Gain Greater Insight into Your SAS® Software with SAS Books.

Discover all that you need on your journey to knowledge and empowerment.

support.sas.com/bookstore
for additional books and resources.
## Contents

Chapter 1. What’s New in SAS Text Miner 13.1 High-Performance Procedures ........................................ 1
Chapter 2. Introduction ....................................................................................................................... 3
Chapter 3. Shared Concepts and Topics .......................................................................................... 7
Chapter 4. The HPTMINE Procedure .............................................................................................. 39
Chapter 5. The HPTMSCORE Procedure ......................................................................................... 79

Subject Index ................................................................................................................................... 89

Syntax Index ..................................................................................................................................... 90
Credits and Acknowledgments

Credits

Documentation

Editing Anne Baxter, Ed Huddleston
Documentation Support Tim Arnold, Melanie Gratton

Software

The SAS high-Performance text mining procedures were implemented by the following members of the development staff. Program development includes design, programming, debugging, support, and documentation. In the following list, the names of the developers who currently provide primary support are listed first; other developers and previous developers are also listed.

HPTMINE Zheng (Alan) Zhao, Russell Albright, David Bultman, Joshua Griffin, James Cox
HPTMSCORE Zheng (Alan) Zhao, Russell Albright, David Bultman, Joshua Griffin, James Cox
High-performance computing foundation Steve E. Krueger
High-performance analytics foundation Robert Cohen, Georges H. Guirguis, Trevor Kearney, Richard Knight, Gang Meng, Oliver Schabenberger, Charles Shorb, Tom P. Weber
Numerical routines Georges H. Guirguis

The following people contribute to the SAS high-Performance text mining procedures with their leadership and support: Saratendu Sethi, Bob Johnson.

Testing

Internationalization Testing

Alex Chai, Mi-Na Chu, Jacky Dong, Feng Gao, Masayuki Iizuka, David Li, Lan Luan, Haiyong Rong, Bin Sun, Frank Wang, Lina Xu, Catherine Yang.

Technical Support

Craig DeVault, Ann Kou.

Acknowledgments

Many people make significant and continuing contributions to the development of SAS software products.

The final responsibility for the SAS System lies with SAS alone. We hope that you will always let us know your opinions about the SAS System and its documentation. It is through your participation that SAS software is continuously improved.

Overview

A keyword has been added in the HPTMINE procedure to control whether document projections are normalized.

Procedure Enhancements

HPTMINE Procedure

You can use the new NONORMDOC keyword in the OUTDOCPRO= option in the SVD statement to control whether document projections are normalized.
Chapter 2
Introduction

Overview of the SAS Text Miner High-Performance procedures

The SAS Text Miner high-performance procedures provide text mining tools that have been specially developed to take advantage of parallel processing in both multithreaded single-machine mode and distributed multiple machine mode. The SAS Text Miner high-performance procedures provide full-spectrum support for text mining, including document parsing, term weighting and filtering, term-by-document matrix creation, dimensionality reduction via singular value decomposition (SVD), and scoring.

In addition to the procedures described in this book, SAS Text Miner includes high-performance utility procedures, which are described in Base SAS Procedures Guide: High-Performance Procedures. You can run all of these procedures in single-machine mode without licensing SAS High-Performance Text Miner. However, to run these procedures in distributed mode, you must license SAS High-Performance Text Miner.

Installation and Configuration

Before you can use the SAS Text Miner high-performance procedures, you need to install and configure the SAS High-Performance Analytics infrastructure. For more information, see the SAS High-Performance Analytics Infrastructure: Installation and Configuration Guide.

Additional installation and configuration is required for SAS Text Miner high-performance procedures. For more information, see the section “System Configuration” on page 61 in Chapter 4, “The HPTMINE Procedure.”
About This Book

This book assumes that you are familiar with Base SAS software and with the books *SAS Language Reference: Concepts* and the *Base SAS Procedures Guide*. It also assumes that you are familiar with basic SAS System concepts, such as using the DATA step to create SAS data sets and using Base SAS procedures (such as, the PRINT and SORT procedures) to manipulate SAS data sets.

Chapter Organization

This book is organized as follows.

Chapter 2, this chapter, provides an overview of SAS Text Miner high-performance procedures.

Chapter 3, “Shared Concepts and Topics,” describes the modes in which SAS Text Miner high-performance procedures can execute.

Subsequent chapters describe the SAS Text Miner high-performance procedures. These chapters appear in alphabetical order by procedure name. Each chapter is organized as follows:

- The “Overview” section provides a brief description of the analysis provided by the procedure.
- The “Getting Started” section provides a quick introduction to the procedure through a simple example.
- The “Syntax” section describes the SAS statements and options that control the procedure.
- The “Details” section discusses methodology and miscellaneous details, such as ODS tables.
- The “Examples” section contains examples that use the procedure.
- The “References” section contains references for the methodology and for examples of the procedure.

Typographical Conventions

This book uses several type styles for presenting information. The following list explains the meaning of the typographical conventions used in this book:

- roman is the standard type style used for most text.
- UPPERCASE ROMAN is used for SAS statements, options, and other SAS language elements when they appear in the text. However, you can enter these elements in your own SAS programs in lowercase, uppercase, or a mixture of the two.
- UPPERCASE BOLD is used in the “Syntax” sections’ initial lists of SAS statements and options.
- oblique is used for user-supplied values for options in the syntax definitions and in text.
- VariableName is used for the names of variables and data sets when they appear in the text.
- bold is used to refer to matrices and vectors.
**Options Used in Examples**

**Output of Examples**

Most of the output shown in this book is produced with the following SAS System options:

```plaintext
options linesize=80 pagesize=500 nonumber nodate;
```

The HTMLBLUE style is used to create the HTML output and graphs that appear in the online documentation. A style template controls stylistic elements such as colors, fonts, and presentation attributes. The style template is specified in the ODS HTML statement as follows:

```plaintext
ods html style=HTMLBlue;
```

If you run the examples, you might get slightly different output. This is a function of the SAS System options used and the precision used by your computer for floating-point calculations.

**Online Documentation**

You can access the documentation by going to [http://support.sas.com/documentation](http://support.sas.com/documentation)

**SAS Technical Support Services**

As with all SAS products, the SAS Technical Support staff is available to respond to problems and answer technical questions regarding the use of the SAS Text Miner high-performance procedures. Go to [http://support.sas.com/techsup](http://support.sas.com/techsup) for more information.
Chapter 3
Shared Concepts and Topics

Contents

Overview ............................................. 7
Processing Modes .................................. 8
  Single-Machine Mode ............................. 8
  Distributed Mode ................................. 8
  Symmetric and Asymmetric Distributed Modes ...... 9
  Controlling the Execution Mode with Environment Variables and Performance Statement Options .......... 9
  Determining Single-Machine Mode or Distributed Mode ............................................. 11
Alongside-the-Database Execution .................. 15
Alongside-LASR Distributed Execution ............ 18
Running High-Performance Analytical Procedures Alongside a SAS LASR Analytic Server in Distributed Mode ............................................. 18
  Starting a SAS LASR Analytic Server Instance ............................................. 18
  Associating a SAS Libref with the SAS LASR Analytic Server Instance ..................... 19
  Running a High-Performance Analytical Procedure Alongside the SAS LASR Analytic Server Instance .......... 19
  Terminating a SAS LASR Analytic Server Instance ............................................. 20
Alongside-LASR Distributed Execution on a Subset of the Appliance Nodes .................... 20
Running High-Performance Analytical Procedures in Asymmetric Mode ......................... 21
  Running in Symmetric Mode ....................... 21
  Running in Asymmetric Mode on One Appliance ............................................. 23
  Running in Asymmetric Mode on Distinct Appliances ............................................. 23
Alongside-HDFS Execution ............................ 26
  Alongside-HDFS Execution by Using the SASHDAT Engine ............................................. 27
  Alongside-HDFS Execution by Using the Hadoop Engine ............................................. 28
Output Data Sets ...................................... 32
Working with Formats ................................ 33
PERFORMANCE Statement .......................... 35

Overview

This chapter describes the modes of execution in which SAS high-performance analytical procedures can execute. If you have SAS Text Miner installed, you can run any procedure in this book on a single machine.
However, to run procedures in this book in distributed mode, you must also have SAS High-Performance Text Miner software installed. For more information about these modes, see the next section.

This chapter provides details of how you can control the modes of execution and includes the syntax for the PERFORMANCE statement, which is common to all high-performance analytical procedures.

### Processing Modes

#### Single-Machine Mode

Single-machine mode is a computing model in which multiple processors or multiple cores are controlled by a single operating system and can access shared resources, such as disks and memory. In this book, single-machine mode refers to an application running multiple concurrent threads on a multicore machine in order to take advantage of parallel execution on multiple processing units. More simply, single-machine mode for high-performance analytical procedures means multithreading on the client machine.

All high-performance analytical procedures are capable of running in single-machine mode, and this is the default mode when a procedure runs on the client machine. The procedure uses the number of CPUs (cores) on the machine to determine the number of concurrent threads. High-performance analytical procedures use different methods to map core count to the number of concurrent threads, depending on the analytic task. Using one thread per core is not uncommon for the procedures that implement data-parallel algorithms.

#### Distributed Mode

Distributed mode is a computing model in which several nodes in a distributed computing environment participate in the calculations. In this book, the distributed mode of a high-performance analytical procedure refers to the procedure performing the analytics on an appliance that consists of a cluster of nodes. This appliance can be one of the following:

- a database management system (DBMS) appliance on which the SAS High-Performance Analytics infrastructure is also installed
- a cluster of nodes that have the SAS High-Performance Analytics infrastructure installed but no DBMS software installed

Distributed mode has several variations:

- Client-data (or local-data) mode: The input data for the analytic task are not stored on the appliance or cluster but are distributed to the distributed computing environment by the SAS High-Performance Analytics infrastructure when the procedure runs.
- Alongside-the-database mode: The data are stored in the distributed database and are read from the DBMS in parallel into a high-performance analytical procedure that runs on the database appliance.
Symmetric and Asymmetric Distributed Modes

- Alongside-HDFS mode: The data are stored in the Hadoop Distributed File System (HDFS) and are read in parallel from the HDFS. This mode is available if you install the SAS High-Performance Deployment of Hadoop on the appliance or when you configure a Cloudera 4 Hadoop deployment on the appliance to operate with the SAS High-Performance Analytics infrastructure. For more information about installing the SAS High-Performance Deployment of Hadoop, see the SAS High-Performance Analytics Infrastructure: Installation and Configuration Guide.

- Alongside-LASR mode: The data are loaded from a SAS LASR Analytic Server that runs on the appliance.

**Symmetric and Asymmetric Distributed Modes**

SAS high-performance analytical procedures can run alongside the database or alongside HDFS in asymmetric mode. The primary reason for providing the asymmetric mode is to enable you to manage and house data on one appliance (the data appliance) and to run the high-performance analytical procedure on a second appliance (the computing appliance). You can also run in asymmetric mode on a single appliance that functions as both the data appliance and the computing appliance. This enables you to run alongside the database or alongside HDFS, where computations are done on a different set of nodes from the nodes that contain the data. The following subsections provide more details.

**Symmetric Mode**

When SAS high-performance analytical procedures run in symmetric distributed mode, the data appliance and the computing appliance must be the same appliance. Both the SAS Embedded Process and the high-performance analytical procedures execute in a SAS process that runs on the same hardware where the DBMS process executes. This is called symmetric mode because the number of nodes on which the DBMS executes is the same as the number of nodes on which the high-performance analytical procedures execute. The initial data movement from the DBMS to the high-performance analytical procedure does not cross node boundaries.

**Asymmetric Mode**

When SAS high-performance analytical procedures run in asymmetric distributed mode, the data appliance and computing appliance are usually distinct appliances. The high-performance analytical procedures execute in a SAS process that runs on the computing appliance. The DBMS and a SAS Embedded Process run on the data appliance. Data are requested by a SAS data feeder that runs on the computing appliance and communicates with the SAS Embedded Process on the data appliance. The SAS Embedded Process transfers the data in parallel to the SAS data feeder that runs on each of the nodes of the computing appliance. This is called asymmetric mode because the number of nodes on the data appliance does not need to be the same as the number of nodes on the computing appliance.

**Controlling the Execution Mode with Environment Variables and Performance Statement Options**

You control the execution mode by using environment variables or by specifying options in the PERFORMANCE statement in high-performance analytical procedures, or by a combination of these methods.
Chapter 3: Shared Concepts and Topics

The important environment variables follow:

- **grid host** identifies the domain name system (DNS) or IP address of the appliance node to which the SAS High-Performance Text Miner software connects to run in distributed mode.

- **installation location** identifies the directory where the SAS High-Performance Text Miner software is installed on the appliance.

- **data server** identifies the database server on Teradata appliances as defined in the hosts file on the client. This data server is the same entry that you usually specify in the SERVER= entry of a LIBNAME statement for Teradata. For more information about specifying LIBNAME statements for Teradata and other engines, see the DBMS-specific section of SAS/ACCESS for Relational Databases: Reference for your engine.

- **grid mode** specifies whether the high-performance analytical procedures execute in symmetric or asymmetric mode. Valid values for this variable are 'sym' for symmetric mode and 'asym' for asymmetric mode. The default is symmetric mode.

You can set an environment variable directly from the SAS program by using the OPTION SET= command. For example, the following statements define three variables for a Teradata appliance (the grid mode is the default symmetric mode):

```sas
option set=GRIDHOST ="hpa.sas.com";
option set=GRIDINSTALLLOC="/opt/TKGrid";
option set=GRIDDATASERVER="myserver";
```

Alternatively, you can set the parameters in the PERFORMANCE statement in high-performance analytical procedures. For example:

```sas
performance host ="hpa.sas.com"
    install ="/opt/TKGrid"
    dataserver="myserver";
```

The following statements define three variables that are needed to run asymmetrically on a computing appliance.

```sas
option set=GRIDHOST ="compute_appliance.sas.com";
option set=GRIDINSTALLLOC="/opt/TKGrid";
option set=GRIDMODE ="asym";
```

Alternatively, you can set the parameters in the PERFORMANCE statement in high-performance analytical procedures. For example:

```sas
performance host ="compute_appliance.sas.com"
    install ="/opt/TKGrid"
    gridmode ="asym"
```

A specification in the PERFORMANCE statement overrides a specification of an environment variable without resetting its value. An environment variable that you set in the SAS session by using an OPTION SET= command remains in effect until it is modified or until the SAS session terminates.
Specifying a data server is necessary only on Teradata systems when you do not explicitly set the GRIDMODE environment variable or specify the GRIDMODE= option in the PERFORMANCE statement. The data server specification depends on the entries in the (client) hosts file. The file specifies the server (suffixed by cop and a number) and an IP address. For example:

```
myservercop1 33.44.55.66
```

The key variable that determines whether a high-performance analytical procedure executes in single-machine or distributed mode is the grid host. The installation location and data server are needed to ensure that a connection to the grid host can be made, given that a host is specified. This book assumes that the installation location and data server (if necessary) have been set by your system administrator.

The following sets of SAS statements are functionally equivalent:

```
proc hpreduce;
  reduce unsupervised x;
  performance host="hpa.sas.com";
run;

option set=GRIDHOST="hpa.sas.com";
proc hpreduce;
  reduce unsupervised x;
run;
```

Determining Single-Machine Mode or Distributed Mode

High-performance analytical procedures use the following rules to determine whether they run in single-machine mode or distributed mode:

- If a grid host is not specified, the analysis is carried out in single-machine mode on the client machine that runs the SAS session.

- If a grid host is specified, the behavior depends on whether the execution is alongside the database or alongside HDFS. If the data are local to the client (that is, not stored in the distributed database or HDFS on the appliance), you need to use the NODES= option in the PERFORMANCE statement to specify the number of nodes on the appliance or cluster that you want to engage in the analysis. If the procedure executes alongside the database or alongside HDFS, you do not need to specify the NODES= option.

The following example shows single-machine and client-data distributed configurations for a data set of 100,000 observations that are simulated from a logistic regression model. The following DATA step generates the data:

```
data simData;
  array _a{8} _temporary_ (0,0,0,1,0,1,1,1);  
  array _b{8} _temporary_ (0,0,1,0,1,0,1,1);  
  array _c{8} _temporary_ (0,1,0,0,1,1,0,1);  
```
do obsno=1 to 100000;
  x = rantbl(1,0.28,0.18,0.14,0.14,0.03,0.09,0.08,0.06);
  a = _a{x};
  b = _b{x};
  c = _c{x};
  x1 = int(ranuni(1)*400);
  x2 = 52 + ranuni(1)*38;
  x3 = ranuni(1)*12;
  lp = 6. -0.015*(1-a) + 0.7*(1-b) + 0.6*(1-c) + 0.02*x1 -0.05*x2 - 0.1*x3;
  y = ranbin(1,1,(1/(1+exp(lp))));
  output;
end;
run;

The following statements run PROC HPLOGISTIC to fit a logistic regression model:

```
proc hplogistic data=simData;
  class a b c;
  model y = a b c x1 x2 x3;
run;
```

Figure 3.1 shows the results from the analysis.

**Figure 3.1 Results from Logistic Regression in Single-Machine Mode**

<table>
<thead>
<tr>
<th>The HPLOGISTIC Procedure</th>
</tr>
</thead>
<tbody>
<tr>
<td>Performance Information</td>
</tr>
<tr>
<td>Execution Mode</td>
</tr>
<tr>
<td>Number of Threads</td>
</tr>
<tr>
<td>Model Information</td>
</tr>
<tr>
<td>Data Source</td>
</tr>
<tr>
<td>Response Variable</td>
</tr>
<tr>
<td>Class Parameterization</td>
</tr>
<tr>
<td>Distribution</td>
</tr>
<tr>
<td>Link Function</td>
</tr>
<tr>
<td>Optimization Technique</td>
</tr>
</tbody>
</table>
The entries in the “Performance Information” table show that the HPLOGISTIC procedure runs in single-machine mode and uses four threads, which are chosen according to the number of CPUs on the client machine. You can force a certain number of threads on any machine that is involved in the computations by specifying the NTHREADS option in the PERFORMANCE statement. Another indication of execution on the client is the following message, which is issued in the SAS log by all high-performance analytical procedures:

NOTE: The HPLOGISTIC procedure is executing on the client.

The following statements use 10 nodes (in distributed mode) to analyze the data on the appliance; results appear in Figure 3.2:

```sas
proc hplogistic data=simData;
   class a b c;
   model y = a b c x1 x2 x3;
   performance host="hpa.sas.com" nodes=10;
run;
```

Figure 3.2 Results from Logistic Regression in Distributed Mode
Figure 3.2 continued

<table>
<thead>
<tr>
<th>Model Information</th>
</tr>
</thead>
<tbody>
<tr>
<td>Data Source: WORK.SIMDATA</td>
</tr>
<tr>
<td>Response Variable: y</td>
</tr>
<tr>
<td>Class Parameterization: GLM</td>
</tr>
<tr>
<td>Distribution: Binary</td>
</tr>
<tr>
<td>Link Function: Logit</td>
</tr>
<tr>
<td>Optimization Technique: Newton-Raphson with Ridging</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Parameter Estimates</th>
</tr>
</thead>
<tbody>
<tr>
<td>Standard Error: [Error]</td>
</tr>
<tr>
<td>DF: [DF]</td>
</tr>
<tr>
<td>t Value: [t Value]</td>
</tr>
<tr>
<td>Pr &gt;</td>
</tr>
</tbody>
</table>

The specification of a host causes the “Performance Information” table to display the name of the host node of the appliance. The “Performance Information” table also indicates that the calculations were performed in a distributed environment on the appliance. Twenty-four threads on each of 10 nodes were used to perform the calculations—for a total of 240 threads.

Another indication of distributed execution on the appliance is the following message, which is issued in the SAS log by all high-performance analytical procedures:

NOTE: The HPLOGISTIC procedure is executing in the distributed computing environment with 10 worker nodes.

You can override the presence of a grid host and force the computations into single-machine mode by specifying the NODES=0 option in the PERFORMANCE statement:

```sas
proc hplogistic data=simData;
  class a b c;
  model y = a b c x1 x2 x3;
  performance host="hpa.sas.com" nodes=0;
run;
```

Figure 3.3 shows the “Performance Information” table. The numeric results are not reproduced here, but they agree with the previous analyses, which are shown in Figure 3.1 and Figure 3.2.
The “Performance Information” table indicates that the HPLOGISTIC procedure executes in single-machine mode on the client. This information is also reported in the following message, which is issued in the SAS log:

NOTE: The HPLOGISTIC procedure is executing on the client.

In the analysis shown previously in Figure 3.2, the data set Work.simData is local to the client, and the HPLOGISTIC procedure distributed the data to 10 nodes on the appliance. The High-Performance Analytics infrastructure does not keep these data on the appliance. When the procedure terminates, the in-memory representation of the input data on the appliance is freed.

When the input data set is large, the time that is spent sending client-side data to the appliance might dominate the execution time. In practice, transfer speeds are usually lower than the theoretical limits of the network connection or disk I/O rates. At a transfer rate of 40 megabytes per second, sending a 10-gigabyte data set to the appliance requires more than four minutes. If analytic execution time is in the range of seconds, the “performance” of the process is dominated by data movement.

The alongside-the-database execution model, unique to high-performance analytical procedures, enables you to read and write data in distributed form from the database that is installed on the appliance.
Before you can run alongside the database, you must distribute the data to the appliance. The following statements use the HPDS2 procedure to distribute the data set Work.simData into the mydb database on the hpa.sas.com appliance. In this example, the appliance houses a Greenplum database.

```sas
option set=GRIDHOST="hpa.sas.com";
libname applianc greenplm
  server    ="hpa.sas.com"
  user      =XXXXXX
  password=YYYYY
  database=mydb;

proc datasets lib=applianc nolist; delete simData;
proc hpds2 data=simData
  out =applianc.simData(distributed_by='distributed randomly');
performance commit=10000 nodes=all;
data DS2GTF.out;
  method run();
  set DS2GTF.in;
end;
enddata;
run;
```

If the output table applianc.simData exists, the DATASETS procedure removes the table from the Greenplum database because a DBMS does not usually support replacement operations on tables.

Note that the libref for the output table points to the appliance. The data set option informs the HPDS2 procedure to distribute the records randomly among the data segments of the appliance. The statements that follow the PERFORMANCE statement are the DS2 program that copies the input data to the output data without further transformations.

Because you loaded the data into a database on the appliance, you can use the following HPLOGISTIC statements to perform the analysis on the appliance in the alongside-the-database mode. These statements are almost identical to the first PROC HPLOGISTIC example in the previous section, which executed in single-machine mode.

```sas
proc hplogistic data=applianc.simData;
  class a b c;
  model y = a b c x1 x2 x3;
run;
```

The subtle differences are as follows:

- The grid host environment variable that you specified in an OPTION SET= command is still in effect.
- The DATA= option in the high-performance analytical procedure uses a libref that identifies the data source as being housed on the appliance. This libref was specified in a prior LIBNAME statement.

Figure 3.4 shows the results from this analysis. The “Performance Information” table shows that the execution was in distributed mode. In this case the execution was alongside the Greenplum database. The numeric results agree with the previous analyses, which are shown in Figure 3.1 and Figure 3.2.
When high-performance analytical procedures execute symmetrically alongside the database, any nonzero specification of the NODES= option in the PERFORMANCE statement is ignored. If the data are read alongside the database, the number of compute nodes is determined by the layout of the database and cannot be modified. In this example, the appliance contains 16 nodes. (See the “Performance Information” table.)

However, when high-performance analytical procedures execute asymmetrically alongside the database, the number of compute nodes that you specify in the PERFORMANCE statement can differ from the number of nodes across which the data are partitioned. For an example, see the section “Running High-Performance Analytical Procedures in Asymmetric Mode” on page 21.
Alongside-LASR Distributed Execution

You can execute high-performance analytical procedures in distributed mode alongside a SAS LASR Analytic Server. When high-performance analytical procedures execute in this mode, the data are preloaded in distributed form in memory that is managed by a LASR Analytic Server. The data on the nodes of the appliance are accessed in parallel in the process that runs the LASR Analytic Server, and they are transferred to the process where the high-performance analytical procedure runs. In general, each high-performance analytical procedure copies the data to memory that persists only while that procedure executes. Hence, when a high-performance analytical procedure runs alongside a LASR Analytic Server, both the high-performance analytical procedure and the LASR Analytic Server have a copy of the subset of the data that is used by the high-performance analytical procedure. The advantage of running high-performance analytical procedures alongside a LASR Analytic Server (as opposed to running alongside a DBMS table or alongside HDFS) is that the initial transfer of data from the LASR Analytic Server to the high-performance analytical procedure is a memory-to-memory operation that is faster than the disk-to-memory operation when the procedure runs alongside a DBMS or HDFS. When the cost of preloading a table into a LASR Analytic Server is amortized by multiple uses of these data in separate runs of high-performance analytical procedures, using the LASR Analytic Server can result in improved performance.

Running High-Performance Analytical Procedures Alongside a SAS LASR Analytic Server in Distributed Mode

This section provides an example of steps that you can use to start and load data into a SAS LASR Analytic Server instance and then run high-performance analytical procedures alongside this LASR Analytic Server instance.

Starting a SAS LASR Analytic Server Instance

The following statements create a SAS LASR Analytic Server instance and load it with the simData data set that is used in the preceding examples. The data that are loaded into the LASR Analytic Server persist in memory across procedure boundaries until these data are explicitly deleted or until the server instance is terminated.

```
proc lasr port=12345
  data=simData
  path="/tmp/"
  performance host="hpa.sas.com" nodes=ALL;
run;
```

The PORT= option specifies a network port number to use. The PATH= option specifies the directory in which the server and table signature files are to be stored. The specified directory must exist on each machine in the cluster. The DATA= option specifies the name of a data set that is loaded into this LASR Analytic Server instance. (You do not need to specify the DATA= option at this time because you can add tables to
the LASR Analytic Server instance at any stage of its life.) For more information about starting and using a LASR Analytic Server, see the *SAS LASR Analytic Server: Administration Guide*.

The NODES=ALL option in the PERFORMANCE statement specifies that the LASR Analytic Server run on all the nodes on the appliance. You can start a LASR Analytic Server on a subset of the nodes on an appliance, but this might affect whether high-performance analytical procedures can run alongside the LASR Analytic Server. For more information, see the section “Alongside-LASR Distributed Execution on a Subset of the Appliance Nodes” on page 20.

Figure 3.5 shows the “Performance Information” table, which shows that the LASR procedure executes in distributed mode on 16 nodes.

**Figure 3.5 Performance Information**

<table>
<thead>
<tr>
<th>The LASR Procedure</th>
<th>Performance Information</th>
</tr>
</thead>
<tbody>
<tr>
<td>Host Node</td>
<td>hpa.sas.com</td>
</tr>
<tr>
<td>Execution Mode</td>
<td>Distributed</td>
</tr>
<tr>
<td>Grid Mode</td>
<td>Symmetric</td>
</tr>
<tr>
<td>Number of Compute Nodes</td>
<td>8</td>
</tr>
</tbody>
</table>

---

**Associating a SAS Libref with the SAS LASR Analytic Server Instance**

The following statements use a LIBNAME statement that associates a SAS libref (named MyLasr) with tables on the server instance as follows:

```
libname MyLasr sasiola port=12345;
```

The SASIOLA option requests that the MyLasr libref use the SASIOLA engine, and the PORT= value associates this libref with the appropriate server instance. For more information about creating a libref that uses the SASIOLA engine, see the *SAS LASR Analytic Server: Administration Guide*.

---

**Running a High-Performance Analytical Procedure Alongside the SAS LASR Analytic Server Instance**

You can use the MyLasr libref to specify the input data for high-performance analytical procedures. You can also create output data sets in the SAS LASR Analytic Server instance by using this libref to request that the output data set be held in memory by the server instance as follows:

```
proc hplogistic data=MyLasr.simData;
   class a b c;
   model y = a b c x1 x2 x3;
   output out=MyLasr.simulateScores pred=PredictedProbabiity;
run;
```
Because you previously specified the GRIDHOST= environment variable and the input data are held in distributed form in the associated server instance, this PROC HPLOGISTIC step runs in distributed mode alongside the LASR Analytic Server, as indicated in the “Performance Information” table shown in Figure 3.6.

**Figure 3.6 Performance Information**

<table>
<thead>
<tr>
<th>Performance Information</th>
</tr>
</thead>
<tbody>
<tr>
<td>Host Node</td>
</tr>
<tr>
<td>Execution Mode</td>
</tr>
<tr>
<td>Grid Mode</td>
</tr>
<tr>
<td>Number of Compute Nodes</td>
</tr>
<tr>
<td>Number of Threads per Node</td>
</tr>
</tbody>
</table>

The preceding OUTPUT statement creates an output table that is added to the LASR Analytic Server instance. Output data sets do not have to be created in the same server instance that holds the input data. You can use a different LASR Analytic Server instance to hold the output data set. However, in order for the output data to be created in alongside mode, all the nodes that are used by the server instance that holds the input data must also be used by the server instance that holds the output data.

**Terminating a SAS LASR Analytic Server Instance**

You can continue to run high-performance analytical procedures and add and delete tables from the SAS LASR Analytic Server instance until you terminate the server instance as follows:

```sas
proc lasr term port=12345;
run;
```

**Alongside-LASR Distributed Execution on a Subset of the Appliance Nodes**

When you run PROC LASR to start a SAS LASR Analytic Server, you can specify the NODES= option in a PERFORMANCE statement to control how many nodes the LASR Analytic Server executes on. Similarly, a high-performance analytical procedure can execute on a subset of the nodes either because you specify the NODES= option in a PERFORMANCE statement or because you run alongside a DBMS or HDFS with an input data set that is distributed on a subset of the nodes on an appliance. In such situations, if a high-performance analytical procedure uses nodes on which the LASR Analytic Server is not running, then running alongside LASR is not supported. You can avoid this issue by specifying the NODES=ALL in the PERFORMANCE statement when you use PROC LASR to start the LASR Analytic Server.
Running High-Performance Analytical Procedures in
Asymmetric Mode

This section provides examples of how you can run high-performance analytical procedures in asymmetric mode. It also includes examples that run in symmetric mode to highlight differences between the modes. For a description of asymmetric mode, see the section “Symmetric and Asymmetric Distributed Modes” on page 9.

Asymmetric mode is commonly used when the data appliance and the computing appliance are distinct appliances. In order to be able to use an appliance as a data provider for high-performance analytical procedures that run in asymmetric mode on another appliance, it is not necessary that SAS High-Performance Text Miner be installed on the data appliance. However, it is essential that a SAS Embedded Process be installed on the data appliance and that SAS High-Performance Text Miner be installed on the computing appliance.

The following examples use a 24-node data appliance named “data_appliance.sas.com,” which houses a Teradata DBMS and has a SAS Embedded Process installed. Because SAS High-Performance Text Miner is also installed on this appliance, it can be used to run high-performance analytical procedures in both symmetric and asymmetric modes.

The following statements load the simData data set of the preceding sections onto the data appliance:

```sas
libname dataLib teradata
   server ="tera2650"
   user =XXXXXX
   password=YYYYY
   database=mydb;

data dataLib.simData;
   set simData;
run;

NOTE: You can provision the appliance with data even if SAS High-Performance Text Miner software is not installed on the appliance.

The following subsections show how you can run the HPLOGISTIC procedure symmetrically and asymmetrically on a single data appliance and asymmetrically on distinct data and computing appliances.

Running in Symmetric Mode

The following statements run the HPLOGISTIC procedure in symmetric mode on the data appliance:

```sas
proc hplogistic data=dataLib.simData;
   class a b c;
   model y = a b c x1 x2 x3;
   performance host = "data_appliance.sas.com"
   nodes = 10
   gridmode = sym;
```
Because you explicitly specified the GRIDMODE= option, you do not need to also specify the DATASERVER= option in the PERFORMANCE statement. Figure 3.7 shows the results of this analysis.

**Figure 3.7** Alongside-the-Database Execution in Symmetric Mode on Teradata

<table>
<thead>
<tr>
<th>The HPLOGISTIC Procedure</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Performance Information</strong></td>
</tr>
<tr>
<td><strong>Host Node</strong></td>
</tr>
<tr>
<td><strong>Execution Mode</strong></td>
</tr>
<tr>
<td><strong>Grid Mode</strong></td>
</tr>
<tr>
<td><strong>Number of Compute Nodes</strong></td>
</tr>
<tr>
<td><strong>Number of Threads per Node</strong></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th><strong>Model Information</strong></th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Data Source</strong></td>
</tr>
<tr>
<td><strong>Response Variable</strong></td>
</tr>
<tr>
<td><strong>Class Parameterization</strong></td>
</tr>
<tr>
<td><strong>Distribution</strong></td>
</tr>
<tr>
<td><strong>Link Function</strong></td>
</tr>
<tr>
<td><strong>Optimization Technique</strong></td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th><strong>Parameter Estimates</strong></th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Parameter</strong></td>
</tr>
<tr>
<td>Intercept</td>
</tr>
<tr>
<td>a 0</td>
</tr>
<tr>
<td>a 1</td>
</tr>
<tr>
<td>b 0</td>
</tr>
<tr>
<td>b 1</td>
</tr>
<tr>
<td>c 0</td>
</tr>
<tr>
<td>c 1</td>
</tr>
<tr>
<td>x1</td>
</tr>
<tr>
<td>x2</td>
</tr>
<tr>
<td>x3</td>
</tr>
</tbody>
</table>

The “Performance Information” table shows that the execution occurs in symmetric mode on the 24 nodes of the data appliance. In this case, the NODES=10 option in the PERFORMANCE statement is ignored because the number of nodes that are used is determined by the number of nodes across which the data are distributed, as indicated in the following warning message in the SAS log:

**WARNING:** The NODES=10 option in the PERFORMANCE statement is ignored because you are running alongside the distributed data source DATALIB.simData.DATA. The number of compute nodes is determined by the configuration of the distributed DBMS.
Running in Asymmetric Mode on One Appliance

You can switch to running the HPLOGISTIC procedure in asymmetric mode by specifying the GRID-MODE=ASYM option in the PERFORMANCE statement as follows:

```
proc hplogistic data=dataLib.simData;
  class a b c;
  model y = a b c x1 x2 x3;
  performance host = "data_appliance.sas.com"
    nodes = 10
    gridmode = asym;
run;
```

Figure 3.8 shows the “Performance Information” table.

<table>
<thead>
<tr>
<th>The HPLOGISTIC Procedure</th>
</tr>
</thead>
<tbody>
<tr>
<td>Performance Information</td>
</tr>
<tr>
<td>Host Node</td>
</tr>
<tr>
<td>Execution Mode</td>
</tr>
<tr>
<td>Grid Mode</td>
</tr>
<tr>
<td>Number of Compute Nodes</td>
</tr>
<tr>
<td>Number of Threads per Node</td>
</tr>
</tbody>
</table>

You can see that now the grid mode is asymmetric. Furthermore, the NODES=10 option that you specified in the PERFORMANCE statement is honored. The data are moved in parallel from the 24 nodes on which the data are stored to the 10 nodes on which the execution occurs. The numeric results are not reproduced here, but they agree with the previous analyses.

Running in Asymmetric Mode on Distinct Appliances

Usually, there is no advantage to executing high-performance analytical procedures in asymmetric mode on one appliance, because data might have to be unnecessarily moved between nodes. The following example demonstrates the more typical use of asymmetric mode. In this example, the specified grid host “compute_appliance.sas.com” is a computing appliance that has 15 compute nodes, and it is a different appliance from the 24-node data appliance “data_appliance.sas.com,” which houses the Teradata DBMS where the data reside.

The advantage of using different computing and data appliances is that the data appliance is not affected by the execution of high-performance analytical procedures except during the initial parallel data transfer. A potential disadvantage of this asymmetric mode of execution is that the performance can be limited by the bandwidth with which data can be moved between the appliances. However, because this data movement takes place in parallel from the nodes of the data appliance to the nodes of the computing appliance, this
potential performance bottleneck can be overcome with appropriately provisioned hardware. The following statements show how this is done:

```sas
proc hplogistic data=dataLib.simData;
class a b c;
model y = a b c x1 x2 x3;
performance host = "compute_appliance.sas.com"
    gridmode = asym;
run;
```

Figure 3.9 shows the “Performance Information” table.

![Figure 3.9 Asymmetric Mode with Distinct Data and Computing Appliances](image)

The HPLOGISTIC Procedure

**Performance Information**

<table>
<thead>
<tr>
<th>Host Node</th>
<th>compute_appliance.sas.com</th>
</tr>
</thead>
<tbody>
<tr>
<td>Execution Mode</td>
<td>Distributed</td>
</tr>
<tr>
<td>Grid Mode</td>
<td>Asymmetric</td>
</tr>
<tr>
<td>Number of Compute Nodes</td>
<td>15</td>
</tr>
<tr>
<td>Number of Threads per Node</td>
<td>24</td>
</tr>
</tbody>
</table>

PROC HPLOGISTIC ran on the 15 nodes of the computing appliance, even though the data are partitioned across the 24 nodes of the data appliance. The numeric results are not reproduced here, but they agree with the previous analyses shown in Figure 3.1 and Figure 3.2.

Every time you run a high-performance analytical procedure in asymmetric mode that uses different computing and data appliances, data are transferred between these appliances. If you plan to make repeated use of the same data, then it might be advantageous to temporarily persist the data that you need on the computing appliance. One way to persist the data is to store them as a table in a SAS LASR Analytic Server that runs on the computing appliance. By running PROC LASR in asymmetric mode, you can load the data in parallel from the data appliance nodes to the nodes on which the LASR Analytic Server runs on the computing appliance. You can then use a LIBNAME statement that associates a SAS libref with tables on the LASR Analytic Server. The following statements show how you do this:

```sas
proc lasr port=54321
    data=dataLib.simData
    path="/tmp/";
performance host = "compute_appliance.sas.com"
    gridmode = asym;
run;
libname MyLasr sasiola tag="dataLib" port=54321 host="compute_appliance.sas.com" ;
```
PROC LASR ran in asymmetric mode on the computing appliance, which has 15 compute nodes. In this mode, the data are loaded in parallel from the 24 data appliance nodes to the 15 compute nodes on the computing appliance. By default, all the nodes on the computing appliance are used. You can use the NODES= option in the PERFORMANCE statement to run the LASR Analytic Server on a subset of the nodes on the computing appliance. If you omit the GRIDMODE=ASYM option from the PERFORMANCE statement, PROC LASR still runs successfully but much less efficiently. The Teradata access engine transfers the simData data set to a temporary table on the client, and the High-Performance Analytics infrastructure then transfers these data from the temporary table on the client to the grid nodes on the computing appliance.

After the data are loaded into a LASR Analytic Server that runs on the computing appliance, you can run high-performance analytical procedures alongside this LASR Analytic Server. Because these procedures run on the same computing appliance where the LASR Analytic Server is running, it is best to run these procedures in symmetric mode, which is the default or can be explicitly specified in the GRIDMODE=SYM option in the PERFORMANCE statement. The following statements provide an example. The OUTPUT statement creates an output data set that is held in memory by the LASR Analytic Server. The data appliance has no role in executing these statements.

```sas
proc hplogistic data=MyLasr.simData;
  class a b c;
  model y = a b c x1 x2 x3;
  output out=MyLasr.myOutputData pred=myPred;
  performance host = "compute_appliance.sas.com";
run;
```

The following note, which appears in the SAS log, confirms that the output data set is created successfully:

```
NOTE: The table DATALIB.MYOUTPUTDATA has been added to the LASR Analytic Server with port 54321. The Libname is MYLASR.
```

You can use the dataLib libref that you used to load the data onto the data appliance to create an output data set on the data appliance. In order for this output to be directly written in parallel from the nodes of the computing appliance to the nodes of the data appliance, you need to run the HPLOGISTIC procedure in asymmetric mode by specifying the GRIDMODE=ASYM option in the PERFORMANCE statement as follows:
proc hplogistic data=MyLasr.simData;
  class a b c;
  model y = a b c x1 x2 x3;
  output out=dataLib.myOutputData pred=myPred;
  performance host = "compute_appliance.sas.com"
                    gridmode = asym;
run;

The following note, which appears in the SAS log, confirms that the output data set is created successfully on the data appliance:

NOTE: The data set DATALIB.myOutputData has 100000 observations and 1 variables.

When you run a high-performance analytical procedure on a computing appliance and either read data from or write data to a different data appliance, it is important to run the high-performance analytical procedures in asymmetric mode so that the Read and Write operations take place in parallel without any movement of data to and from the SAS client. If you omit running the preceding PROC HPLOGISTIC step in asymmetric mode, then the output data set would be created much less efficiently: the output data would be moved sequentially to a temporary table on the client, after which the Teradata access engine sequentially would write this table to the data appliance.

When you no longer need the data in the SAS LASR Analytic Server, you should terminate the server instance as follows:

proc lasr term port=54321;
  performance host="compute_appliance.sas.com";
run;

If you configured Hadoop on the computing appliance, then you can create output data tables that are stored in the HDFS on the computing appliance. You can do this by using the SASHDAT engine as described in the section “Alongside-HDFS Execution” on page 26.

---

Alongside-HDFS Execution

Running high-performance analytical procedures alongside HDFS shares many features with running alongside the database. You can execute high-performance analytical procedures alongside HDFS by using either the SASHDAT engine or the Hadoop engine.

You use the SASHDAT engine to read and write data that are stored in HDFS in a proprietary SASHDAT format. In SASHDAT format, metadata that describe the data in the Hadoop files are included with the data. This enables you to access files in SASHDAT format without supplying any additional metadata. Additionally, you can also use the SASHDAT engine to read data in CSV (comma-separated value) format, but you need supply metadata that describe the contents of the CSV data. The SASHDAT engine provides highly optimized access to data in HDFS that are stored in SASHDAT format.

The Hadoop engine reads data that are stored in various formats from HDFS and writes data to HDFS in CSV format. This engine can use metadata that are stored in Hive, which is a data warehouse that supplies
metadata about data that are stored in Hadoop files. In addition, this engine can use metadata that you create by using the HDMD procedure.

The following subsections provide details about using the SASHDAT and Hadoop engines to execute high-performance analytical procedures alongside HDFS.

---

**Alongside-HDFS Execution by Using the SASHDAT Engine**

If the grid host is a cluster that houses data that have been distributed by using the SASHDAT engine, then high-performance analytical procedures can analyze those data in the alongside-HDFS mode. The procedures use the distributed computing environment in which an analytic process is co-located with the nodes of the cluster. Data then pass from HDFS to the analytic process on each node of the cluster.

Before you can run a procedure alongside HDFS, you must distribute the data to the cluster. The following statements use the SASHDAT engine to distribute to HDFS the simData data set that was used in the previous two sections:

```sas
option set=GRIDHOST="hpa.sas.com";
libname hdatLib sashdat
    path="/hps";

data hdatLib.simData (replace = yes) ;
    set simData;
run;
```

In this example, the GRIDHOST is a cluster where the SAS Data in HDFS Engine is installed. If a data set that is named `simData` already exists in the `hps` directory in HDFS, it is overwritten because the REPLACE=YES data set option is specified. For more information about using this LIBNAME statement, see the section “LIBNAME Statement for the SAS Data in HDFS Engine” in the *SAS LASR Analytic Server: Administration Guide*.

The following HPLOGISTIC procedure statements perform the analysis in alongside-HDFS mode. These statements are almost identical to the PROC HPLOGISTIC example in the previous two sections, which executed in single-machine mode and alongside-the-database distributed mode, respectively.

```sas
proc hplogistic data=hdatLib.simData;
    class a b c;
    model y = a b c x1 x2 x3;
run;
```

*Figure 3.11* shows the “Performance Information” table. You see that the procedure ran in distributed mode. The numeric results shown in *Figure 3.12* agree with the previous analyses shown in *Figure 3.1*, *Figure 3.2*, and *Figure 3.4*. 
Figure 3.11 Alongside-HDFS Execution Performance Information

<table>
<thead>
<tr>
<th>Performance Information</th>
</tr>
</thead>
<tbody>
<tr>
<td>Host Node</td>
</tr>
<tr>
<td>Execution Mode</td>
</tr>
<tr>
<td>Grid Mode</td>
</tr>
<tr>
<td>Number of Compute Nodes</td>
</tr>
<tr>
<td>Number of Threads per Node</td>
</tr>
</tbody>
</table>

Figure 3.12 Alongside-HDFS Execution Model Information

<table>
<thead>
<tr>
<th>Model Information</th>
</tr>
</thead>
<tbody>
<tr>
<td>Data Source</td>
</tr>
<tr>
<td>Response Variable</td>
</tr>
<tr>
<td>Class Parameterization</td>
</tr>
<tr>
<td>Distribution</td>
</tr>
<tr>
<td>Link Function</td>
</tr>
<tr>
<td>Optimization Technique</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th>Parameter Estimates</th>
</tr>
</thead>
<tbody>
<tr>
<td>Parameter</td>
</tr>
<tr>
<td>Intercept</td>
</tr>
<tr>
<td>a 0</td>
</tr>
<tr>
<td>a 1</td>
</tr>
<tr>
<td>b 0</td>
</tr>
<tr>
<td>b 1</td>
</tr>
<tr>
<td>c 0</td>
</tr>
<tr>
<td>c 1</td>
</tr>
<tr>
<td>x1</td>
</tr>
<tr>
<td>x2</td>
</tr>
<tr>
<td>x3</td>
</tr>
</tbody>
</table>

Alongside-HDFS Execution by Using the Hadoop Engine

The following LIBNAME statement sets up a libref that you can use to access data that are stored in HDFS and have metadata in Hive:

```plaintext
libname hdoopLib hadoop
    server = "hpa.sas.com"
    user = XXXXX
    password = YYYYY
    database = myDB
    config = "demo.xml" ;
```
For more information about LIBNAME options available for the Hadoop engine, see the LIBNAME topic in the Hadoop section of *SAS/ACCESS for Relational Databases: Reference*. The configuration file that you specify in the CONFIG= option contains information that is needed to access the Hive server. It also contains information that enables this configuration file to be used to access data in HDFS without using the Hive server. This information can also be used to specify replication factors and block sizes that are used when the engine writes data to HDFS. The following XML shows the contents of the file `demo.xml` that is used in this example:

```xml
<configuration>
  <property>
    <name>fs.default.name</name>
    <value>hdfs://hpa.sas.com:8020</value>
  </property>
  <property>
    <name>mapred.job.tracker</name>
    <value>hpa.sas.com:8021</value>
  </property>
  <property>
    <name>dfs.replication</name>
    <value>1</value>
  </property>
  <property>
    <name>dfs.block.size</name>
    <value>33554432</value>
  </property>
</configuration>
```

The following DATA step uses the Hadoop engine to distribute to HDFS the `simData` data set that was used in the previous sections. The engine creates metadata for the data set in Hive.

```sas
data hdoopLib.simData;
  set simData;
run;
```

After you have loaded data or if you are accessing preexisting data in HDFS that have metadata in Hive, you can access this data alongside HDFS by using high-performance analytics procedures. The following `HPLOGISTIC` procedure statements perform the analysis in alongside-HDFS mode. These statements are similar to the `PROC HPLOGISTIC` example in the previous sections. However, whenever you use the Hadoop engine, you must execute the analysis in asymmetric mode to cause the execution to occur alongside HDFS.

```sas
proc hplogistic data=hdoopLib.simData;
  class a b c;
  model y = a b c x1 x2 x3;
  performance host = "compute_appliance.sas.com"
    gridmode = asym;
run;
```

*Figure 3.13 shows the “Performance Information” table. You see that the procedure ran asymmetrically in distributed mode. The numeric results shown in Figure 3.14 agree with the previous analyses.*
The Hadoop engine also enables you to access tables in HDFS that are stored in various formats and that are not registered in Hive. You can use the HDMD procedure to generate metadata for tables that are stored in the following file formats:

- delimited text
- fixed-record length binary
- sequence files
- XML text
To read any other kind of file in Hadoop, you can write a custom file reader plug-in in Java for use with PROC HDMD. For more information about LIBNAME options available for the Hadoop engine, see the LIBNAME topic in the Hadoop section of SAS/ACCESS for Relational Databases: Reference.

The following example shows how you can use PROC HDMD to register metadata for CSV data independently from Hive and then analyze these data by using high-performance analytics procedures. The CSV data in the table csvExample.csv is stored in HDFS in the directory /user/demo/data. Each record in this table consists of the following fields, in the order shown and separated by commas.

1. a string of at most six characters
2. a numeric field with values of 0 or 1
3. a numeric field with real numbers

Suppose you want to fit a logistic regression model to these data, where the second field represents a target variable named Success, the third field represents a regressor named Dose, and the first field represents a classification variable named Group.

The first step is to use PROC HDMD to create metadata that are needed to interpret the table, as in the following statements:

```sas
libname hdoopLib hadoop
server = "hpa.sas.com"
user = XXXXX
password = YYYYY
HDFS_PERMDIR = "/user/demo/data"
HDFS_METADIR = "/user/demo/meta"
config = "demo.xml"
DBCREATE_TABLE_EXTERNAL=YES;
proc hdmd name=hdoopLib.csvExample data_file='csvExample.csv'
    format=delimited encoding=utf8 sep = ',';
    column Group char(6);
    column Success double;
    column Dose double;
run;
```

The metadata that are created by PROC HDMD for this table are stored in the directory /user/demo/meta that you specified in the HDFS_METADIR = option in the preceding LIBNAME statement. After you create the metadata, you can execute high-performance analytics procedures with these data by using the hdoopLib libref. For example, the following statements fit a logistic regression model to the CSV data that are stored in csvExample.csv table.

```sas
proc hplogistic data=hdoopLib.csvExample;
    class Group;
    model Success = Dose;
    performance host = "compute_appliance.sas.com"
        gridmode = asym;
run;
```
Figure 3.15 shows the results of this analysis. You see that the procedure ran asymmetrically in distributed mode. The metadata that you created by using the HDMD procedure have been used successfully in executing this analysis.

**Figure 3.15** Alongside-HDFS Execution with CSV Data

```
The HPLOGISTIC Procedure

Performance Information

Host Node                  compute_appliance.sas.com
Execution Mode             Distributed
Grid Mode                  Asymmetric
Number of Compute Nodes    15
Number of Threads per Node 24

Model Information

Data Source                GRIDLIB.CSVEXAMPLE
Response Variable          Success
Class Parameterization     GLM
Distribution               Binary
Link Function              Logit
Optimization Technique     Newton-Raphson with Ridging

Class Level Information

Class       Levels     Values
Group       3           group1 group2 group3

Number of Observations Read 1000
Number of Observations Used 1000

Parameter Estimates

| Parameter | Estimate | Error   | DF   | t Value | Pr > |t| |
|-----------|----------|---------|------|---------|------|-------|
| Intercept | 0.1243   | 0.1295  | Infty| 0.96    | 0.3371|
| Dose      | -0.2674  | 0.2216  | Infty| -1.21   | 0.2277|
```

**Output Data Sets**

In the alongside-the-database mode, the data are read in distributed form, minimizing data movement for best performance. Similarly, when you write output data sets and a high-performance analytical procedure executes in distributed mode, the data can be written in parallel into the database.
For example, in the following statements, the HPLOGISTIC procedure executes in distributed mode by using eight nodes on the appliance to perform the logistic regression on work.simData:

```
proc hplogistic data=simData;
  class a b c;
  model y = a b c x1 x2 x3;
  id a;
  output out=applianc.simData_out pred=p;
  performance host="hpa.sas.com" nodes=8;
run;
```

The output data set applianc.simData_out is written in parallel into the database. Although the data are fed on eight nodes, the database might distribute the data on more nodes.

When a high-performance analytical procedure executes in single-machine mode, all output objects are created on the client. If the libref of the output data sets points to the appliance, the data are transferred to the database on the appliance. This can lead to considerable performance degradation compared to execution in distributed mode.

Many procedures in SAS software add the variables from the input data set when an observationwise output data set is created. The assumption of high-performance analytical procedures is that the input data sets can be large and contain many variables. For performance reasons, the output data set contains the following:

- variables that are explicitly created by the statement
- variables that are listed in the ID statement
- distribution keys or hash keys that are transferred from the input data set

Including this information enables you to add to the output data set information necessary for subsequent SQL joins without copying the entire input data set to the output data set.

---

**Working with Formats**

You can use SAS formats and user-defined formats with high-performance analytical procedures as you can with other procedures in the SAS System. However, because the analytic work is carried out in a distributed environment and might depend on the formatted values of variables, some special handling can improve the efficiency of work with formats.

High-performance analytical procedures examine the variables that are used in an analysis for association with user-defined formats. Any user-defined formats that are found by a procedure are transmitted automatically to the appliance. If you are running multiple high-performance analytical procedures in a SAS session and the analysis variables depend on user-defined formats, you can preprocess the formats. This step involves generating an XML stream (a file) of the formats and passing the stream to the high-performance analytical procedures.
Suppose that the following formats are defined in your SAS program:

```sas
proc format;
  value YesNo    1='Yes'    0='No';
  value checkThis 1='ThisisOne' 2='ThisisTwo';
  value $cityChar 1='Portage' 2='Kinston';
run;
```

The next group of SAS statements create the XML stream for the formats in the file `Myfmt.xml`, associate that file with the file reference `myxml`, and pass the file reference with the `FMTLIBXML=` option in the `PROC HPLOGISTIC` statement:

```sas
filename myxml 'Myfmt.xml';
libname myxml XML92 xmltype=sasfmt tagset=tagsets.XMLsuv;
proc format cntlout=myxml.allfmts;
run;

proc hplogistic data=six fmtlibxml=myxml;
  class wheeze cit age;
  format wheeze best4. cit $cityChar.;
  model wheeze = cit age;
run;
```

Generation and destruction of the stream can be wrapped in convenience macros:

```sas
%macro Make_XMLStream(name=tempxml);
  filename &name 'fmt.xml';
  libname &name XML92 xmltype=sasfmt tagset=tagsets.XMLsuv;
  proc format cntlout=&name..allfmts;
  run;
%mend;

%macro Delete_XMLStream(fref);
  %let rc=%sysfunc(fdelete(&fref));
%mend;
```

If you do not pass an XML stream to a high-performance analytical procedure that supports the `FMTLIBXML=` option, the procedure generates an XML stream as needed when it is invoked.
PERFORMANCE Statement

**PERFORMANCE** `<performance-options>`;

The PERFORMANCE statement defines performance parameters for multithreaded and distributed computing, passes variables that describe the distributed computing environment, and requests detailed results about the performance characteristics of a high-performance analytical procedure.

You can also use the PERFORMANCE statement to control whether a high-performance analytical procedure executes in single-machine or distributed mode.

You can specify the following `performance-options` in the PERFORMANCE statement:

**COMMIT=n**
requests that the high-performance analytical procedure write periodic updates to the SAS log when observations are sent from the client to the appliance for distributed processing.

High-performance analytical procedures do not have to use input data that are stored on the appliance. You can perform distributed computations regardless of the origin or format of the input data, provided that the data are in a format that can be read by the SAS System (for example, because a SAS/ACCESS engine is available).

In the following example, the HPREG procedure performs LASSO variable selection where the input data set is stored on the client:

```sas
proc hpreg data=work.one;
    model y = x1-x500;
    selection method=lasso;
    performance nodes=10 host='mydca' commit=10000;
run;
```

In order to perform the work as requested using 10 nodes on the appliance, the data set Work.One needs to be distributed to the appliance.

High-performance analytical procedures send the data in blocks to the appliance. Whenever the number of observations sent exceeds an integer multiple of the `COMMIT=` size, a SAS log message is produced. The message indicates the actual number of observations distributed, and not an integer multiple of the `COMMIT=` size.

**DATASERVER=**"name"

specifies the name of the server on Teradata systems as defined through the *hosts* file and as used in the LIBNAME statement for Teradata. For example, assume that the *hosts* file defines the server for Teradata as follows:

```plaintext
myservercop1  33.44.55.66
```

Then a LIBNAME specification would be as follows:
libname TDLib teradata server=myserver user= password= database= ;

A PERFORMANCE statement to induce running alongside the Teradata server would specify the following:

```
performance dataserver="myserver";
```

The DATASERVER= option is not required if you specify the GRIDMODE=option in the PERFORMANCE statement or if you set the GRIDMODE environment variable.

Specifying the DATASERVER= option overrides the GRIDDATASERVER environment variable.

DETAILS
requests a table that shows a timing breakdown of the procedure steps.

GRIDHOST="name"
HOST="name"
specifies the name of the appliance host in single or double quotation marks. If this option is specified, it overrides the value of the GRIDHOST environment variable.

GRIDMODE=SYM | ASYM
MODE=SYM | ASYM
specifies whether the high-performance analytical procedure runs in symmetric (SYM) mode or asymmetric (ASYM) mode. The default is GRIDMODE=SYM. For more information about these modes, see the section “Symmetric and Asymmetric Distributed Modes” on page 9.

If this option is specified, it overrides the GRIDMODE environment variable.

GRIDTIMEOUT=s
TIMEOUT=s
specifies the time-out in seconds for a high-performance analytical procedure to wait for a connection to the appliance and establish a connection back to the client. The default is 120 seconds. If jobs are submitted to the appliance through workload management tools that might suspend access to the appliance for a longer period, you might want to increase the time-out value.

INSTALL="name"
INSTALLLOC="name"
specifies the directory in which the shared libraries for the high-performance analytical procedure are installed on the appliance. Specifying the INSTALL= option overrides the GRIDINSTALLLOC environment variable.

LASRSERVER="path"
LASR="path"
specifies the fully qualified path to the description file of a SAS LASR Analytic Server instance. If the input data set is held in memory by this LASR Analytic Server instance, then the procedure runs alongside LASR. This option is not needed to run alongside LASR if the DATA= specification of the input data uses a libref that is associated with a LASR Analytic Server instance. For more information, see the section “Alongside-LASR Distributed Execution” on page 18 and the SAS LASR Analytic Server: Administration Guide.
NODES=ALL | n
NNODES=ALL | n

specifies the number of nodes in the distributed computing environment, provided that the data are not processed alongside the database.

Specifying NODES=0 indicates that you want to process the data in single-machine mode on the client machine. If the input data are not alongside the database, this is the default. The high-performance analytical procedures then perform the analysis on the client. For example, the following sets of statements are equivalent:

```sas
proc hplogistic data=one;
  model y = x;
run;
```

```sas
proc hplogistic data=one;
  model y = x;
  performance nodes=0;
run;
```

If the data are not read alongside the database, the NODES= option specifies the number of nodes on the appliance that are involved in the analysis. For example, the following statements perform the analysis in distributed mode by using 10 units of work on the appliance that is identified in the HOST= option:

```sas
proc hplogistic data=one;
  model y = x;
  performance nodes=10 host="hpa.sas.com";
run;
```

If the number of nodes can be modified by the application, you can specify a NODES=n option, where n exceeds the number of physical nodes on the appliance. The SAS High-Performance Text Miner software then oversubscribes the nodes and associates nodes with multiple units of work. For example, on a system that has 16 appliance nodes, the following statements oversubscribe the system by a factor of 3:

```sas
proc hplogistic data=one;
  model y = x;
  performance nodes=48 host="hpa.sas.com";
run;
```
Usually, it is not advisable to oversubscribe the system because the analytic code is optimized for a certain level of multithreading on the nodes that depends on the CPU count. You can specify NODES=ALL if you want to use all available nodes on the appliance without oversubscribing the system.

If the data are read alongside the distributed database on the appliance, specifying a nonzero value for the NODES= option has no effect. The number of units of work in the distributed computing environment is then determined by the distribution of the data and cannot be altered. For example, if you are running alongside an appliance with 24 nodes, the NODES= option in the following statements is ignored:

```plaintext
libname GPLib greenplm server=gpdca user=XXX password=YYY
database=ZZZ;
proc hplogistic data=gplib.one;
  model y = x;
  performance nodes=10 host="hpa.sas.com";
run;
```

**NTHREADS=n**

**THREADS=n**

specifies the number of threads for analytic computations and overrides the SAS system option THREADS | NOTHREADS. If you do not specify the NTHREADS= option, the number of threads is determined based on the number of CPUs on the host on which the analytic computations execute. The algorithm by which a CPU count is converted to a thread count is specific to the high-performance analytical procedure. Most procedures create one thread per CPU for the analytic computations.

By default, high-performance analytical procedures execute in multiple concurrent threads unless multithreading has been turned off by the NOTHREADS system option or you force single-threaded execution by specifying NTHREADS=1. The largest number that can be specified for \( n \) is 256. Individual high-performance analytical procedures can impose more stringent limits if called for by algorithmic considerations.

**Note:** The SAS system options THREADS | NOTHREADS apply to the client machine on which the SAS high-performance analytical procedures execute. They do not apply to the compute nodes in a distributed environment.
Chapter 4
The HPTMINE Procedure

Contents

Overview: HPTMINE Procedure .................................................. 40
PROC HPTMINE Features ....................................................... 40
PROC HPTMINE Contrasted with Other SAS Procedures ....................... 41
Getting Started: HPTMINE Procedure ....................................... 41
Syntax: HPTMINE Procedure .................................................. 46
  PROC HPTMINE Statement .................................................. 46
  DOC_ID Statement ........................................................... 47
  PARSE Statement ............................................................ 47
  PERFORMANCE Statement .................................................... 50
  SVD Statement ............................................................... 50
  TARGET Statement ........................................................... 52
  VARIABLE Statement .......................................................... 52
Details: HPTMINE Procedure .................................................. 52
  Natural Language Processing .................................................. 53
    Stemming ................................................................. 53
    Part-of-Speech Tagging ...................................................... 53
    Noun Group Extraction ..................................................... 54
    Entity Identification ....................................................... 54
    Multiword Terms Handling .................................................. 54
    Language Support .......................................................... 54
  Term and Cell Weighting ..................................................... 55
  Singular Value Decomposition ................................................ 55
    Applications in Text Mining ................................................ 56
    Computation ............................................................... 56
Output Data Sets ............................................................... 57
  The OUTCONFIG= Data Set .................................................. 57
  The OUTDOCPRO= Data Set .................................................. 58
  The OUTPARENT= Data Set ................................................... 58
  The OUTCHILD= Data Set .................................................... 59
  The OUTTERMS= Data Set .................................................... 59
Displayed Output .............................................................. 61
  Performance Information .................................................... 61
  Procedure Task Timing ........................................................ 61
ODS Table Names ............................................................... 61
System Configuration ........................................................... 61
Prerequisites for Running PROC HPTMINE .................................. 61
Overview: HPTMINE Procedure

The HPTMINE procedure is a high-performance procedure that analyzes large-scale textual data. PROC HPTMINE provides an essential capability for high-performance text mining and supports a wide range of fundamental text analysis features, which include tokenizing, stemming, part-of-speech tagging, noun group extraction, default or customized stop or start lists, entity parsing, multiword tokens, synonym lists, term weighting, term-by-document matrix creation, and dimension reduction by term filtering and singular value decomposition (SVD).

PROC HPTMINE integrates the functionalities that are provided by the TGPARSE, TMUTIL, and SPSVD procedures from SAS Text Miner, and achieves high efficiency and scalability through parallel processing. The HPTMINE procedure can also generate results that can be used to facilitate scoring by the HPTMSCORE procedure. These results include a configuration data set, a term data set, and a data set that contains the SVD projection matrix.

You can use the HPTMINE procedure to read data in distributed form and perform text analysis in parallel in single-machine mode or distributed mode. For more information about how to configure the execution mode of SAS high-performance analytical procedures, see the section “Processing Modes” on page 8 in Chapter 3, “Shared Concepts and Topics.”

NOTE: Distributed mode requires SAS High-Performance Text Miner.

PROC HPTMINE Features

The HPTMINE procedure processes large-scale textual data in parallel to achieve efficiency and scalability. The following list summarizes the basic features of PROC HPTMINE:

- Functionalities that are related to document parsing, term-by-document matrix creation, and dimension reduction are integrated into one procedure to process data more efficiently.
• Parsing supports essential natural language processing (NLP) features, which include tokenizing, stemming, part-of-speech tagging, noun group extraction, default or customized stop or start lists, entity parsing, multiword tokens, synonym lists.

• Term weighting and filtering are supported for term-by-document matrix creation.

• Parsing and term-by-document matrix creation are processed in parallel.

• Computation of singular value decomposition (SVD) is parallelized.

• Analysis can be performed on a massively parallel SAS high-performance appliance.

• All phases of processing make use of a high degree of multithreading.

• Input data can be read in parallel when the data source is the appliance database.

---

**PROC HPTMINE Contrasted with Other SAS Procedures**

The following remarks compare the HPTMINE procedure with the TGPARSE, TMUTIL, and SPSVD procedures in SAS Text Miner software.

PROC HPTMINE maximizes performance by combining the key functionalities that are found in the TGPARSE, TMUTIL, and SPSVD procedures. By functioning as an “all-in-one” procedure, PROC HPTMINE avoids expensive I/O operations. The PARSE option in the HPTMINE procedure effectively replaces the necessary options that are provided by specifying PROC TGPARSE followed by PROC TMUTIL in the conventional SAS environment. You can use the PARSE option to control parsing, accumulation, and ultimately the construction of the underlying weighted term-document frequency matrix. This matrix serves as input for singular value decomposition, which is controlled by the SVD statement in the HPTMINE procedure.

PROC HPTMINE’s functionality also replaces many of the core functions that were available in the SPSVD procedure. Both PROC HPTMINE and PROC SPSVD produce an output that holds the reduced dimensional representation for the input documents. They also generate an output U matrix that can be used to project new documents in this same reduced space during scoring.

---

**Getting Started: HPTMINE Procedure**

The following DATA step contains 36 observations that have two variables. The text variable contains the input documents, and the did variable contains the ID of the documents. Each row in the data set represents a document for analysis.

```sas
data getstart;
  infile cards delimiter='|' missover;
  length text $150;
  input text$ did$;
  cards;
  High-performance analytics hold the key to |d01
```

```sas
  High-performance analytics hold the key to |d01
```
unlocking the unprecedented business value of big data. Organizations looking for optimal ways to gain insights from big data in shorter reporting windows are turning to SAS. As the gold-standard leader in business analytics for more than 36 years, SAS frees enterprises from the limitations of traditional computing and enables them to draw instant benefits from big data. Faster Time to Insight. From banking to retail to health care to insurance, SAS is helping industries glean insights from data that once took days or weeks in just hours, minutes or seconds. It's all about getting to and analyzing relevant data faster. Revealing previously unseen patterns, sentiments and relationships. Identifying unknown risks. And speeding the time to insights. High-Performance Analytics from SAS Combining industry-leading analytics software with high-performance computing technologies produces fast and precise answers to unsolvable problems and enables our customers to gain greater competitive advantage. SAS In-Memory Analytics eliminate the need for disk-based processing allowing for much faster analysis. SAS In-Database executes analytic logic into the database itself for improved agility and governance. SAS Grid Computing creates a centrally managed, shared environment for processing large jobs and supporting a growing number of users efficiently. Together, the components of this integrated, supercharged platform are changing the decision-making landscape and redefining how the world solves big data business problems. Big data is a popular term used to describe the exponential growth, availability and use of information, both structured and unstructured. Much has been written on the big data trend and how it can serve as the basis for innovation, differentiation and growth.

The following statements use singular value decomposition to parse the input text data and generate a lower-dimensional representation of the documents. The statements specify that only the terms that appear at least twice in the document collection be kept for generating the term-by-document matrix. The summary information about the terms in the document collection is stored in a SAS data set named terms. The SVD statement requests that the top 10 singular values and singular vectors be computed. The projection of the documents is stored in a SAS data set named docpro.

```
proc hptmine data=getstart;
  doc_id did;
  variable text;
  parse
    outterms = terms
    reducef = 2;
  svd
    k = 10
    outdocpro = docpro;
  performance details;
run;
```
The output from this analysis is presented in Figure 4.1 through Figure 4.5.

Figure 4.1 shows the “Performance Information” table, which indicates that PROC HPTMINE executes in single-machine mode. That is, PROC HPTMINE runs on the machine where the SAS system is running. The table also shows that two threads are used for computing.

**Figure 4.1** Performance Information

<table>
<thead>
<tr>
<th>Performance Information</th>
</tr>
</thead>
<tbody>
<tr>
<td>Execution Mode</td>
</tr>
<tr>
<td>Number of Threads</td>
</tr>
</tbody>
</table>

Figure 4.2 shows the “Procedure Task Timing” table, which provides details about how much time is used by each processing step.

**Figure 4.2** Procedure Task Timing

<table>
<thead>
<tr>
<th>Procedure Task Timing</th>
<th>Time (sec.)</th>
<th>%</th>
</tr>
</thead>
<tbody>
<tr>
<td>Parse documents</td>
<td>0.92</td>
<td>90.9%</td>
</tr>
<tr>
<td>Analyze terms</td>
<td>0.00</td>
<td>0.03%</td>
</tr>
<tr>
<td>Obtain term frequency</td>
<td>0.03</td>
<td>3.20%</td>
</tr>
<tr>
<td>Filter terms</td>
<td>0.02</td>
<td>1.61%</td>
</tr>
<tr>
<td>Generate term-document matrix</td>
<td>0.01</td>
<td>1.40%</td>
</tr>
<tr>
<td>Output OUTTERMS table</td>
<td>0.01</td>
<td>1.41%</td>
</tr>
<tr>
<td>Compute SVD</td>
<td>0.01</td>
<td>1.14%</td>
</tr>
<tr>
<td>Output OUTDOCPRO table</td>
<td>0.00</td>
<td>0.26%</td>
</tr>
</tbody>
</table>

Figure 4.3 shows the SAS log that is generated by PROC HPTMINE; the log provides information about the default configurations used by the procedure, about where the procedure runs, and about the input and output files. The log shows that the terms data set contains 271 observations. This means that the HPTMINE procedure identified 271 individual terms in the input document collection. Because K=10 in the SVD statement, the docpro data set contains 11 variables: the first variable is the document ID, and the remaining 10 variables are obtained by projecting the original document to the 10 left singular vectors that are computed by singular value decomposition.
Chapter 4: The HPTMINE Procedure

Figure 4.3  SAS Log

NOTE: Stemming will be used in parsing.
NOTE: Tagging will be used in parsing.
NOTE: Noun groups will be used in parsing.
NOTE: No TERMWGT option is specified. TERMWGT=ENTROPY will be run by default.
NOTE: No CELLWGT option is specified. CELLWGT=LOG will be run by default.
NOTE: No ENTITIES option is specified. ENTITIES=STD will be run by default.
NOTE: The HPTMINE procedure is executing on the client.
NOTE: There were 36 observations read from the data set WORK.GETSTART.
NOTE: The data set WORK.TERMS has 271 observations and 11 variables.
NOTE: The data set WORK.DOCPRO has 36 observations and 11 variables.

The following statements use PROC PRINT to show the contents of the first 10 rows of the docpro data set that is generated by the HPTMINE procedure:

```sas
proc print data = docpro (obs=10) round;
run;
```

Figure 4.4 shows the output of PROC PRINT. For information about the output of the OUTDOCPRO= option, see the section “The OUTDOCPRO= Data Set” on page 58.

**Figure 4.4  The docpro Data Set**

<table>
<thead>
<tr>
<th>Obs</th>
<th>did</th>
<th>COL1</th>
<th>COL2</th>
<th>COL3</th>
<th>COL4</th>
<th>COL5</th>
<th>COL6</th>
<th>COL7</th>
<th>COL8</th>
<th>COL9</th>
<th>COL10</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>d01</td>
<td>0.26</td>
<td>-0.76</td>
<td>0.21</td>
<td>-0.22</td>
<td>0.33</td>
<td>-0.18</td>
<td>0.25</td>
<td>0.08</td>
<td>-0.01</td>
<td>-0.24</td>
</tr>
<tr>
<td>2</td>
<td>d02</td>
<td>0.74</td>
<td>0.24</td>
<td>0.39</td>
<td>-0.04</td>
<td>-0.15</td>
<td>-0.02</td>
<td>-0.09</td>
<td>-0.09</td>
<td>-0.45</td>
<td>-0.01</td>
</tr>
<tr>
<td>3</td>
<td>d03</td>
<td>0.22</td>
<td>-0.23</td>
<td>-0.62</td>
<td>-0.23</td>
<td>-0.17</td>
<td>0.07</td>
<td>-0.48</td>
<td>-0.04</td>
<td>0.03</td>
<td>-0.44</td>
</tr>
<tr>
<td>4</td>
<td>d04</td>
<td>0.74</td>
<td>-0.01</td>
<td>0.13</td>
<td>0.35</td>
<td>0.08</td>
<td>-0.13</td>
<td>-0.41</td>
<td>0.19</td>
<td>0.18</td>
<td>0.23</td>
</tr>
<tr>
<td>5</td>
<td>d05</td>
<td>0.30</td>
<td>-0.23</td>
<td>0.49</td>
<td>-0.54</td>
<td>-0.30</td>
<td>-0.18</td>
<td>-0.19</td>
<td>0.38</td>
<td>0.06</td>
<td>0.14</td>
</tr>
<tr>
<td>6</td>
<td>d06</td>
<td>0.11</td>
<td>-0.24</td>
<td>-0.30</td>
<td>-0.23</td>
<td>-0.62</td>
<td>0.46</td>
<td>-0.08</td>
<td>-0.10</td>
<td>0.42</td>
<td>-0.11</td>
</tr>
<tr>
<td>7</td>
<td>d07</td>
<td>0.42</td>
<td>-0.33</td>
<td>0.06</td>
<td>0.40</td>
<td>-0.23</td>
<td>0.07</td>
<td>-0.18</td>
<td>-0.06</td>
<td>-0.55</td>
<td>0.39</td>
</tr>
<tr>
<td>8</td>
<td>d08</td>
<td>0.09</td>
<td>-0.16</td>
<td>-0.12</td>
<td>-0.40</td>
<td>0.26</td>
<td>-0.14</td>
<td>-0.24</td>
<td>-0.71</td>
<td>0.06</td>
<td>0.37</td>
</tr>
<tr>
<td>9</td>
<td>d09</td>
<td>0.85</td>
<td>0.10</td>
<td>-0.03</td>
<td>0.36</td>
<td>0.16</td>
<td>-0.23</td>
<td>-0.20</td>
<td>0.05</td>
<td>-0.06</td>
<td>-0.09</td>
</tr>
<tr>
<td>10</td>
<td>d10</td>
<td>0.23</td>
<td>-0.08</td>
<td>-0.60</td>
<td>-0.28</td>
<td>0.22</td>
<td>0.08</td>
<td>0.15</td>
<td>0.53</td>
<td>-0.16</td>
<td>0.34</td>
</tr>
</tbody>
</table>

The following statements use PROC SORT and PROC PRINT to show the contents of the terms data set that is generated by the HPTMINE procedure. In the output, only the terms that have a key value from 1 to 35 are shown. And for these terms only the values of the variables term, role, freq, numdocs, key, and parent are displayed.

```sas
proc sort data = terms; by key; run;
proc print data = terms (obs=35);
var term role freq numdocs key parent;
run;
```
Figure 4.5 shows the output of PROC PRINT, which provides details about the terms that are identified by the HPTMINE procedure. For example, it shows that the key of “problems” is 32 and its parent’s key is 9, which is the term “problem.” The HPTMINE procedure also identified that “sas” is a proper noun and that “big data” and “high-performance analytics” are noun groups. For information about the output of the OUTTERMS= option, see the section “The OUTTERMS= Data Set” on page 59.

Figure 4.5  The terms Data Set

<table>
<thead>
<tr>
<th>Obs</th>
<th>Term</th>
<th>Role</th>
<th>Freq</th>
<th>numdocs</th>
<th>Key</th>
<th>Parent</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>high-performance</td>
<td>Adj</td>
<td>3</td>
<td>3</td>
<td>1</td>
<td>.</td>
</tr>
<tr>
<td>2</td>
<td>big data</td>
<td>NOUN_GROUP</td>
<td>5</td>
<td>5</td>
<td>2</td>
<td>.</td>
</tr>
<tr>
<td>3</td>
<td>to</td>
<td>Prep</td>
<td>13</td>
<td>11</td>
<td>3</td>
<td>.</td>
</tr>
<tr>
<td>4</td>
<td>fast</td>
<td>Adj</td>
<td>4</td>
<td>4</td>
<td>4</td>
<td>.</td>
</tr>
<tr>
<td>5</td>
<td>fast</td>
<td>Adj</td>
<td>1</td>
<td>1</td>
<td>4</td>
<td>.</td>
</tr>
<tr>
<td>6</td>
<td>data</td>
<td>Noun</td>
<td>8</td>
<td>8</td>
<td>5</td>
<td>.</td>
</tr>
<tr>
<td>7</td>
<td>gain</td>
<td>Verb</td>
<td>2</td>
<td>2</td>
<td>6</td>
<td>.</td>
</tr>
<tr>
<td>8</td>
<td>big</td>
<td>Adj</td>
<td>6</td>
<td>6</td>
<td>7</td>
<td>.</td>
</tr>
<tr>
<td>9</td>
<td>for</td>
<td>Prep</td>
<td>6</td>
<td>6</td>
<td>8</td>
<td>.</td>
</tr>
<tr>
<td>10</td>
<td>problem</td>
<td>Noun</td>
<td>2</td>
<td>2</td>
<td>9</td>
<td>9</td>
</tr>
<tr>
<td>11</td>
<td>as</td>
<td>Prep</td>
<td>2</td>
<td>2</td>
<td>10</td>
<td>.</td>
</tr>
<tr>
<td>12</td>
<td>business</td>
<td>Noun</td>
<td>3</td>
<td>3</td>
<td>11</td>
<td>.</td>
</tr>
<tr>
<td>13</td>
<td>analytics</td>
<td>Prop</td>
<td>2</td>
<td>2</td>
<td>12</td>
<td>.</td>
</tr>
<tr>
<td>14</td>
<td>the</td>
<td>Det</td>
<td>13</td>
<td>13</td>
<td>13</td>
<td>.</td>
</tr>
<tr>
<td>15</td>
<td>time</td>
<td>Noun</td>
<td>2</td>
<td>2</td>
<td>14</td>
<td>.</td>
</tr>
<tr>
<td>16</td>
<td>in</td>
<td>Prep</td>
<td>3</td>
<td>3</td>
<td>15</td>
<td>.</td>
</tr>
<tr>
<td>17</td>
<td>process</td>
<td>Verb</td>
<td>2</td>
<td>2</td>
<td>16</td>
<td>16</td>
</tr>
<tr>
<td>18</td>
<td>and</td>
<td>Conj</td>
<td>13</td>
<td>13</td>
<td>17</td>
<td>.</td>
</tr>
<tr>
<td>19</td>
<td>insight</td>
<td>Noun</td>
<td>3</td>
<td>3</td>
<td>18</td>
<td>18</td>
</tr>
<tr>
<td>20</td>
<td>a</td>
<td>Det</td>
<td>3</td>
<td>3</td>
<td>19</td>
<td>.</td>
</tr>
<tr>
<td>21</td>
<td>compute</td>
<td>Verb</td>
<td>2</td>
<td>2</td>
<td>20</td>
<td>20</td>
</tr>
<tr>
<td>22</td>
<td>be</td>
<td>Verb</td>
<td>3</td>
<td>3</td>
<td>21</td>
<td>21</td>
</tr>
<tr>
<td>23</td>
<td>high-performance analytics</td>
<td>NOUN_GROUP</td>
<td>2</td>
<td>2</td>
<td>22</td>
<td>.</td>
</tr>
<tr>
<td>24</td>
<td>be</td>
<td>Aux</td>
<td>3</td>
<td>3</td>
<td>23</td>
<td>23</td>
</tr>
<tr>
<td>25</td>
<td>from</td>
<td>Prep</td>
<td>6</td>
<td>6</td>
<td>24</td>
<td>.</td>
</tr>
<tr>
<td>26</td>
<td>how</td>
<td>Conj</td>
<td>2</td>
<td>2</td>
<td>25</td>
<td>.</td>
</tr>
<tr>
<td>27</td>
<td>growth</td>
<td>Noun</td>
<td>2</td>
<td>2</td>
<td>26</td>
<td>.</td>
</tr>
<tr>
<td>28</td>
<td>it</td>
<td>Pron</td>
<td>2</td>
<td>2</td>
<td>27</td>
<td>.</td>
</tr>
<tr>
<td>29</td>
<td>enable</td>
<td>Verb</td>
<td>2</td>
<td>2</td>
<td>28</td>
<td>28</td>
</tr>
<tr>
<td>30</td>
<td>of</td>
<td>Prep</td>
<td>5</td>
<td>5</td>
<td>29</td>
<td>.</td>
</tr>
<tr>
<td>31</td>
<td>analytics</td>
<td>Noun</td>
<td>3</td>
<td>3</td>
<td>30</td>
<td>.</td>
</tr>
<tr>
<td>32</td>
<td>sas</td>
<td>Prop</td>
<td>7</td>
<td>7</td>
<td>31</td>
<td>.</td>
</tr>
<tr>
<td>33</td>
<td>problems</td>
<td>Noun</td>
<td>2</td>
<td>2</td>
<td>32</td>
<td>9</td>
</tr>
<tr>
<td>34</td>
<td>insights</td>
<td>Noun</td>
<td>3</td>
<td>3</td>
<td>33</td>
<td>18</td>
</tr>
<tr>
<td>35</td>
<td>enables</td>
<td>Verb</td>
<td>2</td>
<td>2</td>
<td>34</td>
<td>28</td>
</tr>
</tbody>
</table>
Syntax: HPTMINE Procedure

The following statements are available in the HPTMINE procedure:

```
PROC HPTMINE < options > ;
    VARIABLE variable ;
    TARGET variable ;
    DOC_ID variable ;
    PARSE < parse-options > ;
    SVD < svd-options > ;
    PERFORMANCE performance-options ;
```

PROC HPTMINE Statement

```
PROC HPTMINE < options > ;
```

The PROC HPTMINE statement invokes the procedure. Table 4.1 summarizes the `options` in the statement by function. The `options` are then described fully in alphabetical order.

<table>
<thead>
<tr>
<th>option</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Basic Options</strong></td>
<td></td>
</tr>
<tr>
<td>DATA</td>
<td>Specifies the input document data set</td>
</tr>
<tr>
<td>DOC=</td>
<td></td>
</tr>
<tr>
<td><strong>Output Options</strong></td>
<td>Suppresses ODS output</td>
</tr>
<tr>
<td>NOPRINT</td>
<td></td>
</tr>
</tbody>
</table>

You can specify the following `options`:

**DATA=**SAS-data-set

names the input SAS data set of documents to be used by PROC HPTMINE. The default is the most recently created data set. If PROC HPTMINE executes in distributed mode, the input data are distributed to memory on the appliance nodes and analyzed in parallel, unless the data are already distributed in the appliance database. When data are already distributed, PROC HPTMINE reads the data alongside the distributed database. For more information, see the sections “Processing Modes” on page 8 and “Alongside-the-Database Execution” on page 15 in Chapter 3, “Shared Concepts and Topics.” Each row of the input data set must contain one text field and one ID field that correspond to the text and the unique ID of a document, respectively.

**NOPRINT**

suppresses the generation of ODS output.
**DOC_ID Statement**

```
DOC_ID variable < docid-option > ;
```

This statement specifies the `variable` that contains the ID of each document. In the input data set, each row corresponds to one document. The ID of each document must be unique; it can be either a number or a string of characters.

You can specify the following `docid-option`:

- `DUPLICATIONCHECK < ( WARNING | STOP ) >`
- `DUPCHK < ( WARNING | STOP ) >`

`DUPLICATIONCHECK` checks whether the ID of each document is unique. When you do not specify this option, no duplication check is performed. When you specify `DUPLICATIONCHECK(WARNING)`, the HPTMINE procedure outputs a warning message that shows the number of duplicate document IDs that have been detected, unless the procedure finds no duplicate IDs, in which case no message is generated. When you specify `DUPLICATIONCHECK(STOP)`, the HPTMINE procedure outputs an error message that shows the number of duplicate document IDs that have been detected, and then the procedure terminates. If you specify the `DUPLICATIONCHECK` option without the `WARNING` or `STOP` keyword, the HPTMINE procedure takes the `WARNING` keyword by default.

**PARSE Statement**

```
PARSE < parse-options > ;
```

The `PARSE` statement specifies the options for parsing the input documents and creating the term-by-document matrix. Table 4.2 summarizes the `parse-options` in the statement by function. The `parse-options` are then described fully in alphabetical order.

<table>
<thead>
<tr>
<th>parse-option</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Parsing Options</strong></td>
<td></td>
</tr>
<tr>
<td>NOSTEMMING</td>
<td>Specifies no stemming in parsing</td>
</tr>
<tr>
<td>NOTAGGING</td>
<td>Specifies no part-of-speech tagging in parsing</td>
</tr>
<tr>
<td>NONOUNGROUPS X NONG</td>
<td>Specifies no noun group extraction in parsing</td>
</tr>
<tr>
<td>ENTITIES=</td>
<td>Specifies whether to extract entities in parsing</td>
</tr>
<tr>
<td>STOP=</td>
<td>Specifies the stop list</td>
</tr>
<tr>
<td>START=</td>
<td>Specifies the start list</td>
</tr>
<tr>
<td>MULTITERM=</td>
<td>Specifies the multiword term list</td>
</tr>
<tr>
<td>SYNONYM</td>
<td>SYN=</td>
</tr>
<tr>
<td><strong>Term-by-Document Matrix Creation Options</strong></td>
<td></td>
</tr>
<tr>
<td>TERMWGT=</td>
<td>Specifies how terms are weighted</td>
</tr>
<tr>
<td>CELLLWGT=</td>
<td>Specifies how cells are weighted</td>
</tr>
<tr>
<td>REDUCECF=</td>
<td>Specifies the frequency for term filtering</td>
</tr>
</tbody>
</table>

**Table 4.2** PARSE Statement Options
Chapter 4: The HPTMINE Procedure

### Table 4.2 continued

<table>
<thead>
<tr>
<th>parse-option</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Output Options</strong></td>
<td></td>
</tr>
<tr>
<td>OUTPARENT=</td>
<td>Specifies the data set to contain the term-by-document frequency matrix. Child terms are not represented in the data set. The frequencies of child terms are attributed to their corresponding parents.</td>
</tr>
<tr>
<td>OUTCHILD=</td>
<td>Specifies a secondary choice to contain the term-by-document frequency matrix. All kept terms, whether or not they are child terms, are represented in the data set with their corresponding frequency.</td>
</tr>
<tr>
<td>OUTTERMS=</td>
<td>Specifies the data set to contain the summary information about the terms in the document collection</td>
</tr>
<tr>
<td>OUTCONFIG=</td>
<td>Specifies the data set to contain the option settings that PROC HPTMINE uses in the current run</td>
</tr>
</tbody>
</table>

You can specify the following **options** in the PROC HPTMINE statement.

**CELLWGT=LOG | NONE**

specifies how the elements in the term-by-document matrix are weighted. The available cell weights are as follows:

- **LOG** specifies that cells be weighted by using the LOG formulation.
- **NONE** specifies that no cell weight be applied.

For information about the LOG formulation for cell weighting, see the section “Term and Cell Weighting” on page 55.

**ENTITIES=STD | NONE**

determines whether the entity extractor should use the standard list of entities. When ENTITIES=STD, entity extraction is turned on and standard entities are used; terms such as “George W. Bush” are recognized as an entity and given the corresponding entity role and attribute. For this example, the entity role is PERSON and the attribute is Entity. Although the entity is treated as the single term, “george w. bush,” the individual tokens “george,” “w,” and “bush” are also included. By default, ENTITIES=NONE.

**MULTITERM=OS-file-reference**

specifies the path to a file that contains a list of multiword terms. These terms are case-sensitive and are treated as a single entry by the HPTMINE procedure. Thus, the terms “Thank You” and “thank you” are processed differently. Consequently, you must convert all text strings to lowercase or add each of the multiterm’s case variations to the list before using the HPTMINE procedure to create consistent multiword terms. The multiterm file can be any file type as long as it is formatted in the following manner:

```
multiterm: 3: pos
```

Specifically, the first item is the multiword term itself followed by a colon, the second item is a number that is specific to Teragram and represents the token type followed by a colon, and the third item is the part of speech that the multiword term represents. **NOTE:** The token type 3 is the most common token type for multiterm lists and represents compound words.
NONOUNGROUPS | NONG determines whether the noun group extractor should be used. When this option is not specified, noun groups are extracted and the HPTMINE procedure returns maximal groups and subgroups (which do not include groups that contain determiners or prepositions). If stemming is turned on, then noun group elements are also stemmed.

NOSTEMMING determines whether words should be stemmed. When this option is not specified, words are stemmed and terms such as “advises” and “advising” are mapped to the parent term “advise.”

NOTAGGING determines whether terms should be tagged. When this option is not specified, terms are tagged and the HPTMINE procedure identifies a term’s part of speech based on context clues. The identified part of speech is provided in the Role variable of the OUTTERMS= data set.

OUTCHILD= SAS-data-set specifies the data set to contain a compressed representation of the sparse term-by-document frequency matrix. The term counts cannot be weighted. The data set saves only the kept, representative terms. The child frequencies are not attributed to their corresponding parent (as they are in the OUTPARENT= data set). For more information about the compressed representation of the sparse term-by-document frequency matrix, see the section “Output Data Sets” on page 57.

OUTCONFIG= SAS-data-set specifies the data set to contain configuration information that is used for the current run of PROC HPTMINE. The primary purpose of this data set is to relay the configuration information from the HPTMINE procedure to the HPTMSCORE procedure. The HPTMSCORE procedure is forced to use options that are consistent with the HPTMINE procedure. Thus, the data set that is created by using the OUTCONFIG= option becomes an input data set for PROC HPTMSCORE and ensures that the parsing options are consistent between the two runs. For more information about this data set, see the section “Output Data Sets” on page 57.

OUTPARENT= SAS-data-set specifies the data set to contain a compressed representation of the sparse term-by-document frequency matrix. The term counts can be weighted, if requested. The data set contains only the kept, representative terms, and the child frequencies are attributed to the corresponding parent. To obtain information about the children, use the OUTCHILD= option. For more information about the compressed representation of the sparse term-by-document frequency matrix, see the section “Output Data Sets” on page 57.

OUTTERMS= SAS-data-set specifies the data set to contain the summary information about the terms in the document collection. For more information about this data set, see the section “Output Data Sets” on page 57.

REDUCEF= n removes terms that are not in at least n documents. The value of n must be a positive integer. By default, REDUCEF=4.

START= SAS-data-set specifies the data set that contains the terms that are to be kept for the analysis. These terms are displayed in the OUTTERMS= data set with a keep status of Y, and all other terms have a keep status
Chapter 4: The HPTMINE Procedure

of N. The START data set must have a Term variable and can also have a Role variable. You cannot specify both the START and the STOP options.

STOP=SAS-data-set

specifies the data set that contains the terms to exclude from the analysis. These terms are displayed in the OUTTERMS= data set with a keep status of N. They are not identified as parents or children. The STOP data set must have a Term variable and can also have a Role variable. You cannot specify both the START and the STOP options.

SYNONYM=SAS-data-set

SYN=SAS-data-set

specifies the data set that contains user-defined synonyms to be used in the analysis. The data set specifies parent-child relationships that enable you to map child terms to a representative parent. The synonym relationship is indicated in the data set that is specified in the OUTTERMS= option and is also reflected in the term-by-document data set that is specified in the OUTPARENT= option. The input synonym data set must have either the two variables Term and Parent or the four variables Term, Parent, Termrole, and Parentrole. When stemming is turned on, this list overrides any relationships that are identified when terms are stemmed.

TERMWGT=ENTROPY | MI | NONE

specifies how terms are weighted. The available term weights are as follows:

- **ENTROPY** requests that terms be weighted using the entropy formulation.
- **MI** requests that terms be weighted using the mutual information formulation.
- **NONE** requests that no term weight be applied.

For more information about the entropy formulation and the mutual information formulation for term weighting, see the section “Term and Cell Weighting” on page 55.

PERFORMANCE Statement

**PERFORMANCE < performance-options> ;**

The PERFORMANCE statement defines performance parameters for multithreaded and distributed computing, passes variables that describe the distributed computing environment, and requests detailed results about the performance characteristics of the HPTMINE procedure. You can also use the PERFORMANCE statement to control whether the HPTMINE procedure executes in single-machine or distributed mode. The PERFORMANCE statement is documented further in the section “PERFORMANCE Statement” on page 35 of Chapter 3, “Shared Concepts and Topics.”

SVD Statement

**SVD < svd-options> ;**

The SVD statement specifies the options for calculating a truncated singular value decomposition (SVD) of the large, sparse term-by-document frequency matrix that was created during the parsing phase of PROC
HPTMINE. Table 4.3 summarizes the *svd-options* in the statement by function. The *svd-options* are then described fully in alphabetical order.

**Table 4.3** SVD Statement Options

<table>
<thead>
<tr>
<th>svd-option</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>SVD Computation Options</strong></td>
<td></td>
</tr>
<tr>
<td>K=</td>
<td>Specifies the number of dimensions to be extracted</td>
</tr>
<tr>
<td>MAX_K=</td>
<td>Specifies the maximum number of dimensions to be extracted</td>
</tr>
<tr>
<td>TOL=</td>
<td>Specifies the maximum allowable tolerance for the singular value</td>
</tr>
<tr>
<td>RESOLUTION = RES=</td>
<td>Specifies the recommended number of dimensions (resolution) extracted by SVD, when the MAX_K= option is used</td>
</tr>
<tr>
<td><strong>Output Options</strong></td>
<td></td>
</tr>
<tr>
<td>SVDU=</td>
<td>Specifies the U matrix, which contains the left singular vectors</td>
</tr>
<tr>
<td>SVDV=</td>
<td>Specifies the V matrix, which contains the right singular vectors</td>
</tr>
<tr>
<td>SVDS=</td>
<td>Specifies the S matrix, whose diagonal elements are the singular values</td>
</tr>
<tr>
<td>OUTDOCPRO=</td>
<td>Specifies the projections of the documents</td>
</tr>
</tbody>
</table>

You can specify the following *svd-options*:

**K=**

specifies the number of columns in the matrices U, V, and S. This value is the number of dimensions of the data set after the SVD is performed. If the value of \( k \) is too large, then the HPTMINE procedure runs for an unnecessarily long time. You cannot specify both this option and the MAX_K= option.

**MAX_K=**

specifies the maximum value that the HPTMINE procedure should return as the recommended value of \( n \). If the RES= argument is specified to recommend the value of \( k \), this option limits that value to at most \( n \). The HPTMINE procedure attempts to calculate (as opposed to recommends) \( k \) dimensions when it performs SVD. You cannot specify both this option and the K= option.

**OUTDOCPRO=SAS-data-set <KEEPVARIABLES=variable-list><NONORMDOC>**

specifies the data set to contain the projections of the columns of the term-by-document frequency matrix onto the columns of U. Because each column of the term-by-document frequency matrix corresponds to a document, the output forms a new representation of the input documents in a space that has much lower dimensionality.

You can copy the variables from the data set that is specified in the DATA= option in the PROC HPTMINE statement to the *SAS-data-set* that is specified in this option. When you specify KEEP-VARIABLES=variable-list, the content of the variables that is specified in the variable-list is attached to the output. These variables must appear in the data set that is specified in the DATA= option in the PROC HPTMINE statement.

When you specify NONORMDOC, the columns that contain the projections of documents are not normalized to have a unit norm.
RESOLUTION=LOW | MED | HIGH
RES=LOW | MED | HIGH

specifies the recommended number of dimensions (resolution) for the singular value decomposition. If you specify this option, you must also specify the MAX_K= option. A low-resolution singular value decomposition returns fewer dimensions than a high-resolution singular value decomposition. This option recommends the value of $k$ heuristically based on the value specified in the MAX_K= option. Assume that the MAX_K= option is set to $n$ and a singular value decomposition with $n$ dimensions accounts for $t\%$ of the total variance. The option HIGH always recommends the maximum number of dimensions; that is, $k = n$. The option MED recommends a $k$ that explains $(5/6) * t\%$ of the total variance. The option LOW recommends a $k$ that explains $(2/3) * t\%$ of the total variance.

SVDS=SAS-data-set

specifies the data set to contain the calculated singular values.

SVDU=SAS-data-set

specifies the data set to contain the calculated left singular vectors.

SVDV=SAS-data-set

specifies the data set to contain the calculated right singular vectors.

TOL=$\epsilon$

specifies the maximum allowable tolerance for the singular value. Let $A$ be a matrix. Suppose $\lambda_i$ is the $i$th singular value of $A$ and $\xi_i$ is the corresponding right singular vector. The SVD computation terminates when for all $i \in \{1, \ldots, k\}$, $\lambda_i$ and $\xi_i$ satisfy $\|A^{T}A\xi - \lambda \xi\|^2 \leq \epsilon$. The default value of $\epsilon$ is $10^{-6}$, which is more than adequate for most text mining problems.

TARGET Statement

TARGET variable ;

This statement specifies the variable that contains the information about the category that a document belongs to. A target variable can be any nominal or ordinal variable. It is used in calculating supervised term weighting, such as the mutual information formulation.

VARIABLE Statement

VARIABLE variable ;

VAR variable ;

This statement specifies the variable that contains the text to be processed.

Details: HPTMINE Procedure

PROC HPTMINE integrates the functionalities from both natural language processing and statistical analysis to provide essential support for high-performance text mining. This section provides details about various
aspects of the HPTMINE procedure.

Natural Language Processing

Natural language processing (NLP) techniques can be used to extracting meaningful information from natural language input. The following sections describe features from SAS linguistic technologies that are supported in the HPTMINE procedure to support nature language processing.

Stemming

Stemming (a special case of morphological analysis) identifies the possible root form of an inflected word. For example, the word “talk” is the stem of the words “talk,” “talks,” “talking,” and “talked.” In this case “talk” is the parent, and “talk,” “talks,” “talking,” and “talked” are its children.

Part-of-Speech Tagging

Using SAS linguistic technologies, tagging identifies, or disambiguates, the grammatical category of a word by analyzing it within its context. For example:

I like to bank at the local branch of my bank.

In this case the first “bank” is tagged as a verb (V), and the second “bank” is tagged as a noun (N). Table 4.4 shows all possible part-of-speech tags.

<table>
<thead>
<tr>
<th>Part-of-Speech Tag</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>ABBR</td>
<td>Abbreviation</td>
</tr>
<tr>
<td>ADJ</td>
<td>Adjective</td>
</tr>
<tr>
<td>ADV</td>
<td>Adverb</td>
</tr>
<tr>
<td>AUX</td>
<td>Auxiliary or modal term</td>
</tr>
<tr>
<td>CONJ</td>
<td>Conjunction</td>
</tr>
<tr>
<td>DET</td>
<td>Determiner</td>
</tr>
<tr>
<td>INTERJ</td>
<td>Interjection</td>
</tr>
<tr>
<td>NOUN</td>
<td>Noun</td>
</tr>
<tr>
<td>NOUN_GROUP</td>
<td>Compound noun</td>
</tr>
<tr>
<td>NUM</td>
<td>Number or numeric expression</td>
</tr>
<tr>
<td>PART</td>
<td>Infinitive marker, negative participle, or possessive marker</td>
</tr>
<tr>
<td>PREF</td>
<td>Prefix</td>
</tr>
<tr>
<td>PREP</td>
<td>Preposition</td>
</tr>
<tr>
<td>PRON</td>
<td>Pronoun</td>
</tr>
<tr>
<td>PROP</td>
<td>Proper noun</td>
</tr>
<tr>
<td>PUNCT</td>
<td>Punctuation</td>
</tr>
<tr>
<td>VERB</td>
<td>Verb</td>
</tr>
<tr>
<td>VERBADJ</td>
<td>Verbal adjective</td>
</tr>
</tbody>
</table>
Noun Group Extraction

Noun groups provide more relevant information than simple nouns. A noun group is defined as a sequence of nouns and their modifiers. Noun group extraction uses part-of-speech tagging to identify nouns and their related words that together form a noun group. Examples of noun groups are “weeklong cruises” and “Middle Eastern languages.”

Entity Identification

Entity identification uses SAS linguistic technologies to classify sequences of words into predefined classes. These classes are assigned as roles for the corresponding sequences. For example, “Person,” “Location,” “Company,” and “Measurement” are identified as classes for “George W. Bush,” “Boston,” “SAS Institute,” “2.5 inches,” respectively. Table 4.5 shows all valid entities.

<table>
<thead>
<tr>
<th>Entities</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>ADDRESS</td>
<td>Postal address or number and street name</td>
</tr>
<tr>
<td>COMPANY</td>
<td>Company name</td>
</tr>
<tr>
<td>CURRENCY</td>
<td>Currency or currency expression</td>
</tr>
<tr>
<td>INTERNET</td>
<td>E-mail address or URL</td>
</tr>
<tr>
<td>LOCATION</td>
<td>City, county, state, political or geographical place or region</td>
</tr>
<tr>
<td>MEASURE</td>
<td>Measurement or measurement expression</td>
</tr>
<tr>
<td>NOUN_GROUP</td>
<td>Phrases that contain multiple words</td>
</tr>
<tr>
<td>ORGANIZATION</td>
<td>Government, legal, or service agency</td>
</tr>
<tr>
<td>PERCENT</td>
<td>Percentage or percentage expression</td>
</tr>
<tr>
<td>PERSON</td>
<td>Person’s name</td>
</tr>
<tr>
<td>PHONE</td>
<td>Telephone number</td>
</tr>
<tr>
<td>PROP_MISC</td>
<td>Proper noun with an ambiguous classification</td>
</tr>
<tr>
<td>SSN</td>
<td>Social Security number</td>
</tr>
<tr>
<td>TIME</td>
<td>Time or time expression</td>
</tr>
<tr>
<td>TIME_PERIOD</td>
<td>Measure of time expressions</td>
</tr>
<tr>
<td>TITLE</td>
<td>Person’s title or position</td>
</tr>
<tr>
<td>VEHICLE</td>
<td>Motor vehicle, including color, year, make, and model</td>
</tr>
</tbody>
</table>

Multiword Terms Handling

By default, SAS linguistic technologies tokenize the text to individual words and operate at the word level. Multiword terms provide a control that enables you to specify sequences of words to be interpreted as individual units. For example, “greater than,” “in spite of,” and “as well as” can be defined as multiword terms.

Language Support

In the current release, English is the only supported language. By turning off some of the advanced parsing functionality, you might be able to use PROC HPTMINE effectively with other space-delimited languages.
Term and Cell Weighting

The TERMWGT= option and the CELLWGT= option control how to weight the frequencies in the compressed term-by-document matrix. The term weight is a positive number that is assigned to each term based on the distribution of that term in the document collection. This weight can be interpreted as an indication of the importance of that term to the document collection. The cell weight is a function that is applied to every entry in the term-by-document matrix; it moderates the effect of a term that is repeated within a document.

Let $f_{i,j}$ be the entry in the $i$th row and $j$th column of the term-by-document matrix, which indicates the time of appearance of term $i$ in document $j$. Assuming that the term weight of term $i$ is $w_i$ and the cell weight function is $g(x)$, the weighted frequency of each entry in the term-by-document matrix is given by $w_i \times g(f_{i,j})$.

When the CELLWGT=LOG option is specified, the following equation is used to weight cells:

$$g(x) = \log_2(f_{i,j} + 1)$$

The equation reduces the influence of highly frequent terms by applying the log function.

When the TERMWGT=ENTROPY option is specified, the following equation is used to weight terms:

$$w_i = 1 + \sum_j p_{i,j} \log_2(p_{i,j}) / \log_2(n)$$

In this equation, $n$ is the number of documents, and $p_{i,j}$ is the probability that term $i$ appears in document $j$, which can be estimated by $p_{i,j} = f_{i,j} / g_i$, where $g_i$ is the global term frequency for term $i$.

When the TERMWGT=MI option is specified, the following equation is used to weight terms:

$$w_i = \max_{C_k} \left( \log \left( \frac{P(t_i, C_k)}{P(t_i) P(C_k)} \right) \right)$$

In this equation, $C_k$ is the set of documents that belong to category $k$, $P(C_k)$ is the percentage of documents that belong to category $k$, and $P(t_i, C_k)$ is the percentage of documents that contain term $t_i$ and belong to category $k$. Let $d_i$ be the number of documents that term $i$ appears in. Then $P(t_i) = d_i / n$.

Singular Value Decomposition

Singular value decomposition (SVD) of a matrix $A$ factors $A$ into three matrices such that $A = U \Sigma V^T$. The singular value decomposition also requires that the columns of $U$ and $V$ be orthogonal and that $\Sigma$ be a real-valued diagonal matrix with monotonically decreasing, nonnegative entries. The entries of $\Sigma$ are called singular values. The columns of $U$ and $V$ are called left and right singular vectors, respectively. A truncated singular value decomposition calculates only the first $k$ singular values and their corresponding left and right singular vectors. In information retrieval, singular value decomposition is also known as latent semantic indexing (LSI).
Applications in Text Mining

Let $A \in \mathbb{R}^{m \times n}$ be a term-by-document matrix, where $m$ is the number of terms and $n$ is the number of documents. The SVD statement has two main functions: to calculate a truncated singular value decomposition (SVD) of $A$, and to project the columns of $A$ onto the left singular vectors to generate a new representation of the documents with much lower dimensionality. The output of the SVD statement is a truncated singular value decomposition of $A$, for which you specify the parameter $k$ to define how many singular values and singular vectors to compute. A singular value decomposition reduces the dimension of the term-by-document matrix and reveals themes that are present in the document collection.

In general, the value of $k$ must be large enough to capture the meaning of the document collection, yet small enough to ignore the noise. You can specify this value explicitly or accept a value that is recommended by the HPTMINE procedure. A value between 50 and 200 should work well for a document collection that contains thousands of documents.

An important purpose of singular value decomposition is to reduce a high-dimensional term-by-document matrix into a low-dimensional representation that reveals information about the document collection. The columns of the $A$ form the coordinates of the document space, and the rows form the coordinates of the term space. Each document in the collection is represented as a vector in $m$-dimensional space and each term as a vector in $n$-dimensional space. The singular value decomposition captures this same information by using a smaller number of basis vectors than would be necessary if you analyzed $A$ directly.

For example, consider the columns of $A$, which represent the document space. By construction, the columns of $U$ also reside in $m$-dimensional space. If $U$ has only one column, the line between that vector and the origin would form the best fit line, in a least squares sense, to the original document space. If $U$ has two columns, then these columns would form the best fit plane to the original document space. In general, the first $k$ columns of $U$ form the best fit $k$-dimensional subspace for the document space. Thus, you can project the columns of $A$ onto the first $k$ columns of $U$ in order to optimally reduce the dimension of the document space from $m$ to $k$.

The projection of a document $d$ (one column of $A$) onto $U$ results in $k$ real numbers that are defined by the inner product $d$ with each column of $U$. That is, $p_i = d^T u_i$. With this representation, each document forms a $k$-dimensional vector that can be considered a theme in the document collection. You can then calculate the Euclidean distance between each document and each column of $U$ to determine the documents that are described by this theme.

In a similar fashion, you can repeat the previous process by using the rows of $A$ and the first $k$ columns of $V$. This generates a best fit $k$-dimensional subspace for the term space. This representation is used to group terms into similar clusters. These clusters also represent concepts that are prevalent in the document collection. Thus, singular value decomposition can be used to cluster both the terms and documents into meaningful representations of the entire document collection.

Computation

The computation of the singular vector decomposition is fully parallelized in PROC HPTMINE to support both single-machine mode via multithreading and distributed model via distributed computing. In the current release, computing singular value decomposition requires the input data to contain at least 25 documents and at least as many documents as there are nodes in the grid. If $p$ nodes are used for computing singular value decomposition in a distributed computing environment, then the input data must contain at least $\max(p, 25)$ documents. Computing singular value decomposition is an iterative process that involves considerable communication among the computer nodes in a distributed computing environment. Therefore, adding
more computer nodes for computing singular value decomposition might not always improve efficiency. Conversely, when the data size is not large enough, adding too many computer nodes for computation might lead to a noticeable increase in communication time and sometimes might even slow down the overall computation.

Output Data Sets

This section describes the output data sets that PROC HPTMINE produces when you specify the corresponding option.

The OUTCONFIG= Data Set

The OUTCONFIG= option specifies a SAS data set to contain the configuration that PROC HPTMINE uses in the current run. The primary purpose of this data set is to relay the configuration information from the HPTMINE procedure to the HPTMSCORE procedure. Thus, the HPTMSCORE procedure can use the options that are consistent with the HPTMINE procedure during scoring.

Table 4.6 shows the configuration information that is contained in this data set.

<table>
<thead>
<tr>
<th>Field</th>
<th>Indicates</th>
</tr>
</thead>
<tbody>
<tr>
<td>namedfile</td>
<td>Whether the parsed variable is in an external file or in the input data set (this field is not applicable in the current release)</td>
</tr>
<tr>
<td>language</td>
<td>The source language of the documents (English is the only supported language)</td>
</tr>
<tr>
<td>encoding</td>
<td>The encoding of the documents (&quot;wlatin1&quot; is the only supported encoding)</td>
</tr>
<tr>
<td>plugin</td>
<td>The plug-in that is used to parse the document collection (&quot;Tera-seg&quot; is the only supported plug-in)</td>
</tr>
<tr>
<td>stemming</td>
<td>Whether stemming is used: &quot;Y&quot; indicates that stemming is used, and &quot;N&quot; indicates that it is not used</td>
</tr>
<tr>
<td>tagging</td>
<td>Whether tagging is used: &quot;Y&quot; indicates that tagging is used, and &quot;N&quot; indicates that it is not used</td>
</tr>
<tr>
<td>NG</td>
<td>Whether noun group is used: &quot;Y&quot; indicates that noun group is used, and &quot;N&quot; indicates that it is not used</td>
</tr>
<tr>
<td>entities</td>
<td>Whether entities should be extracted: &quot;STD&quot; indicates that entities should be extracted, and &quot;N&quot; indicates that entities should not be extracted</td>
</tr>
<tr>
<td>entList</td>
<td>The entities that are kept (this field is not applicable in the current release)</td>
</tr>
<tr>
<td>attributes</td>
<td>Whether attributes are kept or dropped: &quot;Y&quot; indicates that attributes are kept, and &quot;N&quot; indicates that attributes are dropped</td>
</tr>
<tr>
<td>attrList</td>
<td>The attributes that are kept or dropped (this field is not applicable in the current release)</td>
</tr>
<tr>
<td>pos</td>
<td>Whether parts of speech are kept or dropped (this field is not applicable in the current release)</td>
</tr>
</tbody>
</table>
Table 4.6  continued

<table>
<thead>
<tr>
<th>Field</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>posList</td>
<td>The parts of speech that are kept or dropped (this field is not applicable in the current release)</td>
</tr>
<tr>
<td>multiterm</td>
<td>The path to the multiterm list file</td>
</tr>
<tr>
<td>litiList</td>
<td>The path to your custom entities (this field is not applicable in the current release)</td>
</tr>
<tr>
<td>cellwgt</td>
<td>How the cells of the term-by-document frequency matrix are weighted</td>
</tr>
<tr>
<td>nonormdoc</td>
<td>Whether document projects are normalized: “Y” indicates document projects are normalized to have unit norm.</td>
</tr>
</tbody>
</table>

The content of this data set is case-sensitive.

The OUTDOCPRO= Data Set

The OUTDOCPRO= option specifies a SAS data set to contain the projections of the columns of the term-by-document frequency matrix onto the columns of U. Because each column of the term-by-document frequency matrix corresponds to a document, the output forms a new representation of the input documents in a space with much lower dimensionality. If the K= option in the SVD statement is set to $k$ and the input data set contains $n$ documents, the output will have $n$ rows and $k + 1$ columns. Each row of the output corresponds to a document. The first column of the output contains the ID of the documents, and the name of the column is the same as the variable that is specified in the DOC_ID statement. The remaining $k$ columns are the projections and are named “COL1” to “COL$k$.”

The OUTPARENT= Data Set

The OUTPARENT= option specifies a SAS data set to contain a compressed representation of the sparse term-by-document frequency matrix. The term-by-document frequency matrix is usually very sparse.\(^1\) To save space, the transactional style is used as a compressed representation to store the sparse matrix. In this format, the matrix is represented as a set of triples $(i, j, x)$, where $x$ is an entry in the matrix and $i$ and $j$ denote its row and column indices, respectively. The transactional style can also be used when the output contains too many columns to fit in a data set. For example, in most database systems the maximum number of columns in a table is usually limited to several thousand. If an output matrix contains more columns than the limit, you must use the transactional style to avoid errors that would arise from writing too many columns to the table. When the transactional style is used, all 0 entries in the matrix are ignored in the output, thereby saving storing space when the matrix is sparse.

The data set of the OUTPARENT= option contains three columns: _TERMNUM_, _DOCUMENT_, and _COUNT_. The _TERMNUM_ column contains the ID of the terms, which corresponds to the “Key” column of the data set that is generated by the OUTTERMS= option. The _DOCUMENT_ column contains the ID of the documents, and the _COUNT_ column contains the term counts. For example, $(t1 \ d1 \ k)$ means that term $t1$ appears $k$ times in document $d1$.

The term counts can be weighted, if requested. The data set saves only the terms that are marked as kept in the data set that is specified in the OUTTERMS= option of the PARSE statement. In the data set set the child

---

\(^1\)Many elements of the matrix are 0.
frequencies are attributed to the corresponding parent. For example, if “said” has term ID t1 and appears eight times in the document d1, “say” has term ID t2 and appears one time in the document d1. Assume that “say” is the parent of “said” and that neither cell weighting nor term weighting is applied. Then the data set that is specified in the OUTPARENT= option will have an entry

\[ t2 \quad d1 \quad 9 \]

The term count of “said” in d1 is attributed to its parent, “say.”

The OUTCHILD= Data Set

The OUTCHILD= option specifies the data set to contain a compressed representation of the sparse term-by-document frequency matrix, which is usually very sparse. To save storage space, the transactional style is used as a compressed representation to store the sparse matrix. Like the data set that is specified in the OUTPARENT= option, the data set specified in the OUTCHILD= option also contains three columns: _TERMNUM_, _DOCUMENT_, and _COUNT_. The _TERMNUM_ column contains the ID of the terms, which corresponds to the “Key” column of the data set generated by the OUTTERMS= option. The _DOCUMENT_ column contains the ID of the documents, and the _COUNT_ column contains the term counts. For example, (t1 d1 k) means that term t1 appears k times in document d1.

The data set saves only the terms that are marked as kept in the data set specified in the OUTTERMS= option of the PARSE statement. The child frequencies are not attributed to their corresponding parent (as they are in the data set specified in the OUTPARENT= option). Using the example in the previous section, the data set that is generated by the OUTCHILD= option will have two entries:

\[ t1 \quad d1 \quad 8 \]
\[ t2 \quad d1 \quad 1 \]

The term count of “said” in d1 is not attributed to its parent, “say.” The data set that is specified in the OUTCHILD= option can be combined with the data set that is specified in the OUTTERMS= option to construct the data set that is specified in the OUTPARENT= option.

The OUTTERMS= Data Set

The data set specified in the OUTTERMS= option contains the summary information about the terms in the document collection. Table 4.7 shows the fields in this data set.

<table>
<thead>
<tr>
<th>Fields</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>Term</td>
<td>A lowercase version of the term</td>
</tr>
<tr>
<td>Role</td>
<td>The term’s part of speech (this variable is empty if the NOTAGGING option is specified in the PARSE statement)</td>
</tr>
</tbody>
</table>
### Table 4.7  continued

<table>
<thead>
<tr>
<th>Fields</th>
<th>Description</th>
</tr>
</thead>
</table>
| Attribute | An indication of the characters that compose the term. Possible attributes are as follows:  
  - **Alpha**: only alphabetic characters  
  - **Mixed**: a combination of attributes  
  - **Num**: only numbers  
  - **Punct**: punctuation characters  
  - **Entity**: an identified entity |
| Freq     | The frequency of a term in the entire document collection |
| numdocs  | The number of documents that contain the term |
| _keep    | The keep status of the term: “Y” indicates that the term is kept for analysis, and “N” indicates that the term should be dropped in later stages of analysis. To ensure that the OUTTERMS= data set is of a reasonable size, only terms that have _keep=Y are kept in the OUTTERMS= data set. |
| Key      | The assigned term number (each unique term in the parsed documents and each unique parent term has a unique Key value) |
| Parent   | The Key value of the term’s parent or a “.” (period):  
  - If a term has a parent, this variable gives the term number of that parent.  
  - If a term does not have a parent, this value is a “.” (period).  
  - If the values of Key, Parent, and Parent_id are identical, the parent occurs as itself.  
  - If the values of Parent and Parent_id are identical but differ from Key, the observation is a child. |
| Parent_id| Another description of the term’s parent: Parent gives the parent’s term number if a term is a child, but Parent_id gives this value for all terms. |
| _ispar   | An indication of term’s status as a parent, child, or neither:  
  - A “+” (plus sign) indicates that the term is a parent.  
  - A “.” (period) indicates that the term is a child.  
  - A missing value indicates that the term is neither a parent nor a child. |
| Weight   | The weights of the terms |
Displayed Output

The following sections describe the output that PROC HPTMINE produces. The output is organized into various tables, which are discussed in their order of appearance.

Performance Information

The “Performance Information” table is produced by default; it displays information about the grid host for distributed execution and information about whether the procedure executes in single-machine mode, distributed mode, or alongside-the-database mode. The numbers of computing nodes and threads are also displayed, depending on the environment.

Procedure Task Timing

When the DETAILS option is specified in the PERFORMANCE statement, PROC HPTMINE produces a “Procedure Task Timing” table, which displays the elapsed time (absolute and relative) for the main tasks.

ODS Table Names

Each table that is created by the HPTMINE procedure has a name associated with it, and you must use this name to refer to the table when you use ODS statements. These names are listed in Table 4.8.

<table>
<thead>
<tr>
<th>Table Name</th>
<th>Description</th>
<th>Required Statement / Option</th>
</tr>
</thead>
<tbody>
<tr>
<td>PerformanceInfo</td>
<td>Information about the high-performance computing environment</td>
<td>Default output</td>
</tr>
<tr>
<td>Timing</td>
<td>Absolute and relative times for tasks performed by the procedure</td>
<td>DETAILS option in the PERFORMANCE statement</td>
</tr>
</tbody>
</table>

System Configuration

Prerequisites for Running PROC HPTMINE

To use the HPTMINE procedure, you must have a valid SAS Text Miner license, and the language binary files that are provided under that license must be available to PROC HPTMINE for parsing text. It is critical that your account for the encoding and text length of the character variables that you use for storing your text data, particularly for data sets that are predistributed on various databases. Different database systems might store character variables in different encodings, and the maximum length of character variables might also be different. Failure to configure your SAS client appropriately might result in character corruption or
truncated documents. To avoid any potential transcoding issue, you need to use a SAS session encoding that is compatible with the encoding that your text data uses. For more information about proper configuration, see the appropriate SAS/ACCESS documents.

## Accessing Input Data

When you run PROC HPTMINE in distributed mode, input data sets that are specified by the following options need to be accessible to the client machine:

- `MULTITERM=`
- `START=`
- `STOP=`
- `SYNONYM=`

When the HPTMINE procedure runs in distributed mode, it first reads these data sets on the client machine and then sends them to the grid.

## Configuring for Distributed Mode

When PROC HPTMINE runs in distributed mode, it needs to find the grid location of the language binary files that are used in parsing text. These binary files must be deployed to the grid, and the `GRID_TEXTANALYTICS_BIN_LOC` macro variable must be specified to indicate the location of the binary files on the grid.

### Deploying Language Binary Files on the Grid

The language binary files must be deployed to the grid in one of the following ways:

- For grid systems that consist of Linux nodes and run SAS Text Miner 13.1 or later, the SAS grid installation script can automatically install the language binary files.

- For earlier releases or other grid systems, you must copy the binary files from the `$SASROOT` folder to the grid either by placing the files in a location that is accessible by all nodes of the grid or by placing a copy of the files on each node of the grid. The binary files are very large, and a shared location requires less space to store them. However, using a shared location means that the files must be distributed to the grid when PROC HPTMINE runs, and this can be time-consuming. Consult your grid administrator for a recommended binary location on the grid.
Language Binary Files

If you have licensed SAS Text Miner, the language binary files are originally installed in the following locations:

- In a Windows environment, the files are installed in the subdirectory `tktg\sasmisc` under `$SASROOT`. For example: `C:\Program Files\SASHome\SASFoundation\9.4\tktg\sasmisc`.
- In a UNIX environment, the files are installed in `$SASROOT/misc/tktg`.

PROC HPTMINE currently supports only English. The following binary files are used for English processing:

- `en-compounds.txt`
- `en-ne.li`
- `en-sastags.txt`
- `en-utf8.stkzo`
- `en-utf8.tkzo`
- `en-utf8-AnalInfSinFas.mdic`
- `en-utf8-std.htagger`
- `case_mapping.bin`
- `chlangid.bin`
- `chmap.bin`
- `tix.config`
- `language-data.yml`

If you have licensed SAS Text Miner languages other than English, you will see other files in your installation directory. But you do not need to copy the files for other licensed languages to the grid.

The GRID_TEXTANALYTICS_BIN_LOC Macro

You can use the `GRID_TEXTANALYTICS_BIN_LOC` macro to tell the HPTMINE procedure where to find these binary files. This macro is required only when you run PROC HPTMINE on the grid. When you run PROC HPTMINE in single-machine mode, the language binary files that were installed in the `$SASROOT` location during your SAS Text Miner installation are used for parsing text.

Assume that the grid administrator has installed the language binary files to a directory named `/global_dir/tktg/misc`, which is accessible to all nodes of the grid. To tell PROC HPTMINE the location of the language binary files, insert the following statement before calling the procedure:

```sas
%let GRID_TEXTANALYTICS_BIN_LOC=/global_dir/tktg/misc;
```

When storage space permits, you can ensure optimal performance by placing a copy of the language binary files on each node and using a relative pathname for the `GRID_TEXTANALYTICS_BIN_LOC` macro variable.
For example, the grid administrator can create a directory whose pathname is /local_dir/tktg/misc on each node and can store the language binary files in that directory. When the following statement is specified, the HPTMINE procedure goes to the directory /local_dir/tktg/misc on each node to load the binary files:

```plaintext
%let GRID_TEXTANALYTICS_BIN_LOC=/local_dir/tktg/misc;
```

### Examples: HPTMINE Procedure

#### Example 4.1: Parsing with No Options Turned On

This example parses five documents, which are in a generated data set. The following DATA step generates the five documents:

```plaintext
/* 1) create data set */

data CarNominations;
  infile cards delimiter='|' missover;
  length text $70 ;
  input text$ i;
  cards;
    The Ford Taurus is the World Car of the Year. |1
    Hyundai won the award last year. |2
    Toyota sold the Toyota Tacoma in bright green. |3
    The Ford Taurus is sold in all colors except for lime green. |4
    The Honda Insight was World Car of the Year in 2008. |5
  ;
run;
```

The following statements run PROC HPTMINE to parse the documents. Because no host for distributed computing is specified and the NODES option is not specified in the PERFORMANCE statement, PROC HPTMINE runs automatically in soloist mode.

```plaintext
/* 2) starting code */
proc hptmine data=work.CarNominations;
  doc_id i;
  var text;
  parse
    nostemming notagging nonoungroups
    termwgt = none
    cellwgt = none
    reducef = 0
    entities = none
    outparent = outparent
    outterms = outterms
    outchild = outchild
    outconfig = outconfig
run;
```
Example 4.2: Parsing with Stemming Turned On

This example uses the data set that is generated in Example 4.1. The following statements run PROC HPTMINE to parse the documents. Because the NOSTEMMING option is not specified in the PARSE
statement, words are stemmed (the default).

```sas
proc hptmine data=work.CarNominations;
   doc_id i;
   var text;
   parse
       notagging nonoungroups
       termwgt = none
       cellwgt = none
       reducef = 0
       entities = none

       outparent= outparent
       outterms = outterms
       outchild = outchild
       outconfig= outconfig
   ;
   performance details;
run;
```

Output 4.2.1 shows the content of the outterms data set. In this example, words are stemmed. You can see that the term “sold” now stems to have the parent term “sell.” Also, the outparent data set and the outchild data set are different. The parent term “sell” shows up in outparent (key=11), but not the child term “sold” (key=26). Only “sold” appears in the outchild data set, and “sell” does not appear.
### Output 4.2.1 The outterms Data Set with Stemming

<table>
<thead>
<tr>
<th>Obs</th>
<th>Term</th>
<th>Role</th>
<th>Attribute</th>
<th>Freq</th>
<th>numdocs</th>
<th>_keep</th>
<th>Key</th>
<th>Parent</th>
<th>id</th>
<th>_ispar</th>
<th>Weight</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>.</td>
<td>Punct</td>
<td>N</td>
<td>30</td>
<td>.</td>
<td>30</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>2</td>
<td>all</td>
<td>Alpha</td>
<td>Y</td>
<td>1</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>3</td>
<td>toyota</td>
<td>Alpha</td>
<td>Y</td>
<td>2</td>
<td>.</td>
<td>2</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>4</td>
<td>ford</td>
<td>Alpha</td>
<td>Y</td>
<td>3</td>
<td>.</td>
<td>3</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>5</td>
<td>tacoma</td>
<td>Alpha</td>
<td>Y</td>
<td>4</td>
<td>.</td>
<td>4</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>6</td>
<td>year</td>
<td>Alpha</td>
<td>Y</td>
<td>5</td>
<td>.</td>
<td>5</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>7</td>
<td>taurus</td>
<td>Alpha</td>
<td>Y</td>
<td>6</td>
<td>.</td>
<td>6</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>8</td>
<td>won</td>
<td>Alpha</td>
<td>Y</td>
<td>7</td>
<td>.</td>
<td>7</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>9</td>
<td>honda</td>
<td>Alpha</td>
<td>Y</td>
<td>8</td>
<td>.</td>
<td>8</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>10</td>
<td>bright</td>
<td>Alpha</td>
<td>Y</td>
<td>9</td>
<td>.</td>
<td>9</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>11</td>
<td>be</td>
<td>Alpha</td>
<td>Y</td>
<td>10</td>
<td>10</td>
<td>+ 1</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>12</td>
<td>sold</td>
<td>Alpha</td>
<td>Y</td>
<td>11</td>
<td>11</td>
<td>.</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>13</td>
<td>sell</td>
<td>Alpha</td>
<td>Y</td>
<td>11</td>
<td>11</td>
<td>+ 1</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>14</td>
<td>colors</td>
<td>Alpha</td>
<td>Y</td>
<td>23</td>
<td>23</td>
<td>.</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>15</td>
<td>lime</td>
<td>Alpha</td>
<td>Y</td>
<td>12</td>
<td>.</td>
<td>12</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>16</td>
<td>except</td>
<td>Alpha</td>
<td>Y</td>
<td>13</td>
<td>.</td>
<td>13</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>17</td>
<td>hyundai</td>
<td>Alpha</td>
<td>Y</td>
<td>14</td>
<td>.</td>
<td>14</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>18</td>
<td>in</td>
<td>Alpha</td>
<td>Y</td>
<td>15</td>
<td>.</td>
<td>15</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>19</td>
<td>is</td>
<td>Alpha</td>
<td>Y</td>
<td>10</td>
<td>10</td>
<td>.</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>20</td>
<td>for</td>
<td>Alpha</td>
<td>Y</td>
<td>16</td>
<td>.</td>
<td>16</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>21</td>
<td>world</td>
<td>Alpha</td>
<td>Y</td>
<td>17</td>
<td>.</td>
<td>17</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>22</td>
<td>green</td>
<td>Alpha</td>
<td>Y</td>
<td>18</td>
<td>.</td>
<td>18</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>23</td>
<td>the</td>
<td>Alpha</td>
<td>Y</td>
<td>19</td>
<td>.</td>
<td>19</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>24</td>
<td>of</td>
<td>Alpha</td>
<td>Y</td>
<td>20</td>
<td>.</td>
<td>20</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>25</td>
<td>award</td>
<td>Alpha</td>
<td>Y</td>
<td>21</td>
<td>.</td>
<td>21</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>26</td>
<td>was</td>
<td>Alpha</td>
<td>Y</td>
<td>10</td>
<td>10</td>
<td>.</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>27</td>
<td>2008</td>
<td>Num</td>
<td>N</td>
<td>31</td>
<td>.</td>
<td>31</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>28</td>
<td>car</td>
<td>Alpha</td>
<td>Y</td>
<td>22</td>
<td>.</td>
<td>22</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>29</td>
<td>color</td>
<td>Alpha</td>
<td>Y</td>
<td>23</td>
<td>23</td>
<td>+ 1</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>30</td>
<td>insight</td>
<td>Alpha</td>
<td>Y</td>
<td>24</td>
<td>.</td>
<td>24</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>31</td>
<td>last</td>
<td>Alpha</td>
<td>Y</td>
<td>25</td>
<td>.</td>
<td>25</td>
<td>.</td>
<td>.</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
</tbody>
</table>
Example 4.3: Adding Entities and Noun Groups

This example uses the data set that is generated in Example 4.1. The following statements run PROC HPTMINE to parse the documents. Because the NONOUNGROUP option is not specified in the PARSE statement, noun group extraction is turned on, and because the ENTITIES option is not specified, entity identification is turned on.

``` Sas
proc hptmine data=work.CarNominations;
  doc_id i;
  var text;
  PARSE
    notagging
    termwgt = none
    cellwgt = none
    reducef = 0
    entities = std
    outparent = outparent
    outterms = outterms
    outchild = outchild
    outconfig = outconfig
  ;
  performance details;
run;

proc print data=outterms; title "OUTTERMS Data Set"; run;
```

Output 4.3.1 shows the content of the outterms data set. Compared to Output 4.2.1, the outterms data set is longer, because it contains entities and noun groups. For example, “honda insight” is included in the outterms data set as an entity with Role=Vehicle, and “bright green” is also included in the outterms data set as a noun group.
### Output 4.3.1 The outterms Data Set with Noun Group Extraction and Entity Identification

<table>
<thead>
<tr>
<th>Obs</th>
<th>Term</th>
<th>Role</th>
<th>Attribute</th>
<th>Freq</th>
<th>numdocs</th>
<th>_keep</th>
<th>Key</th>
<th>Parent</th>
<th>id</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>lime green</td>
<td>NOUN_GROUP</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>2</td>
<td>.</td>
<td>Punct</td>
<td>5</td>
<td>1</td>
<td>N</td>
<td>37</td>
<td>.</td>
<td>37</td>
<td>1</td>
</tr>
<tr>
<td>3</td>
<td>all</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>2</td>
<td>.</td>
<td>2</td>
<td>2</td>
</tr>
<tr>
<td>4</td>
<td>toyota</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>3</td>
<td>.</td>
<td>3</td>
<td>3</td>
</tr>
<tr>
<td>5</td>
<td>ford</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>4</td>
<td>.</td>
<td>4</td>
<td>4</td>
</tr>
<tr>
<td>6</td>
<td>toyota tacoma</td>
<td>VEHICLE</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>5</td>
<td>.</td>
<td>5</td>
</tr>
<tr>
<td>7</td>
<td>tacoma</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>6</td>
<td>.</td>
<td>6</td>
<td>6</td>
</tr>
<tr>
<td>8</td>
<td>year</td>
<td>Alpha</td>
<td>3</td>
<td>1</td>
<td>Y</td>
<td>7</td>
<td>.</td>
<td>7</td>
<td>7</td>
</tr>
<tr>
<td>9</td>
<td>taurus</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>8</td>
<td>.</td>
<td>8</td>
<td>8</td>
</tr>
<tr>
<td>10</td>
<td>won</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>9</td>
<td>.</td>
<td>9</td>
<td>9</td>
</tr>
<tr>
<td>11</td>
<td>honda</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>10</td>
<td>.</td>
<td>10</td>
<td>10</td>
</tr>
<tr>
<td>12</td>
<td>bright</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>11</td>
<td>.</td>
<td>11</td>
<td>11</td>
</tr>
<tr>
<td>13</td>
<td>be</td>
<td>Alpha</td>
<td>3</td>
<td>1</td>
<td>Y</td>
<td>12</td>
<td>12</td>
<td>12</td>
<td>12</td>
</tr>
<tr>
<td>14</td>
<td>sold</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>33</td>
<td>15</td>
<td>15</td>
<td>15</td>
</tr>
<tr>
<td>15</td>
<td>bright green</td>
<td>NOUN_GROUP</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>13</td>
<td>.</td>
<td>13</td>
</tr>
<tr>
<td>16</td>
<td>ford taurus</td>
<td>VEHICLE</td>
<td>Entity</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>14</td>
<td>.</td>
<td>14</td>
</tr>
<tr>
<td>17</td>
<td>sell</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>15</td>
<td>15</td>
<td>15</td>
<td>15</td>
</tr>
<tr>
<td>18</td>
<td>hyundai</td>
<td>COMPANY</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>16</td>
<td>.</td>
<td>16</td>
</tr>
<tr>
<td>19</td>
<td>colors</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>34</td>
<td>30</td>
<td>30</td>
<td>30</td>
</tr>
<tr>
<td>20</td>
<td>lime</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>17</td>
<td>.</td>
<td>17</td>
<td>17</td>
</tr>
<tr>
<td>21</td>
<td>except</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>18</td>
<td>.</td>
<td>18</td>
<td>18</td>
</tr>
<tr>
<td>22</td>
<td>honda insight</td>
<td>VEHICLE</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>19</td>
<td>.</td>
<td>19</td>
</tr>
<tr>
<td>23</td>
<td>in</td>
<td>Alpha</td>
<td>3</td>
<td>1</td>
<td>Y</td>
<td>20</td>
<td>.</td>
<td>20</td>
<td>20</td>
</tr>
<tr>
<td>24</td>
<td>is</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>35</td>
<td>12</td>
<td>12</td>
<td>12</td>
</tr>
<tr>
<td>25</td>
<td>for</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>21</td>
<td>.</td>
<td>21</td>
<td>21</td>
</tr>
<tr>
<td>26</td>
<td>toyota</td>
<td>COMPANY</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>22</td>
<td>.</td>
<td>22</td>
</tr>
<tr>
<td>27</td>
<td>world</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>23</td>
<td>.</td>
<td>23</td>
<td>23</td>
</tr>
<tr>
<td>28</td>
<td>green</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>24</td>
<td>.</td>
<td>24</td>
<td>24</td>
</tr>
<tr>
<td>29</td>
<td>the</td>
<td>Alpha</td>
<td>8</td>
<td>1</td>
<td>Y</td>
<td>25</td>
<td>.</td>
<td>25</td>
<td>25</td>
</tr>
<tr>
<td>30</td>
<td>of</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>26</td>
<td>.</td>
<td>26</td>
<td>26</td>
</tr>
<tr>
<td>31</td>
<td>world car</td>
<td>PROP_MISC</td>
<td>Entity</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>27</td>
<td>.</td>
<td>27</td>
</tr>
<tr>
<td>32</td>
<td>award</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>28</td>
<td>.</td>
<td>28</td>
<td>28</td>
</tr>
<tr>
<td>33</td>
<td>was</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>36</td>
<td>12</td>
<td>12</td>
<td>12</td>
</tr>
<tr>
<td>34</td>
<td>2008</td>
<td>Num</td>
<td>1</td>
<td>1</td>
<td>N</td>
<td>38</td>
<td>.</td>
<td>38</td>
<td>38</td>
</tr>
<tr>
<td>35</td>
<td>car</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>29</td>
<td>.</td>
<td>29</td>
<td>29</td>
</tr>
<tr>
<td>36</td>
<td>color</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>30</td>
<td>30</td>
<td>30</td>
<td>30</td>
</tr>
<tr>
<td>37</td>
<td>insight</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>31</td>
<td>.</td>
<td>31</td>
<td>31</td>
</tr>
<tr>
<td>38</td>
<td>last</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>32</td>
<td>.</td>
<td>32</td>
<td>32</td>
</tr>
</tbody>
</table>
Example 4.4: Adding Part-of-Speech Tagging

This example uses the data set that is generated in Example 4.1. The following statements run PROC HPTMINE to parse the documents. Because the NOTAGGING option is not specified in the PARSE statement, PROC HPTMINE uses context clues to determine a term’s part of speech.

```plaintext
proc hptmine data=work.CarNominations;
  doc_id i;
  var text;
  PARSE
    termwgt  = none
    cellwgt  = none
    reducef  = 0
    entities = std
    outparent = outparent
    outterms  = outterms
    outchild  = outchild
    outconfig = outconfig
  ;
  performance details;
run;

proc print data=outterms; title "OUTTERMS Data Set"; run;
```

Output 4.4.1 shows the content of the outterms data set. Compared to Output 4.3.1, the outterms data set also contains the part-of-speech tag for the terms.
### Example 4.4: Adding Part-of-Speech Tagging

Output 4.4.1 The outterms Data Set with Part-of-Speech Tagging

<table>
<thead>
<tr>
<th>Obs</th>
<th>Term</th>
<th>Role</th>
<th>Attribute</th>
<th>Freq</th>
<th>numdocs</th>
<th>_keep</th>
<th>Key</th>
<th>Parent</th>
<th>id</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>car</td>
<td>Prop</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>2</td>
<td>lime green</td>
<td>NOUN_GROUP</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>2</td>
<td>.</td>
<td>2</td>
</tr>
<tr>
<td>3</td>
<td>last</td>
<td>Adj</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>3</td>
<td>.</td>
<td>3</td>
</tr>
<tr>
<td>4</td>
<td>lime</td>
<td>Adj</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>4</td>
<td>.</td>
<td>4</td>
</tr>
<tr>
<td>5</td>
<td>sell</td>
<td>Verb</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>5</td>
<td>5</td>
<td>5</td>
</tr>
<tr>
<td>6</td>
<td>won</td>
<td>Noun</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>6</td>
<td>.</td>
<td>6</td>
</tr>
<tr>
<td>7</td>
<td>toyota tacoma</td>
<td>VEHICLE</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>7</td>
<td>.</td>
<td>7</td>
</tr>
<tr>
<td>8</td>
<td>bright</td>
<td>Adj</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>8</td>
<td>.</td>
<td>8</td>
</tr>
<tr>
<td>9</td>
<td>honda</td>
<td>Prop</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>9</td>
<td>.</td>
<td>9</td>
</tr>
<tr>
<td>10</td>
<td>sold</td>
<td>Verb</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>34</td>
<td>5</td>
<td>5</td>
</tr>
<tr>
<td>11</td>
<td>colors</td>
<td>Noun</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>35</td>
<td>12</td>
<td>12</td>
</tr>
<tr>
<td>12</td>
<td>for</td>
<td>Prep</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>10</td>
<td>.</td>
<td>10</td>
</tr>
<tr>
<td>13</td>
<td>award</td>
<td>Verb</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>11</td>
<td>.</td>
<td>11</td>
</tr>
<tr>
<td>14</td>
<td>color</td>
<td>Noun</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>12</td>
<td>12</td>
<td>12</td>
</tr>
<tr>
<td>15</td>
<td>2008</td>
<td>Num</td>
<td>Num</td>
<td>1</td>
<td>1</td>
<td>N</td>
<td>38</td>
<td>.</td>
<td>38</td>
</tr>
<tr>
<td>16</td>
<td>bright green</td>
<td>NOUN_GROUP</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>13</td>
<td>.</td>
<td>13</td>
</tr>
<tr>
<td>17</td>
<td>the</td>
<td>Det</td>
<td>Alpha</td>
<td>8</td>
<td>1</td>
<td>Y</td>
<td>14</td>
<td>.</td>
<td>14</td>
</tr>
<tr>
<td>18</td>
<td>is</td>
<td>Verb</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>36</td>
<td>26</td>
<td>26</td>
</tr>
<tr>
<td>19</td>
<td>ford taurus</td>
<td>VEHICLE</td>
<td>Entity</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>15</td>
<td>.</td>
<td>15</td>
</tr>
<tr>
<td>20</td>
<td>hyundai</td>
<td>COMPANY</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>16</td>
<td>.</td>
<td>16</td>
</tr>
<tr>
<td>21</td>
<td>toyota</td>
<td>Prop</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>17</td>
<td>.</td>
<td>17</td>
</tr>
<tr>
<td>22</td>
<td>car</td>
<td>Noun</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>18</td>
<td>.</td>
<td>18</td>
</tr>
<tr>
<td>23</td>
<td>except</td>
<td>Verb</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>19</td>
<td>.</td>
<td>19</td>
</tr>
<tr>
<td>24</td>
<td>in</td>
<td>Prep</td>
<td>Alpha</td>
<td>3</td>
<td>1</td>
<td>Y</td>
<td>20</td>
<td>.</td>
<td>20</td>
</tr>
<tr>
<td>25</td>
<td>was</td>
<td>Verb</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>37</td>
<td>26</td>
<td>26</td>
</tr>
<tr>
<td>26</td>
<td>honda insight</td>
<td>VEHICLE</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>21</td>
<td>.</td>
<td>21</td>
</tr>
<tr>
<td>27</td>
<td>insight</td>
<td>Noun</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>22</td>
<td>.</td>
<td>22</td>
</tr>
<tr>
<td>28</td>
<td>ford</td>
<td>Prop</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>23</td>
<td>.</td>
<td>23</td>
</tr>
<tr>
<td>29</td>
<td>world</td>
<td>Prop</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>24</td>
<td>.</td>
<td>24</td>
</tr>
<tr>
<td>30</td>
<td>.</td>
<td>Punct</td>
<td>Punct</td>
<td>5</td>
<td>1</td>
<td>N</td>
<td>39</td>
<td>.</td>
<td>39</td>
</tr>
<tr>
<td>31</td>
<td>year</td>
<td>Noun</td>
<td>Alpha</td>
<td>3</td>
<td>1</td>
<td>Y</td>
<td>25</td>
<td>.</td>
<td>25</td>
</tr>
<tr>
<td>32</td>
<td>be</td>
<td>Verb</td>
<td>Alpha</td>
<td>3</td>
<td>1</td>
<td>Y</td>
<td>24</td>
<td>26</td>
<td>26</td>
</tr>
<tr>
<td>33</td>
<td>all</td>
<td>Adj</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>27</td>
<td>.</td>
<td>27</td>
</tr>
<tr>
<td>34</td>
<td>toyota</td>
<td>COMPANY</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>28</td>
<td>.</td>
<td>28</td>
</tr>
<tr>
<td>35</td>
<td>taurus</td>
<td>Noun</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>29</td>
<td>.</td>
<td>29</td>
</tr>
<tr>
<td>36</td>
<td>world car</td>
<td>PROP_MISC</td>
<td>Entity</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>30</td>
<td>.</td>
<td>30</td>
</tr>
<tr>
<td>37</td>
<td>of</td>
<td>Prep</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>31</td>
<td>.</td>
<td>31</td>
</tr>
<tr>
<td>38</td>
<td>tacoma</td>
<td>Prop</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>32</td>
<td>.</td>
<td>32</td>
</tr>
<tr>
<td>39</td>
<td>green</td>
<td>Noun</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>33</td>
<td>.</td>
<td>33</td>
</tr>
</tbody>
</table>
Example 4.5: Adding Synonyms

This example uses the data set that is generated in Example 4.1. So far, by looking at the outterms data sets that are generated by Example 4.1 to Example 4.4, you can see that the data are very “vehicle focused.” But suppose what you really care about are the Companies. You can use a synonym list in parsing to map each vehicle to the company that produces it. The following statements show this mapping.

```sas
data synds;
  infile cards delimiter=',';
  length TERM $13;
  input TERM $ TERMROLE $ PARENT $ parentrole$;
cards;
  honda insight, VEHICLE, honda, COMPANY,
  ford taurus, VEHICLE, ford, COMPANY,
  toyota tacoma, VEHICLE, toyota, COMPANY,
run;

proc hptmine data=work.CarNominations;
  doc_id i;
  var text;
  parse
    termwgt = none
    cellwgt = none
    reducef = 0
    entities = std
    synonym = synds
    outparent = outparent
    outterms = outterms
    outchild = outchild
    outconfig = outconfig
  ;
  PERFORMANCE details;
RUN;

proc print data=outterms; title "OUTTERMS Data Set"; run;
```

Output 4.5.1 shows the content of the outterms data set. You can see that the term “honda insight” (key=39) is assigned the parent term “honda” (key=5). Only the term “honda” appears in the outparent data set.
### Output 4.5.1  The outterms Data Set with Synonym Mapping

<table>
<thead>
<tr>
<th>Obs</th>
<th>Term</th>
<th>Role</th>
<th>Attribute</th>
<th>Freq</th>
<th>numdocs</th>
<th>_keep</th>
<th>Key</th>
<th>Parent</th>
<th>id</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>car</td>
<td>Prop</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>2</td>
<td>lime green</td>
<td>NOUN_GROUP</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>2</td>
<td>.</td>
<td>2</td>
</tr>
<tr>
<td>3</td>
<td>last</td>
<td>Adj</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>3</td>
<td>.</td>
<td>3</td>
</tr>
<tr>
<td>4</td>
<td>lime</td>
<td>Adj</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>4</td>
<td>.</td>
<td>4</td>
</tr>
<tr>
<td>5</td>
<td>honda</td>
<td>COMPANY</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>5</td>
<td>5</td>
<td>5</td>
</tr>
<tr>
<td>6</td>
<td>sell</td>
<td>Verb</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>6</td>
<td>6</td>
<td>6</td>
</tr>
<tr>
<td>7</td>
<td>won</td>
<td>Noun</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>7</td>
<td>.</td>
<td>7</td>
</tr>
<tr>
<td>8</td>
<td>toyota tacoma</td>
<td>VEHICLE</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>33</td>
<td>26</td>
<td>26</td>
</tr>
<tr>
<td>9</td>
<td>bright</td>
<td>Adj</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>8</td>
<td>.</td>
<td>8</td>
</tr>
<tr>
<td>10</td>
<td>honda</td>
<td>Prop</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>9</td>
<td>.</td>
<td>9</td>
</tr>
<tr>
<td>11</td>
<td>sold</td>
<td>Verb</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>34</td>
<td>6</td>
<td>6</td>
</tr>
<tr>
<td>12</td>
<td>colors</td>
<td>Noun</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>35</td>
<td>12</td>
<td>12</td>
</tr>
<tr>
<td>13</td>
<td>for</td>
<td>Prep</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>10</td>
<td>.</td>
<td>10</td>
</tr>
<tr>
<td>14</td>
<td>award</td>
<td>Verb</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>11</td>
<td>.</td>
<td>11</td>
</tr>
<tr>
<td>15</td>
<td>color</td>
<td>Noun</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>12</td>
<td>12</td>
<td>12</td>
</tr>
<tr>
<td>16</td>
<td>2008</td>
<td>Num</td>
<td>Num</td>
<td>1</td>
<td>1</td>
<td>N</td>
<td>40</td>
<td>.</td>
<td>40</td>
</tr>
<tr>
<td>17</td>
<td>bright green</td>
<td>NOUN_GROUP</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>13</td>
<td>.</td>
<td>13</td>
</tr>
<tr>
<td>18</td>
<td>the</td>
<td>Det</td>
<td>Alpha</td>
<td>8</td>
<td>1</td>
<td>Y</td>
<td>14</td>
<td>.</td>
<td>14</td>
</tr>
<tr>
<td>19</td>
<td>is</td>
<td>Verb</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>36</td>
<td>24</td>
<td>24</td>
</tr>
<tr>
<td>20</td>
<td>ford taurus</td>
<td>VEHICLE</td>
<td>Entity</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>37</td>
<td>27</td>
<td>27</td>
</tr>
<tr>
<td>21</td>
<td>hyundai</td>
<td>COMPANY</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>15</td>
<td>.</td>
<td>15</td>
</tr>
<tr>
<td>22</td>
<td>toyota</td>
<td>Prop</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>16</td>
<td>.</td>
<td>16</td>
</tr>
<tr>
<td>23</td>
<td>car</td>
<td>Noun</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>17</td>
<td>.</td>
<td>17</td>
</tr>
<tr>
<td>24</td>
<td>except</td>
<td>Verb</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>18</td>
<td>.</td>
<td>18</td>
</tr>
<tr>
<td>25</td>
<td>in</td>
<td>Prep</td>
<td>Alpha</td>
<td>3</td>
<td>1</td>
<td>Y</td>
<td>19</td>
<td>.</td>
<td>19</td>
</tr>
<tr>
<td>26</td>
<td>was</td>
<td>Verb</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>38</td>
<td>24</td>
<td>24</td>
</tr>
<tr>
<td>27</td>
<td>honda insight</td>
<td>VEHICLE</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>39</td>
<td>5</td>
<td>5</td>
</tr>
<tr>
<td>28</td>
<td>insight</td>
<td>Noun</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>20</td>
<td>.</td>
<td>20</td>
</tr>
<tr>
<td>29</td>
<td>ford</td>
<td>Prop</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>21</td>
<td>.</td>
<td>21</td>
</tr>
<tr>
<td>30</td>
<td>world</td>
<td>Prop</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>22</td>
<td>.</td>
<td>22</td>
</tr>
<tr>
<td>31</td>
<td>.</td>
<td>Punct</td>
<td>Punct</td>
<td>5</td>
<td>1</td>
<td>N</td>
<td>41</td>
<td>.</td>
<td>41</td>
</tr>
<tr>
<td>32</td>
<td>year</td>
<td>Noun</td>
<td>Alpha</td>
<td>3</td>
<td>1</td>
<td>Y</td>
<td>23</td>
<td>.</td>
<td>23</td>
</tr>
<tr>
<td>33</td>
<td>be</td>
<td>Verb</td>
<td>Alpha</td>
<td>3</td>
<td>1</td>
<td>Y</td>
<td>24</td>
<td>24</td>
<td>24</td>
</tr>
<tr>
<td>34</td>
<td>all</td>
<td>Adj</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>25</td>
<td>.</td>
<td>25</td>
</tr>
<tr>
<td>35</td>
<td>toyota</td>
<td>COMPANY</td>
<td>Entity</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>26</td>
<td>26</td>
<td>26</td>
</tr>
<tr>
<td>36</td>
<td>toyota</td>
<td>COMPANY</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>26</td>
<td>.</td>
<td>26</td>
</tr>
<tr>
<td>37</td>
<td>ford</td>
<td>COMPANY</td>
<td>Entity</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>27</td>
<td>27</td>
<td>27</td>
</tr>
<tr>
<td>38</td>
<td>taurus</td>
<td>Noun</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>28</td>
<td>.</td>
<td>28</td>
</tr>
<tr>
<td>39</td>
<td>world car</td>
<td>PROP_MISC</td>
<td>Entity</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>29</td>
<td>.</td>
<td>29</td>
</tr>
<tr>
<td>40</td>
<td>of</td>
<td>Prep</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>30</td>
<td>.</td>
<td>30</td>
</tr>
<tr>
<td>41</td>
<td>tacoma</td>
<td>Prop</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>31</td>
<td>.</td>
<td>31</td>
</tr>
<tr>
<td>42</td>
<td>green</td>
<td>Noun</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>32</td>
<td>.</td>
<td>32</td>
</tr>
</tbody>
</table>
Example 4.6: Adding a Stop List

This example uses the data set that is generated in Example 4.1. If you want to eliminate TOYOTA from the analysis, you can enter the parent term “Toyota” with role=COMPANY in a stop list. When this stop list is an input, PROC HPTMINE drops the term “Toyota” (role=COMPANY) and all its children terms in the outterms data set by marking Keep=N for these terms.

```sas
data synds;
infile cards delimiter=',';
length TERM $13;
input TERM $ TERMROLE $ PARENT $ parentrole$;
cards;
   honda insight, VEHICLE, honda, COMPANY,
   ford taurus, VEHICLE, ford, COMPANY,
   toyota tacoma, VEHICLE, toyota, COMPANY,
;run;

/* create stop list*/
data stopList;
infile cards delimiter='|' missover;
length term $25 role $40;
input term$ role$ ;
cards;
   toyota| COMPANY
;run;

proc hptmine data=work.CarNominations;
doc_id i;
var text;
parse
   termwgt = none
   cellwgt = none
   reducef = 0
   entities = std
   synonym = synds
   stop = stopList
   outparent = outparent
   outterms = outterms
   outchild = outchild
   outconfig = outconfig
   ;
performance details;
run;

proc print data=outterms; title "OUTTERMS Data Set"; run;
```
Example 4.6: Adding a Stop List

Output 4.6.1 shows the content of the outterms data set. You can see that the term “Toyota” (key=40) and the term “Toyota Tacoma” (key=41) are assigned Keep=N. The outparent data set is shorter than the one generated in Example 4.5.

**Output 4.6.1** The outterms Data Set Filtered Using Stop List

<table>
<thead>
<tr>
<th>Obs</th>
<th>Term</th>
<th>Role</th>
<th>Attribute</th>
<th>Freq</th>
<th>numdocs</th>
<th>_keep</th>
<th>Key</th>
<th>Parent</th>
<th>id</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>car</td>
<td>Prop</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>1</td>
<td>.</td>
<td>1</td>
</tr>
<tr>
<td>2</td>
<td>lime green</td>
<td>NOUN_GROUP</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>2</td>
<td>.</td>
<td>2</td>
</tr>
<tr>
<td>3</td>
<td>last</td>
<td>Adj</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>3</td>
<td>.</td>
<td>3</td>
</tr>
<tr>
<td>4</td>
<td>lime</td>
<td>Adj</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>4</td>
<td>.</td>
<td>4</td>
</tr>
<tr>
<td>5</td>
<td>honda</td>
<td>COMPANY</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>5</td>
<td>5</td>
<td>5</td>
</tr>
<tr>
<td>6</td>
<td>sell</td>
<td>Verb</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>6</td>
<td>6</td>
<td>6</td>
</tr>
<tr>
<td>7</td>
<td>won</td>
<td>Noun</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>7</td>
<td>.</td>
<td>7</td>
</tr>
<tr>
<td>8</td>
<td>toyota tacoma</td>
<td>VEHICLE</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>N</td>
<td>41</td>
<td>40</td>
<td>40</td>
</tr>
<tr>
<td>9</td>
<td>bright</td>
<td>Adj</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>8</td>
<td>.</td>
<td>8</td>
</tr>
<tr>
<td>10</td>
<td>honda</td>
<td>Prop</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>9</td>
<td>.</td>
<td>9</td>
</tr>
<tr>
<td>11</td>
<td>sold</td>
<td>Verb</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>31</td>
<td>6</td>
<td>6</td>
</tr>
<tr>
<td>12</td>
<td>colors</td>
<td>Noun</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>32</td>
<td>12</td>
<td>12</td>
</tr>
<tr>
<td>13</td>
<td>for</td>
<td>Prep</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>10</td>
<td>.</td>
<td>10</td>
</tr>
<tr>
<td>14</td>
<td>award</td>
<td>Verb</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>11</td>
<td>.</td>
<td>11</td>
</tr>
<tr>
<td>15</td>
<td>color</td>
<td>Noun</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>12</td>
<td>12</td>
<td>12</td>
</tr>
<tr>
<td>16</td>
<td>2008</td>
<td>Num</td>
<td>Num</td>
<td>1</td>
<td>1</td>
<td>N</td>
<td>37</td>
<td>.</td>
<td>37</td>
</tr>
<tr>
<td>17</td>
<td>bright green</td>
<td>NOUN_GROUP</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>13</td>
<td>.</td>
<td>13</td>
</tr>
<tr>
<td>18</td>
<td>the</td>
<td>Det</td>
<td>Alpha</td>
<td>8</td>
<td>1</td>
<td>Y</td>
<td>14</td>
<td>.</td>
<td>14</td>
</tr>
<tr>
<td>19</td>
<td>is</td>
<td>Verb</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>23</td>
<td>23</td>
<td>23</td>
</tr>
<tr>
<td>20</td>
<td>ford taurus</td>
<td>VEHICLE</td>
<td>Entity</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>34</td>
<td>25</td>
<td>25</td>
</tr>
<tr>
<td>21</td>
<td>hyundai</td>
<td>COMPANY</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>15</td>
<td>.</td>
<td>15</td>
</tr>
<tr>
<td>22</td>
<td>toyota</td>
<td>Prop</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>N</td>
<td>38</td>
<td>.</td>
<td>38</td>
</tr>
<tr>
<td>23</td>
<td>car</td>
<td>Noun</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>16</td>
<td>.</td>
<td>16</td>
</tr>
<tr>
<td>24</td>
<td>except</td>
<td>Verb</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>17</td>
<td>.</td>
<td>17</td>
</tr>
<tr>
<td>25</td>
<td>in</td>
<td>Prep</td>
<td>Alpha</td>
<td>3</td>
<td>1</td>
<td>Y</td>
<td>18</td>
<td>.</td>
<td>18</td>
</tr>
<tr>
<td>26</td>
<td>was</td>
<td>Verb</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>35</td>
<td>23</td>
<td>23</td>
</tr>
<tr>
<td>27</td>
<td>honda insight</td>
<td>VEHICLE</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>36</td>
<td>5</td>
<td>5</td>
</tr>
<tr>
<td>28</td>
<td>insight</td>
<td>Noun</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>19</td>
<td>.</td>
<td>19</td>
</tr>
<tr>
<td>29</td>
<td>ford</td>
<td>Prop</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>20</td>
<td>.</td>
<td>20</td>
</tr>
<tr>
<td>30</td>
<td>world</td>
<td>Prop</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>21</td>
<td>.</td>
<td>21</td>
</tr>
<tr>
<td>31</td>
<td>.</td>
<td>Punct</td>
<td>Punct</td>
<td>5</td>
<td>1</td>
<td>N</td>
<td>39</td>
<td>.</td>
<td>39</td>
</tr>
<tr>
<td>32</td>
<td>year</td>
<td>Noun</td>
<td>Alpha</td>
<td>3</td>
<td>1</td>
<td>Y</td>
<td>22</td>
<td>.</td>
<td>22</td>
</tr>
<tr>
<td>33</td>
<td>be</td>
<td>Verb</td>
<td>Alpha</td>
<td>3</td>
<td>1</td>
<td>Y</td>
<td>23</td>
<td>23</td>
<td>23</td>
</tr>
<tr>
<td>34</td>
<td>all</td>
<td>Adj</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>24</td>
<td>.</td>
<td>24</td>
</tr>
<tr>
<td>35</td>
<td>toyota</td>
<td>COMPANY</td>
<td>Entity</td>
<td>2</td>
<td>1</td>
<td>N</td>
<td>40</td>
<td>40</td>
<td>40</td>
</tr>
<tr>
<td>36</td>
<td>toyota</td>
<td>COMPANY</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>N</td>
<td>40</td>
<td>.</td>
<td>40</td>
</tr>
<tr>
<td>37</td>
<td>ford</td>
<td>COMPANY</td>
<td>Entity</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>25</td>
<td>25</td>
<td>25</td>
</tr>
<tr>
<td>38</td>
<td>taurus</td>
<td>Noun</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>26</td>
<td>.</td>
<td>26</td>
</tr>
<tr>
<td>39</td>
<td>world car</td>
<td>PROP_MISC</td>
<td>Entity</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>27</td>
<td>.</td>
<td>27</td>
</tr>
<tr>
<td>40</td>
<td>of</td>
<td>Prep</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>28</td>
<td>.</td>
<td>28</td>
</tr>
<tr>
<td>41</td>
<td>tacoma</td>
<td>Prop</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>29</td>
<td>.</td>
<td>29</td>
</tr>
<tr>
<td>42</td>
<td>green</td>
<td>Noun</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>30</td>
<td>.</td>
<td>30</td>
</tr>
</tbody>
</table>
Example 4.7: Adding a Multiterm List

A multiterm list can be used to define terms that consist of multiple words. This example uses the data set that is generated in Example 4.1 to show how to use MULTITERM option.

data synds;
infile cards delimiter=',';
length TERM $13;
input TERM $ TERMROLE $ PARENT $ parentrole$;
cards;
    honda insight, VEHICLE, honda, COMPANY,
    ford taurus, VEHICLE, ford, COMPANY,
    toyota tacoma, VEHICLE, toyota, COMPANY,
; run;

data stopList;
infile cards delimiter='|' missover;
length term $25 role $40;
input term$ role$ ;
cards;
    toyota| COMPANY
; run;

/* create multiterms list*/
data _null_;
    FILE 'encomp.txt' ;
    PUT 'except for :3:Prep';
run;

proc hptmine data=work.CarNominations;
doc_id i;
var text;
parse
    termwgt = none
    cellwgt = none
    reducef = 0
    entities = std
    synonym = synds
    stop = stopList
    multiterm = "encomp.txt"
    outparent = outparent
    outterms = outterms
    outchild = outchild
    outconfig = outconfig
; performance details;
run;

proc print data=outterms; title "OUTTERMS Data Set"; run;
Output 4.7.1 shows the content of the outterms data set. In the preceding statements, “except for” is defined as an individual term. In the outterms data set, you can see that the two terms “except” and “for” have become one term, “except for.”

### Output 4.7.1 The outterms Data Set

<table>
<thead>
<tr>
<th>Obs</th>
<th>Term</th>
<th>Role</th>
<th>Attribute</th>
<th>Freq</th>
<th>numdocs</th>
<th>_keep</th>
<th>Key</th>
<th>Parent</th>
<th>id</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>car</td>
<td>Prop</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>1</td>
<td>1</td>
<td>1</td>
</tr>
<tr>
<td>2</td>
<td>lime green</td>
<td>NOUN_GROUP</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>2</td>
<td>2</td>
<td>2</td>
</tr>
<tr>
<td>3</td>
<td>last</td>
<td>Adj</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>3</td>
<td>3</td>
<td>3</td>
</tr>
<tr>
<td>4</td>
<td>lime</td>
<td>Adj</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>4</td>
<td>4</td>
<td>4</td>
</tr>
<tr>
<td>5</td>
<td>honda</td>
<td>COMPANY</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>5</td>
<td>5</td>
<td>5</td>
</tr>
<tr>
<td>6</td>
<td>sell</td>
<td>Verb</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>6</td>
<td>6</td>
<td>6</td>
</tr>
<tr>
<td>7</td>
<td>won</td>
<td>Noun</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>7</td>
<td>7</td>
<td>7</td>
</tr>
<tr>
<td>8</td>
<td>toyota tacoma</td>
<td>VEHICLE</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>N</td>
<td>40</td>
<td>39</td>
<td>39</td>
</tr>
<tr>
<td>9</td>
<td>bright</td>
<td>Adj</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>8</td>
<td>8</td>
<td>8</td>
</tr>
<tr>
<td>10</td>
<td>honda</td>
<td>Prop</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>9</td>
<td>9</td>
<td>9</td>
</tr>
<tr>
<td>11</td>
<td>sold</td>
<td>Verb</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>30</td>
<td>30</td>
<td>30</td>
</tr>
<tr>
<td>12</td>
<td>colors</td>
<td>Noun</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>31</td>
<td>31</td>
<td>31</td>
</tr>
<tr>
<td>13</td>
<td>award</td>
<td>Verb</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>10</td>
<td>10</td>
<td>10</td>
</tr>
<tr>
<td>14</td>
<td>except for</td>
<td>Pre</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>11</td>
<td>11</td>
<td>11</td>
</tr>
<tr>
<td>15</td>
<td>color</td>
<td>Noun</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>12</td>
<td>12</td>
<td>12</td>
</tr>
<tr>
<td>16</td>
<td>2008</td>
<td>Num</td>
<td>Num</td>
<td>1</td>
<td>1</td>
<td>N</td>
<td>36</td>
<td>36</td>
<td>36</td>
</tr>
<tr>
<td>17</td>
<td>bright green</td>
<td>NOUN_GROUP</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>13</td>
<td>13</td>
<td>13</td>
</tr>
<tr>
<td>18</td>
<td>the</td>
<td>Det</td>
<td>Alpha</td>
<td>8</td>
<td>1</td>
<td>Y</td>
<td>14</td>
<td>14</td>
<td>14</td>
</tr>
<tr>
<td>19</td>
<td>is</td>
<td>Verb</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>32</td>
<td>22</td>
<td>22</td>
</tr>
<tr>
<td>20</td>
<td>ford taurus</td>
<td>VEHICLE</td>
<td>Entity</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>33</td>
<td>24</td>
<td>24</td>
</tr>
<tr>
<td>21</td>
<td>hyundai</td>
<td>COMPANY</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>15</td>
<td>15</td>
<td>15</td>
</tr>
<tr>
<td>22</td>
<td>toyota</td>
<td>Prop</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>N</td>
<td>37</td>
<td>37</td>
<td>37</td>
</tr>
<tr>
<td>23</td>
<td>car</td>
<td>Noun</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>16</td>
<td>16</td>
<td>16</td>
</tr>
<tr>
<td>24</td>
<td>in</td>
<td>Prep</td>
<td>Alpha</td>
<td>3</td>
<td>1</td>
<td>Y</td>
<td>17</td>
<td>17</td>
<td>17</td>
</tr>
<tr>
<td>25</td>
<td>was</td>
<td>Verb</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>34</td>
<td>22</td>
<td>22</td>
</tr>
<tr>
<td>26</td>
<td>honda insight</td>
<td>VEHICLE</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>35</td>
<td>5</td>
<td>5</td>
</tr>
<tr>
<td>27</td>
<td>insight</td>
<td>Noun</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>18</td>
<td>18</td>
<td>18</td>
</tr>
<tr>
<td>28</td>
<td>ford</td>
<td>Prop</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>19</td>
<td>19</td>
<td>19</td>
</tr>
<tr>
<td>29</td>
<td>world</td>
<td>Prop</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>20</td>
<td>20</td>
<td>20</td>
</tr>
<tr>
<td>30</td>
<td>.</td>
<td>Punct</td>
<td>Punct</td>
<td>5</td>
<td>1</td>
<td>N</td>
<td>38</td>
<td>38</td>
<td>38</td>
</tr>
<tr>
<td>31</td>
<td>year</td>
<td>Noun</td>
<td>Alpha</td>
<td>3</td>
<td>1</td>
<td>Y</td>
<td>21</td>
<td>21</td>
<td>21</td>
</tr>
<tr>
<td>32</td>
<td>be</td>
<td>Verb</td>
<td>Alpha</td>
<td>3</td>
<td>1</td>
<td>Y</td>
<td>22</td>
<td>22</td>
<td>22</td>
</tr>
<tr>
<td>33</td>
<td>all</td>
<td>Adj</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>23</td>
<td>23</td>
<td>23</td>
</tr>
<tr>
<td>34</td>
<td>toyota</td>
<td>COMPANY</td>
<td>Entity</td>
<td>2</td>
<td>1</td>
<td>N</td>
<td>39</td>
<td>39</td>
<td>39</td>
</tr>
<tr>
<td>35</td>
<td>toyota</td>
<td>COMPANY</td>
<td>Entity</td>
<td>1</td>
<td>1</td>
<td>N</td>
<td>39</td>
<td>39</td>
<td>39</td>
</tr>
<tr>
<td>36</td>
<td>ford</td>
<td>COMPANY</td>
<td>Entity</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>24</td>
<td>24</td>
<td>24</td>
</tr>
<tr>
<td>37</td>
<td>taurus</td>
<td>Noun</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>25</td>
<td>25</td>
<td>25</td>
</tr>
<tr>
<td>38</td>
<td>world car</td>
<td>PROP_MISC</td>
<td>Entity</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>26</td>
<td>26</td>
<td>26</td>
</tr>
<tr>
<td>39</td>
<td>of</td>
<td>Prep</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>27</td>
<td>27</td>
<td>27</td>
</tr>
<tr>
<td>40</td>
<td>tacoma</td>
<td>Prop</td>
<td>Alpha</td>
<td>1</td>
<td>1</td>
<td>Y</td>
<td>28</td>
<td>28</td>
<td>28</td>
</tr>
<tr>
<td>41</td>
<td>green</td>
<td>Noun</td>
<td>Alpha</td>
<td>2</td>
<td>1</td>
<td>Y</td>
<td>29</td>
<td>29</td>
<td>29</td>
</tr>
</tbody>
</table>
Example 4.8: Running in Distributed Mode

This example uses the data set that is generated in the section “Getting Started: HPTMINE Procedure” on page 41. When a host for distributed computing is specified and the NODES option in the PERFORMANCE statement is specified as in the following statements, PROC HPTMINE uses the specified host for computing and runs in distributed mode.

```sas
%let GRID_TEXTANALYTICS_BIN_LOC=&GRIDTXTBIN;

option set=GRIDHOST="&GRIDHOST";
option set=GRIDINSTALLLOC="&GRIDINSTALLLOC";
proc hptmine data=getstart;
doc_id did;
variable text;
parse
  outterms = terms
  reducef = 2;
svd
  k = 10
  outdocpro = docpro;
performance nodes=2 details;
run;
```

In the preceding statements, the macro variable GRID_TEXTANALYTICS_BIN_LOC specifies the location on the grid where the language binary files are installed. These files are used by the HPTMINE procedure in parsing text.
### Overview: HPTMSCORE Procedure

For high-performance text mining, scoring is the process of applying the parsing and SVD projections to new textual data. The HPTMSCORE procedure performs this scoring of new documents, and its primary output is the document data set `Outdocpro`, which holds the reduced dimensional representation of the score collection. PROC HPTMSCORE uses some of the output data sets of the HPTMINE procedure as input data to ensure consistency between scoring and training. During scoring, the new textual data must be parsed using the same settings that the training data were parsed with, indexed using only the subset of terms that were used during training, and projected onto the reduced dimensional subspace of the singular value decomposition that was derived from the training data. To facilitate this process, you specify the `CONFIG=`, `TERMS=`, and `SVDU=` options in PROC HPTMINE to create three data sets, `Outconfig`, `Outterms`, and `Svdu`, respectively, and then you specify those three data sets as inputs to PROC HPTMSCORE. For more information about these data sets, see their respective options in the section “PROC HPTMSCORE Statement” on page 84.

PROC HPTMSCORE runs in either single-machine mode or distributed mode.

**NOTE:** Distributed mode requires SAS High-Performance Text Miner.
PROC HPTMSCORE Features

The HPTMSCORE procedure processes large-scale textual data in parallel to achieve efficiency and scalability. The following list summarizes the basic features of PROC HPTMSCORE:

- Functionalities that are related to document parsing, term-by-document matrix creation, and dimension reduction are integrated into one procedure to process data more efficiently.
- Parsing and term-by-document matrix creation are performed in parallel.
- Computation of document projection is performed in parallel.
- Analysis can be performed on a massively parallel SAS high-performance appliance.
- All phases of processing make use of a high degree of multithreading.
- Input data can be read in parallel when the data source is the appliance database.

PROC HPTMSCORE Contrasted with Other SAS Procedures

The high-performance PROC HPTMSCORE functionality replaces what has in the past been done by using the DATA step function TGSCORE followed by a call to the SPSVD procedure. The TGSCORE function was used to parse each document and to append each document’s term counts to an Outparent data set. Then PROC SPSVD applied the document projections. PROC HPTMSCORE combines these tasks into a single parallel procedure. For PROC HPTMSCORE, you provide the necessary inputs to perform the parsing, term-by-document table creation, and document projections in a single call. The primary output data set of PROC HPTMSCORE contains the document projections and is specified in the SVDDOCPRO= option.

Getting Started: HPTMSCORE Procedure

The following DATA steps generate two data sets. The getstart data set contains 36 observations, and the getstart_score data set contains 31 observations. Both data sets have two variables. The text variable contains the input documents, and the did variable contains the ID of the documents. Each row in the data sets represents a “document” for analysis.

```
data getstart;
   infile cards delimiter='|' missover;
   length text $150;
   input text$ did$;
   cards;
   High-performance analytics hold the key to |d01
   unlocking the unprecedented business value of big data.|d02
   Organizations looking for optimal ways to gain insights|d03
   from big data in shorter reporting windows are turning to SAS.|d04
   As the gold-standard leader in business analytics |d05
```
for more than 36 years,
SAS frees enterprises from the limitations of traditional computing and enables them to draw instant benefits from big data.

Faster Time to Insight.
From banking to retail to health care to insurance,
SAS is helping industries glean insights from data that once took days or weeks in just hours, minutes or seconds.
It's all about getting to and analyzing relevant data faster.
Revealing previously unseen patterns, sentiments and relationships.
Identifying unknown risks.
And speeding the time to insights.

High-Performance Analytics from SAS
Combining industry-leading analytics software with high-performance computing technologies produces fast and precise answers to unsolvable problems and enables our customers to gain greater competitive advantage.
SAS In-Memory Analytics eliminate the need for disk-based processing allowing for much faster analysis.
SAS In-Database executes analytic logic into the database itself for improved agility and governance.
SAS Grid Computing creates a centrally managed, shared environment for processing large jobs and supporting a growing number of users efficiently.
Together, the components of this integrated, supercharged platform are changing the decision-making landscape and redefining how the world solves big data business problems.

Big data is a popular term used to describe the exponential growth, availability and use of information, both structured and unstructured.

Much has been written on the big data trend and how it can serve as the basis for innovation, differentiation and growth.

data getstart_score;
  infile cards delimiter='|' missover;
  length text $150;
  input text$ did$;
  cards;
Big data according to SAS
At SAS, we consider two other dimensions when thinking about big data:
Variability. In addition to the increasing velocities and varieties of data, data flows can be highly inconsistent with periodic peaks.
Is something big trending in the social media? Perhaps there is a high-profile IPO looming.
Maybe swimming with pigs in the Bahamas is suddenly the must-do vacation activity. Daily, seasonal and event-triggered peak data loads can be challenging to manage - especially with social media involved.
Complexity. When you deal with huge volumes of data, it comes from multiple sources. It is quite an undertaking to link, match, cleanse and transform data across systems. However,
it is necessary to connect and correlate relationships, hierarchies and multiple data linkages or your data can quickly spiral out of control. Data governance can help you determine how disparate data relates to common definitions and how to systematically integrate structured and unstructured data assets to produce high-quality information that is useful, appropriate and up-to-date. Ultimately, regardless of the factors involved, we believe that the term big data is relative whenever an organization's ability to handle, store and analyze data exceeds its current capacity.

run;

The following statements use PROC HPTMINE for processing the input text data set \texttt{getstart} and create three data sets, \texttt{outconfig}, \texttt{outterms}, and \texttt{svdu}, which can be used in PROC HPTMSCORE for scoring. The statements then use PROC HPTMSCORE to score the input text data set \texttt{getstart_score}. The statements take the three data sets that are generated by PROC HPTMINE as input and create a SAS data set named \texttt{docpro}, which contains the projection of the documents in the input data set \texttt{getstart_score}.

```
proc hptmine data = getstart;
  doc_id    did;
  variable  text;
  parse
    outterms = outterms
    outconfig = outconfig
    reducef  = 2;
  svd
    k        = 10
    svdu     = svdu;
  performance details;
run;
```

```
proc hptmscore
  data      = getstart_score
  terms     = outterms
  config    = outconfig
  svdu      = svdu
  svddocpro = docpro;
  doc_id    did;
  variable  text;
  performance details;
run;
```

The output from this analysis is presented in Figure 5.1 through Figure 5.3.

Figure 5.1 shows the “Performance Information” table, which indicates that PROC HPTMSCORE executes in single-machine mode. That is, PROC HPTMSCORE runs on the machine where the SAS system is running. The table also shows that two threads are used for computing.
Figure 5.1 Performance Information

Performance Information

<table>
<thead>
<tr>
<th>Execution Mode</th>
<th>Single-Machine</th>
</tr>
</thead>
<tbody>
<tr>
<td>Number of Threads</td>
<td>2</td>
</tr>
</tbody>
</table>

Figure 5.2 shows the “Procedure Task Timing” table, which provides details about how much time is used by each processing step.

Figure 5.2 Procedure Task Timing

<table>
<thead>
<tr>
<th>Task</th>
<th>Time (sec.)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Parse documents</td>
<td>0.71</td>
</tr>
<tr>
<td>Generate term-document matrix</td>
<td>0.01</td>
</tr>
<tr>
<td>Compute SVD</td>
<td>0.01</td>
</tr>
<tr>
<td>Output OUTDOCPRO table</td>
<td>0.00</td>
</tr>
</tbody>
</table>

The following statements use PROC PRINT to show the content of the first 10 rows of the docpro data set that is generated by the HPTMSCORE procedure.

```
proc print data = docpro (obs=10) round; run;
```

Figure 5.3 shows the output of PROC PRINT. For more information about the output of the OUTDOCPRO= option, see the description of the SVDDOCPRO= option.

Figure 5.3 The DOCPRO Data Set

<table>
<thead>
<tr>
<th>Obs</th>
<th>did</th>
<th>COL1</th>
<th>COL2</th>
<th>COL3</th>
<th>COL4</th>
<th>COL5</th>
<th>COL6</th>
<th>COL7</th>
<th>COL8</th>
<th>COL9</th>
<th>COL10</th>
</tr>
</thead>
<tbody>
<tr>
<td>1</td>
<td>d1</td>
<td>0.90</td>
<td>0.09</td>
<td>-0.01</td>
<td>0.37</td>
<td>0.06</td>
<td>-0.13</td>
<td>-0.14</td>
<td>-0.03</td>
<td>-0.03</td>
<td>-0.04</td>
</tr>
<tr>
<td>2</td>
<td>d2</td>
<td>0.29</td>
<td>-0.84</td>
<td>0.14</td>
<td>0.36</td>
<td>0.12</td>
<td>0.10</td>
<td>0.11</td>
<td>-0.10</td>
<td>-0.09</td>
<td>0.08</td>
</tr>
<tr>
<td>3</td>
<td>d3</td>
<td>0.82</td>
<td>0.36</td>
<td>0.20</td>
<td>0.27</td>
<td>0.08</td>
<td>-0.17</td>
<td>-0.06</td>
<td>-0.12</td>
<td>0.05</td>
<td>-0.18</td>
</tr>
<tr>
<td>4</td>
<td>d4</td>
<td>0.50</td>
<td>-0.38</td>
<td>0.13</td>
<td>-0.21</td>
<td>0.00</td>
<td>-0.27</td>
<td>-0.36</td>
<td>0.47</td>
<td>0.14</td>
<td>0.32</td>
</tr>
<tr>
<td>5</td>
<td>d5</td>
<td>0.69</td>
<td>-0.12</td>
<td>0.11</td>
<td>-0.01</td>
<td>0.12</td>
<td>-0.04</td>
<td>0.21</td>
<td>-0.34</td>
<td>-0.54</td>
<td>0.19</td>
</tr>
<tr>
<td>6</td>
<td>d6</td>
<td>0.27</td>
<td>-0.62</td>
<td>0.01</td>
<td>0.01</td>
<td>0.32</td>
<td>-0.28</td>
<td>0.54</td>
<td>-0.10</td>
<td>0.07</td>
<td>-0.24</td>
</tr>
<tr>
<td>7</td>
<td>d7</td>
<td>0.67</td>
<td>-0.48</td>
<td>0.26</td>
<td>-0.02</td>
<td>0.14</td>
<td>-0.37</td>
<td>0.21</td>
<td>0.14</td>
<td>0.16</td>
<td>-0.04</td>
</tr>
<tr>
<td>8</td>
<td>d8</td>
<td>0.35</td>
<td>-0.10</td>
<td>-0.26</td>
<td>0.16</td>
<td>-0.21</td>
<td>-0.49</td>
<td>0.64</td>
<td>-0.25</td>
<td>-0.11</td>
<td>0.00</td>
</tr>
<tr>
<td>9</td>
<td>d9</td>
<td>0.49</td>
<td>-0.64</td>
<td>0.18</td>
<td>-0.07</td>
<td>0.13</td>
<td>-0.39</td>
<td>0.26</td>
<td>0.19</td>
<td>0.19</td>
<td>0.04</td>
</tr>
<tr>
<td>10</td>
<td>d10</td>
<td>0.36</td>
<td>-0.68</td>
<td>0.20</td>
<td>-0.32</td>
<td>0.27</td>
<td>0.05</td>
<td>0.33</td>
<td>-0.24</td>
<td>-0.06</td>
<td>-0.15</td>
</tr>
</tbody>
</table>
**Syntax: HPTMSCORE Procedure**

The following statements are available in the HPTMSCORE procedure:

```
PROC HPTMSCORE <options> ;
  VARIABLE variable ;
  DOC_ID variable ;
  PERFORMANCE performance-options ;
```

**PROC HPTMSCORE Statement**

```
PROC HPTMSCORE <options> ;
```

The PROC HPTMSCORE statement invokes the procedure. Table 5.1 summarizes the options in the statement by function. The options are then described fully in alphabetical order.

<table>
<thead>
<tr>
<th>option</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Basic Options</strong></td>
<td></td>
</tr>
<tr>
<td>DATA</td>
<td>DOC=</td>
</tr>
<tr>
<td>TERMS=</td>
<td>Specifies the data set that contains the summary information about the terms that are to be used for scoring</td>
</tr>
<tr>
<td>CONFIG=</td>
<td>Specifies the data set that contains the option settings that are to be used for scoring</td>
</tr>
<tr>
<td>SVDU=</td>
<td>Specifies the data set that contains the U matrix whose columns are the left singular vectors</td>
</tr>
<tr>
<td><strong>Output Options</strong></td>
<td></td>
</tr>
<tr>
<td>NOPRINT</td>
<td>Suppresses ODS output</td>
</tr>
<tr>
<td>OUTPARENT=</td>
<td>Specifies the term-by-document frequency matrix that is used to model the document collection, in which the child terms are not represented and child terms’ frequencies are attributed to their corresponding parents</td>
</tr>
<tr>
<td>SVDDOCPRO=</td>
<td>Specifies the projections of the documents</td>
</tr>
</tbody>
</table>

You can specify the following options in the PROC HPTMSCORE statement.

**CONFIG=SAS-data-set**

specifies the input SAS data set that contains configuration information for PROC HPTMSCORE. The data set that is used for this option should be the one that is generated by the OUTCONFIG option in the PARSE statement of the HPTMINE procedure during training. For more information about this data set, see the section “The OUTCONFIG= Data Set” on page 57 of Chapter 4, “The HPTMINE Procedure.”
**DATA | DOC=SAS-data-set**
specifies the input SAS data set of documents to be used by PROC HPTMSCORE. The default is the most recently created data set. If PROC HPTMSCORE executes in distributed mode, the input data are distributed to memory on the appliance nodes and analyzed in parallel, unless the data are already distributed in the appliance database. When data are already distributed, PROC HPTMSCORE reads the data alongside the distributed database. For more information, see the sections “Processing Modes” on page 8 and “Alongside-the-Database Execution” on page 15. Each row of the input data set must contain one text field and one ID field, which correspond to the text and the unique ID of a document, respectively.

**NOPRINT**
suppresses the generation of ODS output.

**OUTPARENT=SAS-data-set**
specifies the output data set to contain a compressed representation of the sparse term-by-document frequency matrix. The data set contains only the kept, representative terms, and the child frequencies are attributed to the corresponding parent. For more information about the compressed representation of the sparse term-by-document frequency matrix, see the section “The OUTPARENT= Data Set” on page 58 of Chapter 4, “The HPTMINE Procedure.”

**SVDDOCPRO=SAS-data-set <KEEPVARS | KEEPVARIABLES=variable-list>**
specifies the output data set to contain the reduced dimensional projections for each document. The contents of this data set are formed by multiplying the term-by-document frequency matrix by the input data set that is specified in the SVDU= option and then normalizing the result.

The variables in the data set that is specified by the DATA= option can be copied to the output of this option. When KEEPVARS | KEEPVARIABLES=variable-list is used, the content of the variables that is specified in the variable-list is attached to the output. These variables must appear in the data set that is specified by the DATA= option.

**SVDU=SAS-data-set**
specifies the input data set that contains the U matrix, which is created during training by PROC HPTMINE. The data set contains the information that is needed to project each document into the reduced dimensional space. For more information about the contents of this data set, see the SVDU= option in Chapter 4, “The HPTMINE Procedure.”

**TERMS=SAS-data-set**
specifies the input SAS data set of terms to be used by PROC HPTMSCORE. The data set that is used for this option should be the one that is generated by the OUTTERMS= option in the PARSE statement of the HPTMINE procedure during training. This data set conveys to PROC HPTMSCORE which terms should be used in the analysis and whether they should be mapped to a parent. The data set also assigns them a key that corresponds to the key that is used in the input data set that is specified by the SVDU= option. For more information about this data set, see the section “The OUTTERMS= Data Set” on page 59 of Chapter 4, “The HPTMINE Procedure.”
Chapter 5: The HPTMSCORE Procedure

**DOC_ID Statement**

```
DOC_ID variable < docid-options > ;
```

This statement specifies the variable that contains the ID of each document. In the input data set, each row corresponds to one document. The ID of each document must be unique; it can be either a number or a string of characters.

You can specify the following `docid-option` in the DOC_ID statement:

- **DUPLICATIONCHECK | DUPCHK < ( WARNING | STOP ) >**
  
  checks whether the ID of each document is unique. When this option is not specified, no duplication check is performed. When you specify this option along with the WARNING keyword and duplicate document IDs are detected, the HPTMSCORE procedure outputs a warning message, which shows the number of duplicate document IDs that have been detected. When you specify this option along with the STOP keyword, the HPTMSCORE procedure outputs an error message and terminates. The error message shows the number of duplicate document IDs that have been detected. If you specify the DUPLICATIONCHECK option without the WARNING or STOP keyword, the HPTMSCORE procedure takes the WARNING keyword by default.

**PERFORMANCE Statement**

```
PERFORMANCE < performance-options > ;
```

The PERFORMANCE statement defines performance parameters for multithreaded and distributed computing, passes variables that describe the distributed computing environment, and requests detailed results about the performance characteristics of the HPTMSCORE procedure.

You can also use the PERFORMANCE statement to control whether the HPTMSCORE procedure executes in single-machine or distributed mode.

The PERFORMANCE statement for is documented further in the section “PERFORMANCE Statement” on page 35 of Chapter 3, “Shared Concepts and Topics.”

**VARIABLE Statement**

```
VARIABLE | VAR variable ;
```

This statement specifies the variable that contains the text to be processed.
Details: HPTMSCORE Procedure

For information about the techniques that are used for nature language processing, term processing, and singular value decomposition, see the section “Details: HPTMINE Procedure” on page 52 of Chapter 4, “The HPTMINE Procedure.”

Displayed Output

The following sections describe the output that PROC HPTMSCORE produces. The output is organized into various tables, which are discussed in their order of appearance.

Performance Information

The “Performance Information” table is produced by default. It displays information about the execution mode. For single-machine mode, the table displays the number of threads used. For distributed mode, the table displays the grid mode (symmetric or asymmetric), the number of compute nodes, and the number of threads per node. If you specify the DETAILS option in the PERFORMANCE statement, the procedure also produces a “Timing” table in which elapsed times (absolute and relative) for the main tasks of the procedure are displayed.

Procedure Task Timing

When the DETAILS option is specified in the PERFORMANCE statement, the procedure produces a “Procedure Task Timing” table, which displays the elapsed time (absolute and relative) for the main tasks.

ODS Table Names

Each table that is created by the HPTMSCORE procedure has a name associated with it, and you must use this name to refer to the table when you use ODS statements. These names are listed in Table 5.2.

<table>
<thead>
<tr>
<th>Table Name</th>
<th>Description</th>
<th>Required Statement / Option</th>
</tr>
</thead>
<tbody>
<tr>
<td>PerformanceInfo</td>
<td>Information about the high-performance computing environment</td>
<td>Default output</td>
</tr>
<tr>
<td>Timing</td>
<td>Absolute and relative times for tasks that are performed by the procedure</td>
<td>DETAILS option in the PERFORMANCE statement</td>
</tr>
</tbody>
</table>
Chapter 5: The HPTMSCORE Procedure

System Configuration

Prerequisites for Running PROC HPTMSCORE

To use the HPTMSCORE procedure, you must have a valid SAS Text Miner license, and the language binary files that are provided under that license must be available to PROC HPTMSCORE for parsing text.

Input Data Accessibility

When you run PROC HPTMSCORE in distributed mode, the following input data sets need to be accessible to the client machine:

- TERMS= data set
- CONFIG= data set
- SVDU= data set

When the HPTMSCORE procedure runs in distributed mode, it first reads these data sets on the client machine and then sends them to the grid.

Configuration for Distributed Mode

When PROC HPTMSCORE runs in distributed mode, it needs to locate on the grid the language binary files that are used in parsing text. These binary files must be deployed to the grid. In addition, the GRID_TEXTANALYTICS_BIN_LOC macro must be specified to indicate to location of the binary files on your grid. For more information, see the section “Configuring for Distributed Mode” on page 62 of Chapter 4, “The HPTMINE Procedure.”
Subject Index

displayed output
  HPTMINE procedure, 61
  HPTMSCORE procedure, 87

HPTMINE procedure, 40
  cell weight, 48
  displayed output, 61
  duplicate document ID checking, 47
  entity, 48
  filtering term by frequency, 49
  input data sets, 46
  multiterm words list, 48
  multithreading, 50
  noun groups, 49
  ODS table names, 61
  performance information, 61
  procedure task timing, 61
  sparse matrix, 58
  start list, 49
  stemming, 49
  stop list, 50
  SVD, singular value decomposition, 55
  synonym list, 50
  system configuration, 61
  tagging, 49
  term weight, 50
  transactional style, 58

HPTMINE procedure, system configuration
  accessing input data, 62
  configuration for distributed mode, 62
  deploying language binary files, 62
  GRID_TEXTANALYTICS_BIN_LOC macro, 63
  language binary files, 63
  prerequisite, 61

HPTMSCORE procedure, 79
  displayed output, 87
  duplicate document ID checking, 86
  input data sets, 85
  multithreading, 86
  ODS table names, 87
  performance information, 87
  procedure task timing, 87
  system configuration, 88

HPTMSCORE procedure, system configuration
  accessibility of the input data, 88
  configuration for distributed mode, 88
  prerequisite, 88

multithreading

options summary
  PARSE statement, 47
  PROC HPTMINE statement, 46
  PROC HPTMSCORE statement, 84
  SVD statement, 51

performance information
  HPTMINE procedure, 61
  HPTMSCORE procedure, 87

procedure task timing
  HPTMINE procedure, 61
  HPTMSCORE procedure, 87

sparse matrix
  HPTMINE procedure, 58

transactional style
  HPTMINE procedure, 58
Syntax Index

CELLWGT= option
   PARSE statement, 48
COMMIT= option
   PERFORMANCE statement (high-performance analytical procedures), 35
CONFIG= option
   HPTMSCORE statement, 84
DATA= option
   PROC HPTMINE statement, 46
   PROC HPTMSCORE statement, 85
DATASERVER= option
   PERFORMANCE statement (high-performance analytical procedures), 35
DETAILS option
   PERFORMANCE statement (high-performance analytical procedures), 36
DOC= option
   PROC HPTMINE statement, 46
   PROC HPTMSCORE statement, 85
DOC_ID statement
   HPTMINE procedure, 47
   HPTMSCORE procedure, 86
DUPCHK option
   DOC_ID statement, 47, 86
DUPLICATIONCHECK option
   DOC_ID statement, 47, 86
ENTITIES= option
   PARSE statement, 48
GRIDHOST= option
   PERFORMANCE statement (high-performance analytical procedures), 36
GRIDMODE= option
   PERFORMANCE statement (high-performance analytical procedures), 36
GRIDTIMEOUT= option
   PERFORMANCE statement (high-performance analytical procedures), 36

high-performance analytical procedures,
   PERFORMANCE statement, 35
COMMIT= option, 35
DATASERVER= option, 35
DETAILS option, 36
GRIDHOST= option, 36
GRIDMODE= option, 36
GRIDTIMEOUT= option, 36

HOST= option, 36
INSTALL= option, 36
INSTALLLOC= option, 36
LASR= option, 36
LASRSERVER= option, 36
MODE= option, 36
NNODES= option, 37
NODES= option, 37
NTHREADS= option, 38
THREADS= option, 38
TIMEOUT= option, 36

HOST= option
   PERFORMANCE statement (high-performance analytical procedures), 36
HPTM procedure
   PERFORMANCE statement, 86
HPTMINE procedure, 46
   PARSE statement, 47
   PERFORMANCE statement, 50
   PROC HPTMINE statement, 46
   SVD statement, 50
   syntax, 46
HPTMINE procedure, DOC_ID statement, 47
   DUPCHK option, 47
   DUPLICATIONCHECK option, 47
HPTMINE procedure, PARSE statement, 47
   CELLWGT= option, 48
   ENTITIES= option, 48
   MULTITERM= option, 48
   NONG option, 49
   NONOUNGROUPS option, 49
   NOSTEMMING option, 49
   NOTAGGING option, 49
   OUTCHILD= option, 49
   OUTCONFIG= option, 49
   OUTPARENT= option, 49
   OUTTERMS= option, 49
   REDUCEF= option, 49
   START= option, 49
   STOP= option, 50
   SYNONYM= option, 50
   TERMWGT= option, 50
HPTMINE procedure, PERFORMANCE statement, 50
HPTMINE procedure, PROC HPTMINE statement, 46
   DATA= option, 46
   DOC= option, 46
   NOPRINT option, 46
HPTMINE procedure, SVD statement, 50
K= option, 51
KEEPVARS, KEEPVARIABLES, 51
MAX_K= option, 51
OUTDOCPRO= option, 51
RES= option, 52
RESOLUTION= option, 52
SVDS= option, 52
SVDU= option, 52
SVDV= option, 52
TOL= option, 52

HPTMINE procedure, TARGET statement, 52
HPTMINE procedure, VARIABLE statement, 52
HPTMSCORE procedure, 84
PROC HPTMSCORE statement, 84
syntax, 84
HPTMSCORE procedure, DOC_ID statement, 86
DUPCHK option, 86
DUPLICATIONCHECK option, 86
HPTMSCORE procedure, HPTMSCORE statement
CONFIG= option, 84
HPTMSCORE procedure, PERFORMANCE statement, 86
HPTMSCORE procedure, PROC HPTMSCORE statement, 84
DATA= option, 85
DOC= option, 85
KEEPVARS, KEEPVARIABLES, 85
NOPRINT option, 85
OUTPARENT= option, 85
SVDDOCPRO= option, 85
SVDU= option, 85
TERMS= option, 85
HPTMSCORE procedure, VAR statement, 86
HPTMSCORE procedure, VARIABLE statement, 86
INSTALL= option
PERFORMANCE statement (high-performance analytical procedures), 36
INSTALLLOC= option
PERFORMANCE statement (high-performance analytical procedures), 36

K= option
SVD statement, 51
KEEPVARS, KEEPVARIABLES
PROC HPTMSCORE statement, 85
SVD statement, 51

LASR= option
PERFORMANCE statement (high-performance analytical procedures), 36
LASRSERVER= option
PERFORMANCE statement (high-performance analytical procedures), 36

MAX_K= option
SVD statement, 51
MODE= option
PERFORMANCE statement (high-performance analytical procedures), 36
MULTITERM= option
PARSE statement, 48

NNODES= option
PERFORMANCE statement (high-performance analytical procedures), 37
NODES= option
PERFORMANCE statement (high-performance analytical procedures), 37
NONG option
PARSE statement, 49
NONOUNGROUPS option
PARSE statement, 49
NOPRINT option
PROC HPTMINE statement, 46
PROC HPTMSCORE statement, 85

NOSTEMMING option
PARSE statement, 49
NOTAGGING option
PARSE statement, 49
NTHREADS= option
PERFORMANCE statement (high-performance analytical procedures), 38

OUTCHILD= option
PARSE statement, 49
OUTCONFIG= option
PARSE statement, 49
OUTDOCPRO= option
SVD statement, 51
OUTPARENT= option
PROC HPTMSCORE statement, 85
OUTTERMS= option
PARSE statement, 49

PARSE statement
HPTMINE procedure, 47
PERFORMANCE statement
high-performance analytical procedures, 35
HPTM procedure, 86
HPTMINE procedure, 50
PROC HPTMINE statement
HPTMINE procedure, 46
PROC HPTMSCORE statement
HPTMSCORE procedure, 84

REDUCEF= option
PARSE statement, 49
RES= option
syntax

SVD statement, 52
RESOLUTION= option
   SVD statement, 52

START= option
   PARSE statement, 49
STOP= option
   PARSE statement, 50

SVDS= option
   SVD statement, 52
SVDP= option
   SVD statement, 52
SVDV= option
   SVD statement, 52
SYNONYM= option
   PARSE statement, 50

TARGET statement
   HPTMINE procedure, 52
TERMS= option
   PROC HPTMSCORE statement, 85
TERMWGT= option
   PARSE statement, 50

THREADS= option
   PERFORMANCE statement (high-performance analytical procedures), 38

TIMEOUT= option
   PERFORMANCE statement (high-performance analytical procedures), 36

TOL= option
   SVD statement, 52

VAR statement
   HPTMINE procedure, 52
   HPTMSCORE procedure, 86

VARIABLE statement
   HPTMINE procedure, 52
   HPTMSCORE procedure, 86