SAS/STAT® 12.3 User's Guide
The PROBIT Procedure
(Chapter)

Chapter 75
The PROBIT Procedure

Contents

Overview: PROBIT Procedure

The PROBIT procedure calculates maximum likelihood estimates of regression parameters and the natural (or threshold) response rate for quantal response data from biological assays or other discrete event data. This includes probit, logit, ordinal logistic, and extreme value (or gompit) regression models.

Probit analysis developed from the need to analyze qualitative (dichotomous or polytomous) dependent variables within the regression framework. Many response variables are binary by nature (yes/no), while others are measured ordinally rather than continuously (degree of severity). Researchers have shown ordinary least squares (OLS) regression to be inadequate when the dependent variable is discrete (Collett 2003; Agresti 2002). Probit or logit analyses are more appropriate in this case.

The PROBIT procedure computes maximum likelihood estimates of the parameters β and C of the probit equation by using a modified Newton-Raphson algorithm. When the response Y is binary, with values 0 and 1, the probit equation is

$$p = \Pr(Y = 0) = C + (1 - C)F(\mathbf{x}'\boldsymbol{\beta})$$

where

$\boldsymbol{\beta}$ is a vector of parameter estimates

F is a cumulative distribution function (normal, logistic, or extreme value)

$\mathbf{x}$ is a vector of explanatory variables

p is the probability of a response

C is the natural (threshold) response rate

Notice that PROC PROBIT, by default, models the probability of the *lower* response levels. The choice of the distribution function F (normal for the probit model, logistic for the logit model, and extreme value or Gompertz for the gompit model) determines the type of analysis. For most problems, there is relatively little difference between the normal and logistic specifications of the model. Both distributions are symmetric about the value zero. The extreme value (or Gompertz) distribution, however, is not symmetric, approaching 0 on the left more slowly than it approaches 1 on the right. You can use the extreme value distribution where such asymmetry is appropriate.

For ordinal response models, the response, Y, of an individual or an experimental unit can be restricted to one of a (usually small) number, $k+1(k \geq 1)$, of ordinal values, denoted for convenience by $1, \ldots, k, k+1$. For example, the severity of coronary disease can be classified into three response categories as 1=no disease, 2=angina pectoris, and 3=myocardial infraction. The PROBIT procedure fits a common slopes cumulative model, which is a parallel-lines regression model based on the cumulative probabilities of the response categories rather than on their individual probabilities. The cumulative model has the form

$$\Pr(Y \leq 1 \mid \mathbf{x}) = F(\mathbf{x}'\boldsymbol{\beta})$$

$$\Pr(Y \leq i \mid \mathbf{x}) = F(\alpha_i + \mathbf{x}'\boldsymbol{\beta}), \quad 2 \leq i \leq k$$

where $\alpha_2, \ldots, \alpha_k$ are $k - 1$ intercept parameters. By default, the covariate vector $\mathbf{x}$ contains an overall intercept term.

You can set or estimate the natural (threshold) response rate C. Estimation of C can begin either from an initial value that you specify or from the rate observed in a control group. By default, the natural response rate is fixed at zero.

An observation in the data set analyzed by the PROBIT procedure might contain the response and explanatory values for one subject. Alternatively, it might provide the number of observed events from a number of subjects at a particular setting of the explanatory variables. In this case, PROC PROBIT models the probability of an event.

The PROBIT procedure uses ODS Graphics to create graphs as part of its output. For general information about ODS Graphics, see Chapter 21, "Statistical Graphics Using ODS." For specific information about the graphics available in the PROBIT procedure, see the section "ODS Graphics" on page 6415.

Getting Started: PROBIT Procedure

The following example illustrates how you can use the PROBIT procedure to compute the threshold response rate and regression parameter estimates for quantal response data.

Estimating the Natural Response Threshold Parameter

Suppose you want to test the effect of a drug at 12 dosage levels. You randomly divide 180 subjects into 12 groups of 15—one group for each dosage level. You then conduct the experiment and, for each subject, record the presence or absence of a positive response to the drug. You summarize the data by counting the number of subjects responding positively in each dose group. Your data set is as follows:

```
data study;
   input Dose Respond @@;
   Number = 15;
   datalines;
0    3   1.1   4   1.3   4   2.0   3   2.2   5   2.8   4
3.7  5   3.9   9   4.4   8   4.8   11  5.9   12  6.8   13
;
```

The variable dose represents the amount of drug administered. The first group, receiving a dose level of 0, is the control group. The variable number represents the number of subjects in each group. All groups are equal in size; hence, number has the value 15 for all observations. The variable respond represents the number of subjects responding to the associated drug dosage.

You can model the probability of positive response as a function of dosage by using the following statements:

```
ods graphics on;

proc probit data=study log10 optc plots=(predpplot ippplot);
   model respond/number=dose;
   output out=new p=p_hat;
run;
```

The DATA= option specifies that PROC PROBIT analyze the SAS data set study. The LOG10 option replaces the first continuous independent variable (dose) with its common logarithm. The OPTC option estimates the natural response rate. When you use the LOG10 option with the OPTC option, any observations with a dose value less than or equal to zero are used in the estimation as a control group.

The PLOTS= option in the PROC PROBIT statement, together with the ODS GRAPHICS statement, requests two plots for the estimated probability values and dosage levels. For general information about ODS Graphics, see Chapter 21, "Statistical Graphics Using ODS." For specific information about the graphics available in the PROBIT procedure, see the section "ODS Graphics" on page 6415.

The MODEL statement specifies a proportional response by using the variables respond and number in *events/trials* syntax. The variable dose is the stimulus or explanatory variable.

The OUTPUT statement creates a new data set, new, that contains all the variables in the original data set, and a new variable, p_hat, that represents the predicted probabilities.

The results from this analysis are displayed in the following figures.

Figure 75.1 displays background information about the model fit. Included are the name of the input data set, the response variables used, and the number of observations, events, and trials. The last line in Figure 75.1 shows the final value of the log-likelihood function.

Figure 75.2 displays the table of parameter estimates for the model. The parameter C, which is the natural response threshold or the proportion of individuals responding at zero dose, is estimated to be 0.2409. Since both the intercept and the slope coefficient have significant p-values (0.0020, 0.0010), you can write the model for

$$\Pr(\text{response}) = C + (1 - C)F(\mathbf{x}'\beta)$$

as

$$\Pr(\text{response}) = 0.2409 + 0.7591(\Phi(-4.1439 + 6.2308 \times \log_{10}(\text{dose})))$$

where Φ is the normal cumulative distribution function.

Finally, PROC PROBIT specifies the resulting tolerance distribution by providing the mean MU and scale parameter SIGMA as well as the covariance matrix of the distribution parameters in Figure 75.3.

Figure 75.1 Model Fitting Information for the PROBIT Procedure

```
                        The Probit Procedure

                        Model Information

        Data Set                              WORK.STUDY
        Events Variable                          Respond
        Trials Variable                           Number
        Number of Observations                        12
        Number of Events                              81
        Number of Trials                             180
        Number of Events In Control Group              3
        Number of Trials In Control Group             15
        Name of Distribution                      Normal
        Log Likelihood                      -104.3945783
```

Figure 75.2 Model Parameter Estimates for the PROBIT Procedure

```
           Analysis of Maximum Likelihood Parameter Estimates


                          Standard    95% Confidence      Chi-
    Parameter    DF Estimate   Error       Limits      Square Pr > ChiSq

    Intercept     1  -4.1438   1.3415  -6.7731  -1.5146   9.54     0.0020
    Log10(Dose)   1   6.2308   1.8996   2.5076   9.9539  10.76     0.0010
    _C_           1   0.2409   0.0523   0.1385   0.3433
```

Figure 75.3 Tolerance Distribution Estimates for the PROBIT Procedure

```
          Estimated Covariance Matrix for Tolerance Parameters

                        MU              SIGMA                _C_

    MU               0.001158        -0.000493           0.000954
    SIGMA           -0.000493         0.002394          -0.000999
    _C_              0.000954        -0.000999           0.002731
```

The PLOT=PREDPPLOT option creates the plot in Figure 75.4, showing the relationship between dosage level, observed response proportions, and estimated probability values. The dashed lines represent pointwise confidence bands for the fitted probabilities, and a reference line is plotted at the estimated threshold value of 0.24.

The PLOT=IPPPLOT option creates the plot in Figure 75.5, showing the inverse relationship between dosage level and observed response proportions/estimated probability values. The dashed lines represent pointwise fiducial limits for the predicted values of the dose variable, and a reference line is also plotted at the estimated threshold value of 0.24.

The two plot options can be put together with the PLOTS= option, as shown in the PROC PROBIT statement.

Figure 75.4 Plot of Observed and Fitted Probabilities versus Dose Level
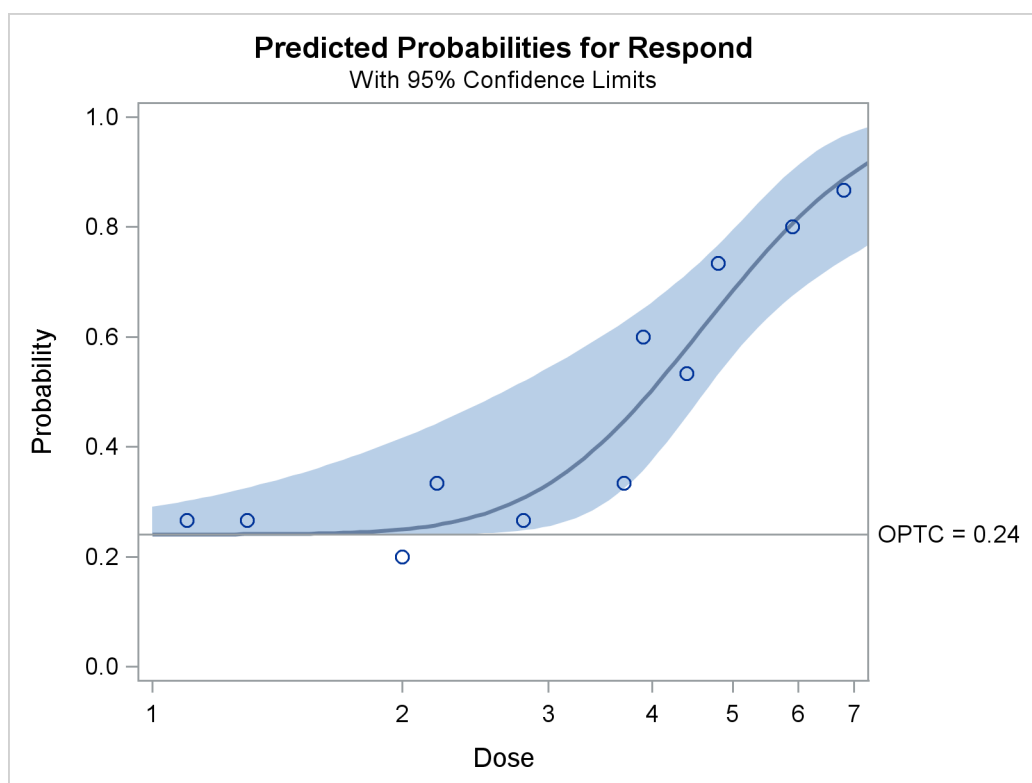
Figure 75.5 Inverse Predicted Probability Plot with Fiducial Limits
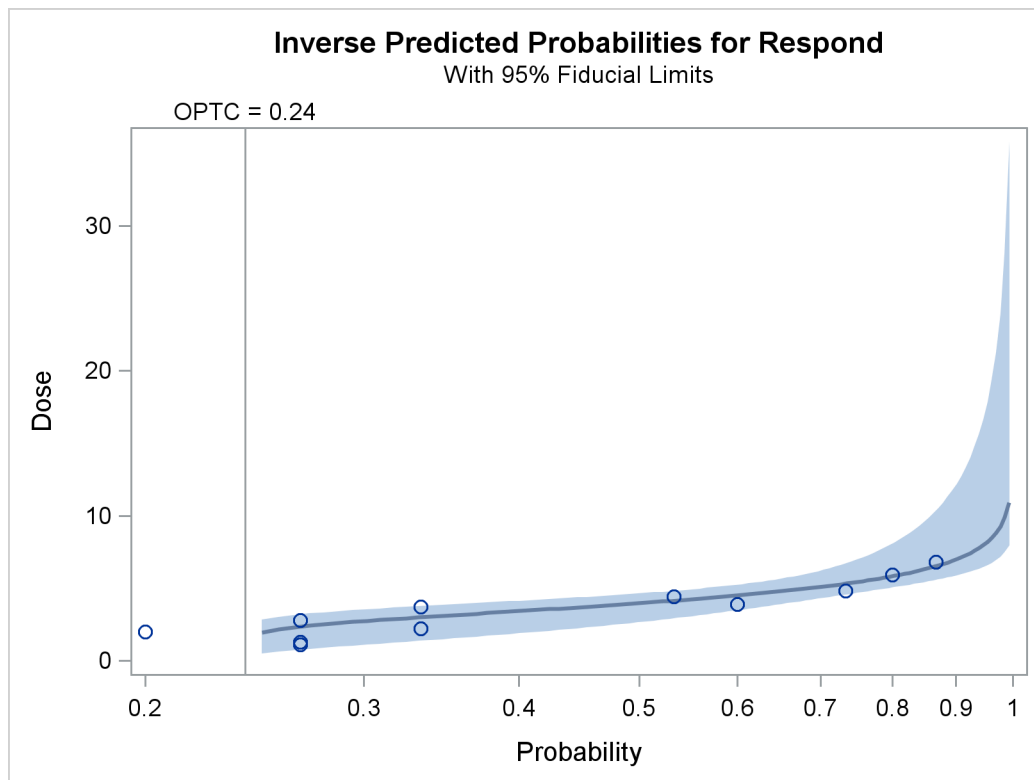
Syntax: PROBIT Procedure

The following statements are available in the PROBIT procedure:

PROC PROBIT *< options >* ;
    BY *variables* ;
    CDFPLOT *< **VAR=**variable > < options >* ;
    CLASS *variables* ;
    ESTIMATE *< 'label' > estimate-specification < (**divisor=**n) >*
              *< , . . . < 'label' > estimate-specification < (**divisor=**n) > > < / options >* ;
    EFFECTPLOT *< plot-type < (plot-definition-options) > > < / options >* ;
    INSET *< keyword-list > < / options >* ;
    IPPPLOT *< **VAR=**variable > < options >* ;
    LPREDPLOT *< **VAR=**variable > < options >* ;
    LSMEANS *< model-effects > < / options >* ;
    LSMESTIMATE *model-effect < 'label' > values < (**divisor=**n) >*
              *< , . . . < 'label' > values < (**divisor=**n) > > < / options >* ;
    MODEL *response = independents < / options >* ;
    OUTPUT *< **OUT=**SAS-data-set > < options >* ;
    PREDPPLOT *< **VAR=**variable > < options >* ;
    SLICE *model-effect < / options >* ;
    STORE *< **OUT=** >item-store-name < / **LABEL=**'label' >* ;
    TEST *< model-effects > < / options >* ;
    WEIGHT *variable* ;

A MODEL statement is required. Only a single MODEL statement can be used with one invocation of the PROBIT procedure. If multiple MODEL statements are present, only the last one is used. Main effects and higher-order terms can be specified in the MODEL statement, as in the GLM procedure. If a CLASS statement is used, it must precede the MODEL statement.

The CDFPLOT, INSET, IPPPLOT, LPREDPLOT, and PREDPPLOT statements are used to produce graphical output. You can use any appropriate combination of the graphical statements after the MODEL statement.

The ESTIMATE, EFFECTPLOT, LSMEANS, LSMESTIMATE, SLICE, STORE, and TEST statements are common to many procedures. Summary descriptions of functionality and syntax for these statements are also given after the PROC PROBIT statement in alphabetical order, and full documentation about them is available in Chapter 19, "Shared Concepts and Topics."

PROC PROBIT Statement

    PROC PROBIT *< options >* ;

The PROC PROBIT statement invokes the PROBIT procedure. Table 75.1 summarizes the options available in the PROC PROBIT statement.

Table 75.1 PROC PROBIT Statement Options

Option	Description
COVOUT	Writes the parameter estimate covariance matrix to the OUTEST= data set
C=	Controls how the natural response is handled
DATA=	Specifies the SAS data set to be used
GOUT=	Specifies a graphics catalog in which to save graphics output
HPROB=	Specifies a minimum probability level for the Pearson's chi-square
INEST=	Specifies an input SAS data set that contains initial estimates
INVERSECL	Computes confidence limits
LACKFIT	Performs two goodness-of-fit tests
LOG	Replaces the first continuous independent variable with its natural logarithm
LOG10	Replaces the first continuous independent variable with log to the base 10
NAMELEN=	Specifies the length of effect names to be *n* characters
NOPRINT	Suppresses the display of all output including graphics
OPTC	Controls how the natural response is handled
ORDER=	Specifies the sort order for the levels of the classification variables
OUTEST=	Specifies a SAS data set to contain the parameter estimates
PLOT | PLOTS	Controls the plots produced though ODS Graphics
XDATA=	Specifies an input SAS data set that contains values for all the independent variables

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

COVOUT

writes the parameter estimate covariance matrix to the OUTEST= data set.

C=rate

OPTC

controls how the natural response is handled. Specify the OPTC option to request that the natural response rate C be estimated. Specify the C=*rate* option to set the natural response rate or to provide the initial estimate of the natural response rate. The natural response rate value must be a number between 0 and 1.

- If you specify neither the OPTC nor the C= option, a natural response rate of zero is assumed.

- If you specify both the OPTC and the C= option, the C= option should be a reasonable initial estimate of the natural response rate. For example, you could use the ratio of the number of responses to the number of subjects in a control group.

- If you specify the C= option but not the OPTC option, the natural response rate is set to the specified value and not estimated.

- If you specify the OPTC option but not the C= option, PROC PROBIT's action depends on the response variable, as follows:

  - If you specify either the LN or LOG10 option and some subjects have the first independent variable (dose) values less than or equal to zero, these subjects are treated as a control group. The initial estimate of C is then the ratio of the number of responses to the number of subjects in this group.

- If you do not specify the LN or LOG10 option or if there is no control group, then one of the following occurs:

  · If all responses are greater than zero, the initial estimate of the natural response rate is the minimal response rate (the ratio of the number of responses to the number of subjects in a dose group) across all dose levels.

  · If one or more of the responses is zero (making the response rate zero in that dose group), the initial estimate of the natural rate is the reciprocal of twice the largest number of subjects in any dose group in the experiment.

DATA=ered*SAS-data-set*

specifies the SAS data set to be used by PROC PROBIT. By default, the procedure uses the most recently created SAS data set.

GOUT=graphics-catalog*

specifies a graphics catalog in which to save graphics output.

HPROB=p*

specifies a minimum probability level for the Pearson's chi-square to indicate a good fit. The default value is 0.10. The LACKFIT option must also be specified for this option to have any effect. For Pearson's goodness-of-fit chi-square values with probability greater than the HPROB= value, the fiducial limits, if requested with the INVERSECL option, are computed by using a critical value of 1.96. For chi-square values with probability less than the value of the HPROB= option, the critical value is a 0.95 two-sided quantile value taken from the t distribution with degrees of freedom equal to $(k-1) \times m - q$, where k is the number of levels for the response variable, m is the number of different sets of independent variable values, and q is the number of parameters fit in the model. Note that the HPROB= option can also appear in the MODEL statement.

INEST=SAS-data-set*

specifies an input SAS data set that contains initial estimates for all the parameters in the model. See the section "INEST= *SAS-data-set*" on page 6407 for a detailed description of the contents of the INEST= data set.

INVERSECL< (PROB=rates) >*

computes confidence limits for the values of the first continuous independent variable (such as dose) that yield selected response rates. You can optionally specify a list of response rates as *rates*. The response rates must be between zero and one, and can be a list separated by blanks, commas, or in the form of a DO list.

For example,

```
PROB = .1 TO .9 by .1
PROB = .1 .2 .3 .4
PROB = .01, .25, .75, .9
```

are valid lists of response rates.

If the algorithm fails to converge (this can happen when C is nonzero), missing values are reported for the confidence limits. See the section "Inverse Confidence Limits" on page 6410 for details. Note that the INVERSECL option can also appear in the MODEL statement.

LACKFIT

performs two goodness-of-fit tests (a Pearson's chi-square test and a log-likelihood ratio chi-square test) for the fitted model.

To compute the test statistics, proper grouping of the observations into subpopulations is needed. You can use the AGGREGATE or AGGREGATE= option for this end. See the entry for the AGGREGATE and AGGREGATE= options under the MODEL statement. If neither AGGREGATE nor AGGREGATE= is specified, PROC PROBIT assumes each observation is from a separate subpopulation and computes the goodness-of-fit test statistics only for the *events/trials* syntax.

NOTE: This test is not appropriate if the data are very sparse, with only a few values at each set of the independent variable values.

If the Pearson's chi-square test statistic is significant, then the covariance estimates and standard error estimates are adjusted. See the section "Lack-of-Fit Tests" on page 6408 for a description of the tests. Note that the LACKFIT option can also appear in the MODEL statement.

LOG

LN

analyzes the data by replacing the first continuous independent variable with its natural logarithm. This variable is usually the level of some treatment such as dosage. In addition to the usual output given by the INVERSECL option, the estimated dose values and 95% fiducial limits for dose are also displayed. If you specify the OPTC option, any observations with a dose value less than or equal to zero are used in the estimation as a control group. If you do not specify the OPTC option with the LOG or LN option, then any observations with the first continuous independent variable values less than or equal to zero are ignored.

LOG10

specifies an analysis like that of the LN or LOG option, except that the common logarithm (log to the base 10) of the dose value is used rather than the natural logarithm.

NAMELEN=n

specifies the length of effect names in tables and output data sets to be n characters, where n is a value between 20 and 200. The default length is 20 characters.

NOPRINT

suppresses the display of all output including graphics. Note that this option temporarily disables the Output Delivery System (ODS). For more information, see Chapter 20, "Using the Output Delivery System."

OPTC

controls how the natural response is handled. See the description of the C= option on page 6353 for details.

ORDER=DATA | FORMATTED | FREQ | INTERNAL

specifies the sort order for the levels of the classification variables (which are specified in the CLASS statement). This option applies to the levels for all classification variables, except when you use the (default) ORDER=FORMATTED option with numeric classification variables that have no explicit format. With this option, the levels of such variables are ordered by their internal value.

The ORDER= option can take the following values:

Value of ORDER=	Levels Sorted By
DATA	Order of appearance in the input data set
FORMATTED	External formatted value, except for numeric variables with no explicit format, which are sorted by their unformatted (internal) value
FREQ	Descending frequency count; levels with the most observations come first in the order
INTERNAL	Unformatted value

By default, ORDER=FORMATTED. For ORDER=FORMATTED and ORDER=INTERNAL, the sort order is machine-dependent.

This order also applies to the levels of the response variable. Response level ordering is important because PROC PROBIT always models the probability of response levels at the beginning of the ordering. See the section "Response Level Ordering" on page 6404 for further details.

For more information about sort order, see the chapter on the SORT procedure in the *Base SAS Procedures Guide* and the discussion of BY-group processing in *SAS Language Reference: Concepts*.

OUTEST=_SAS-data-set_
   specifies a SAS data set to contain the parameter estimates and, if the COVOUT option is specified, their estimated covariances. If you omit this option, the output data set is not created. The contents of the data set are described in the section "OUTEST= *SAS-data-set*" on page 6411.

PLOT | PLOTS < =*plot-request* >
PLOT | PLOTS < =(*plot-request* < . . . *plot-request* >)>

   specifies options that control details of the plots created by ODS Graphics. These plots are related to a dose variable, which is identified as the first single continuous independent variable in the MODEL statement. If there are interaction terms with this variable in the model, the PROBIT procedure will not produce any plot.

   You can specify more than one plot request within the parentheses after PLOTS=. For a single plot request, you can omit the parentheses.

   ODS Graphics must be enabled before plots can be requested. For example:

```
proc probit plots=predplot;
   model r/n = dose;
run;
```

   For more information about enabling and disabling ODS Graphics, see the section "Enabling and Disabling ODS Graphics" on page 600 in Chapter 21, "Statistical Graphics Using ODS."

   The following plot requests are available.

ALL

    creates all appropriate plots.

CDFPLOT< (LEVEL=(*character-list*)) >

    requests the plot of predicted cumulative distribution function (CDF) of the multinomial response variable as a function of a single continuous independent variable (dose variable). This single continuous independent variable must be the first single continuous independent variable listed in the MODEL statement. You can request this plot only with a multinomial model.

    The LEVEL= suboption specifies the levels of the multinomial response variable for which the CDF curves are requested. There are $k - 1$ curves for a k-level multinomial response variable (for the highest level, it is the constant line 1). You can specify any of them to be plotted by the LEVEL= suboption.

IPPPLOT

    requests the inverse plot of the predicted probability against the first single continuous variable (dose variable) in the MODEL statement for the binomial model. You can request this plot only with a binomial model. The confidence limits for the predicted values of the dose variable are the computed fiducial limits, not the inverse of the confidence limits of the predicted probabilities. See the section "Inverse Confidence Limits" on page 6410 for more details.

LPREDPLOT< (LEVEL=(*character-list*)) >

    requests the plot of the linear predictor $\mathbf{x}'\mathbf{b}$ against the first single continuous variable (dose variable) in the MODEL statement for either the binomial model or the multinomial model. The confidence limits for the predicted values are available only for the binomial model.

    For the multinomial model, you can use the LEVEL= suboption to specify the levels for which the linear predictor lines are plotted.

NONE

    suppresses all plots.

PREDPPLOT< (LEVEL=(*character-list*)) >

    requests the plot of the predicted probability against the first single continuous variable (dose variable) in the MODEL statement for both the binomial model and the multinomial model. Confidence limits are available only for the binomial model.

    For the multinomial model, you can use the LEVEL= suboption to specify the levels for which the linear predictor lines are plotted.

XDATA=*SAS-data-set*

    specifies an input SAS data set that contains values for all the independent variables in the MODEL statement and variables in the CLASS statement. If there are covariates specified in a MODEL statement, you specify fixed values for the effects in the MODEL statement by the XDATA= data set when predicted values and/or fiducial limits for a single continuous variable (dose variable) are required. These specified values for the effects in the MODEL statement are also used for generating plots. See the section "XDATA= *SAS-data-set*" on page 6411 for a detailed description of the contents of the XDATA= data set.

BY Statement

> **BY** *variables* ;

You can specify a BY statement with PROC PROBIT to obtain separate analyses of observations in groups that are defined by the BY variables. When a BY statement appears, the procedure expects the input data set to be sorted in order of the BY variables. If you specify more than one BY statement, only the last one specified is used.

If your input data set is not sorted in ascending order, use one of the following alternatives:

- Sort the data by using the SORT procedure with a similar BY statement.

- Specify the NOTSORTED or DESCENDING option in the BY statement for the PROBIT procedure. The NOTSORTED option does not mean that the data are unsorted but rather that the data are arranged in groups (according to values of the BY variables) and that these groups are not necessarily in alphabetical or increasing numeric order.

- Create an index on the BY variables by using the DATASETS procedure (in Base SAS software).

For more information about BY-group processing, see the discussion in *SAS Language Reference: Concepts*. For more information about the DATASETS procedure, see the discussion in the *Base SAS Procedures Guide*.

CDFPLOT Statement

> **CDFPLOT** < **VAR=***variable* > < *options* > ;

The CDFPLOT statement plots the predicted cumulative distribution function (CDF) of the multinomial response variable as a function of a single continuous independent variable (dose variable). You can use this statement only after a multinomial model statement.

VAR=*variable*

> specifies a single continuous variable (dose variable) in the independent variable list of the MODEL statement. If a VAR= *variable* is not specified, the first single continuous variable in the independent variable list of the MODEL statement is used. If such a variable does not exist in the independent variable list of the MODEL statement, an error is reported.

> The predicted cumulative distribution function is defined as

$$\hat{F}_j(\mathbf{x}) = C + (1 - C)F(\hat{a}_j + \mathbf{x}'\hat{\mathbf{b}})$$

where $j = 1, \ldots, k$ are the indexes of the k levels of the multinomial response variable, F is the CDF of the distribution used to model the cumulative probabilities, $\hat{\mathbf{b}}$ is the vector of estimated parameters, $\mathbf{x}$ is the covariate vector, $\hat{a}_j$ are estimated ordinal intercepts with $\hat{a}_1 = 0$, and C is the threshold parameter, either known or estimated from the model. Let x_1 be the covariate corresponding to the dose variable and $\mathbf{x}_{-1}$ be the vector of the rest of the covariates. Let the corresponding estimated parameters be $\hat{b}_1$ and $\hat{\mathbf{b}}_{-1}$. Then

$$\hat{F}_j(\mathbf{x}) = C + (1 - C)F(\hat{a}_j + x_1\hat{b}_1 + \mathbf{x}'_{-1}\hat{\mathbf{b}}_{-1})$$

To plot $\hat{F}_j$ as a function of x_1, $\mathbf{x}_{-1}$ must be specified. You can use the XDATA= option to provide the values of $\mathbf{x}_{-1}$ (see the XDATA= option in the PROC PROBIT statement for details), or use the default values that follow the rules:

- If the effect contains a continuous variable (or variables), the overall mean of this effect is used.

- If the effect is a single classification variable, the highest level of the variable is used.

options

specify the levels of the multinomial response variable for which the CDF curves are requested, and add features to the plot. There are $k - 1$ curves for a k-level multinomial response variable (for the highest level, it is the constant line 1). You can specify any of them to be plotted by the LEVEL= option in the CDFPLOT statement. See the LEVEL= option for how to specify the levels.

An attached box on the right side of the plot is used to label these curves with the names of their levels. You can specify the color of this box by using the CLABBOX= option.

You can use options in the CDFPLOT statement to do the following:

- superimpose specification limits

- specify the levels for which the CDF curves are requested

- specify graphical enhancements (such as color or text height)

Summary of Options

Table 75.2 through Table 75.8 summarize the *options* available in the CDFPLOT statement. The "Dictionary of Options" on page 6362 describes each option in detail.

CDF Options

Table 75.2 Options for CDFPLOT

LEVEL=(*character-list*)	Specifies the names of the levels for which the CDF curves are requested
NOTHRESH	Suppresses the threshold line
THRESHLABPOS=*value*	Specifies the position for the label of the threshold line

General Options

Table 75.3 Color Options

CAXIS=*color*	Specifies color for axis
CFIT=*color*	Specifies color for fitted curves
CFRAME=*color*	Specifies color for frame
CGRID=*color*	Specifies color for grid lines
CHREF=*color*	Specifies color for HREF= lines
CLABBOX=*color*	Specifies color for label box
CTEXT=*color*	Specifies color for text
CVREF=*color*	Specifies color for VREF= lines

Table 75.4 Options to Enhance Plots Produced on Graphics Devices

ANNOTATE= *SAS-data-set*	Specifies an Annotate data set
INBORDER	Requests a border around plot
LFIT=*linetype*	Specifies line style for fitted curves
LGRID=*linetype*	Specifies line style for grid lines
NOFRAME	Suppresses the frame around plotting areas
NOGRID	Suppresses grid lines
NOFIT	Suppresses CDF curves
NOHLABEL	Suppresses horizontal labels
NOHTICK	Suppresses horizontal ticks
NOVTICK	Suppresses vertical ticks
TURNVLABELS	Vertically strings out characters in vertical labels
WFIT=*n*	Specifies thickness for fitted curves
WGRID=*n*	Specifies thickness for grids
WREFL=*n*	Specifies thickness for reference lines

Table 75.5 Axis Options

HAXIS=*value1 to value2* *< by value3 >*	Specifies tick mark values for horizontal axis
HOFFSET=*value*	Specifies offset for horizontal axis
HLOWER=*value*	Specifies lower limit on horizontal axis scale
HUPPER=*value*	Specifies upper limit on horizontal axis scale
NHTICK=*n*	Specifies number of ticks for horizontal axis
NVTICK=*n*	Specifies number of ticks for vertical axis
VAXIS=*value1 to value2* *< by value3 >*	Specifies tick mark values for vertical axis
VAXISLABEL=*'label'*	Specifies label for vertical axis
VOFFSET=*value*	Specifies offset for vertical axis
VLOWER=*value*	Specifies lower limit on vertical axis scale
VUPPER=*value*	Specifies upper limit on vertical axis scale
WAXIS=*n*	Specifies thickness for axis

Table 75.6 Graphics Catalog Options

DESCRIPTION=*'string'*	Specifies description for graphics catalog member
NAME=*'string'*	Specifies name for plot in graphics catalog

Table 75.7 Options for Text Enhancement

FONT=*font*	Specifies software font for text
HEIGHT=*value*	Specifies height of text outside framed areas
INHEIGHT=*value*	Specifies height of text inside framed areas

Table 75.8 Options for Reference Lines

HREF< (INTERSECT)> =value-list	Requests horizontal reference line
HREFLABELS= ('label1',...,'labeln')	Specifies labels for HREF= lines
HREFLABPOS=*n*	Specifies vertical position of labels for HREF= lines
LHREF=*linetype*	Specifies line style for HREF= lines
LVREF=*linetype*	Specifies line style for VREF= lines
VREF<(INTERSECT)> =value-list	Requests vertical reference line
VREFLABELS= ('label1',...,'labeln')	Specifies labels for VREF= lines
VREFLABPOS=*n*	Specifies horizontal position of labels for VREF= lines

Dictionary of Options

The following entries provide detailed descriptions of the *options* in the CDFPLOT statement.

ANNOTATE=*SAS-data-set*

ANNO=*SAS-data-set*

specifies an Annotate data set, as described in *SAS/GRAPH: Reference*, that enables you to add features to the CDF plot. The ANNOTATE= data set you specify in the CDFPLOT statement is used for all plots created by the statement.

CAXIS=*color*

CAXES=*color*

specifies the color used for the axes and tick marks. This option overrides any COLOR= specifications in an AXIS statement. The default is the first color in the device color list.

CFIT=*color*

specifies the color for the fitted CDF curves. The default is the first color in the device color list.

CFRAME=*color*

CFR=*color*

specifies the color for the area enclosed by the axes and frame. This area is not shaded by default.

CGRID=*color*

specifies the color for grid lines. The default is the first color in the device color list.

CLABBOX=*color*

specifies the color for the area enclosed by the label box for CDF curves. This area is not shaded by default.

CHREF=*color*

CH=*color*

> specifies the color for lines requested by the HREF= option. The default is the first color in the device color list.

CTEXT=*color*

> specifies the color for tick mark values and axis labels. The default is the color specified for the CTEXT= option in the most recent GOPTIONS statement.

CVREF=*color*

CV=*color*

> specifies the color for lines requested by the VREF= option. The default is the first color in the device color list.

DESCRIPTION="*string*"

DES="*string*"

> specifies a description, up to 40 characters, that appears in the PROC GREPLAY master menu. The default is the variable name.

FONT=*font*

> specifies a software font for reference line and axis labels. You can also specify fonts for axis labels in an AXIS statement. The FONT= font takes precedence over the FTEXT= font specified in the most recent GOPTIONS statement. Hardware characters are used by default.

HAXIS=*value1 to value2 < by value3 >*

> specifies tick mark values for the horizontal axis; *value1*, *value2*, and *value3* must be numeric, and *value1* must be less than *value2*. The lower tick mark is *value1*. Tick marks are drawn at increments of *value3*. The last tick mark is the greatest value that does not exceed *value2*. If *value3* is omitted, a value of 1 is used.

> Examples of HAXIS= lists follow:

```
haxis = 0 to 10
haxis = 2 to 10 by 2
haxis = 0 to 200 by 10
```

HEIGHT=*value*

> specifies the height of text used outside framed areas. The default value is 3.846 (in percentage).

HLOWER=*value*

> specifies the lower limit on the horizontal axis scale. The HLOWER= option specifies *value* as the lower horizontal axis tick mark. The tick mark interval and the upper axis limit are determined automatically. This option has no effect if the HAXIS= option is used.

HOFFSET=*value*

> specifies offset for horizontal axis. The default value is 1.

HUPPER=*value*

  specifies *value* as the upper horizontal axis tick mark. The tick mark interval and the lower axis limit
  are determined automatically. This option has no effect if the HAXIS= option is used.

HREF < (INTERSECT) > =*value-list*

  requests reference lines perpendicular to the horizontal axis. If (INTERSECT) is specified, a sec-
  ond reference line perpendicular to the vertical axis is drawn that intersects the fit line at the same
  point as the horizontal axis reference line. If a horizontal axis reference line label is specified, the
  intersecting vertical axis reference line is labeled with the vertical axis value. See also the CHREF=,
  HREFLABELS=, and LHREF= options.

HREFLABELS=*'label1',...,'labeln'*

HREFLABEL=*'label1',...,'labeln'*

HREFLAB=*'label1',...,'labeln'*

  specifies labels for the lines requested by the HREF= option. The number of labels must equal the
  number of lines. Enclose each label in quotes. Labels can be up to 16 characters.

HREFLABPOS=*n*

  specifies the vertical position of labels for HREF= lines. The following table shows valid values for *n*
  and the corresponding label placements.

n	**Label Placement**
1	Top
2	Staggered from top
3	Bottom
4	Staggered from bottom
5	Alternating from top
6	Alternating from bottom

INBORDER

  requests a border around CDF plots.

INHEIGHT=*value*

  Specifies height of text inside framed areas.

LEVEL=(*character-list***)**

ORDINAL=(*character-list***)**

  specifies the names of the levels for which CDF curves are requested. Names should be quoted and
  separated by space. If there is no correct name provided, no CDF curve is plotted.

LFIT=*linetype*

  specifies a line style for fitted curves. By default, fitted curves are drawn by connecting solid lines
  (*linetype* = 1).

LGRID=*linetype*

  specifies a line style for all grid lines. *linetype* is between 1 and 46. The default is 35.

LHREF=*linetype*

LH=*linetype*

    specifies the line type for lines requested by the HREF= option. The default is 2, which produces a dashed line.

LVREF=*linetype*

LV=*linetype*

    specifies the line type for lines requested by the VREF= option. The default is 2, which produces a dashed line.

NAME=*'string'*

    specifies a name for the plot, up to eight characters, that appears in the PROC GREPLAY master menu. The default is 'PROBIT'.

NHTICK=*n*

    Specifies number of ticks for horizontal axis.

NVTICK=*n*

    Specifies number of ticks for vertical axis.

NOFIT

    suppresses the fitted CDF curves.

NOFRAME

    suppresses the frame around plotting areas.

NOGRID

    suppresses grid lines.

NOHLABEL

    suppresses horizontal labels.

NOHTICK

    suppresses horizontal tick marks.

NOTHRESH

    suppresses the threshold line.

NOVLABEL

    suppresses vertical labels.

TURNVLABELS

    vertically strings out characters in vertical labels.

NOVTICK

    suppresses vertical tick marks.

THRESHLABPOS=*n*

    specifies the horizontal position of labels for the threshold line. The following table shows valid values for *n* and the corresponding label placements.

n	Label Placement
1	Left
2	Right

VAXIS=value1 to value2 < by value3 >

specifies tick mark values for the vertical axis; *value1*, *value2*, and *value3* must be numeric, and *value1* must be less than *value2*. The lower tick mark is *value1*. Tick marks are drawn at increments of *value3*. The last tick mark is the greatest value that does not exceed *value2*. This method of specification of tick marks is not valid for logarithmic axes. If *value3* is omitted, a value of 1 is used.

Examples of VAXIS= lists follow:

```
vaxis = 0 to 10
vaxis = 0 to 2 by .1
```

VAXISLABEL='string'

specifies a label for the vertical axis.

VLOWER=value

specifies the lower limit on the vertical axis scale. The VLOWER= option specifies *value* as the lower vertical axis tick mark. The tick mark interval and the upper axis limit are determined automatically. This option has no effect if the VAXIS= option is used.

VOFFSET=value

specifies the offset for the vertical axis.

VREF=value-list

requests reference lines perpendicular to the vertical axis. If (INTERSECT) is specified, a second reference line perpendicular to the horizontal axis is drawn that intersects the fit line at the same point as the vertical axis reference line. If a vertical axis reference line label is specified, the intersecting horizontal axis reference line is labeled with the horizontal axis value. See also the CVREF=, LVREF=, and VREFLABELS= options.

VREFLABELS='label1',...,'labeln'

VREFLABEL='label1',...,'labeln'

VREFLAB='label1',...,'labeln'

specifies labels for the lines requested by the VREF= option. The number of labels must equal the number of lines. Enclose each label in quotes. Labels can be up to 16 characters.

VREFLABPOS=n

specifies the horizontal position of labels for VREF= lines. The following table shows valid values for *n* and the corresponding label placements.

n	Label Placement
1	Left
2	Right

VUPPER=*value*

specifies the upper limit on the vertical axis scale. The VUPPER= option specifies *value* as the upper vertical axis tick mark. The tick mark interval and the lower axis limit are determined automatically. This option has no effect if the VAXIS= option is used.

WAXIS=*n*

specifies line thickness for axes and frame. The default value is 1.

WFIT=*n*

specifies line thickness for fitted curves. The default value is 1.

WGRID=*n*

specifies line thickness for grids. The default value is 1.

WREFL=*n*

specifies line thickness for reference lines. The default value is 1.

CLASS Statement

CLASS *variables* < / **TRUNCATE** > **;**

The CLASS statement names the classification variables to be used in the model. Typical classification variables are Treatment, Sex, Race, Group, and Replication. If you use the CLASS statement, it must appear before the MODEL statement.

Classification variables can be either character or numeric. By default, class levels are determined from the entire set of formatted values of the CLASS variables.

NOTE: Prior to SAS 9, class levels were determined by using no more than the first 16 characters of the formatted values. To revert to this previous behavior, you can use the TRUNCATE option in the CLASS statement.

In any case, you can use formats to group values into levels. See the discussion of the FORMAT procedure in the *Base SAS Procedures Guide* and the discussions of the FORMAT statement and SAS formats in *SAS Formats and Informats: Reference*. You can adjust the order of CLASS variable levels with the ORDER= option in the PROC PROBIT statement. You can specify the following option in the CLASS statement after a slash (/):

TRUNCATE

specifies that class levels should be determined by using only up to the first 16 characters of the formatted values of CLASS variables. When formatted values are longer than 16 characters, you can use this option to revert to the levels as determined in releases prior to SAS 9.

EFFECTPLOT Statement

EFFECTPLOT < *plot-type* < (*plot-definition-options*) > > < / *options* > **;**

The EFFECTPLOT statement produces a display of the fitted model and provides options for changing and enhancing the displays. Table 75.9 describes the available *plot-types* and their *plot-definition-options*.

Table 75.9 *Plot-Types* and *Plot-Definition-Options*

Plot-Type and Description	Plot-Definition-Options
BOX Displays a box plot of continuous response data at each level of a CLASS effect, with predicted values superimposed and connected by a line. This is an alternative to the INTERACTION *plot-type*.	PLOTBY= variable or CLASS effect X= CLASS variable or effect
CONTOUR Displays a contour plot of predicted values against two continuous covariates.	PLOTBY= variable or CLASS effect X= continuous variable Y= continuous variable
FIT Displays a curve of predicted values versus a continuous variable.	PLOTBY= variable or CLASS effect X= continuous variable
INTERACTION Displays a plot of predicted values (possibly with error bars) versus the levels of a CLASS effect. The predicted values are connected with lines and can be grouped by the levels of another CLASS effect.	PLOTBY= variable or CLASS effect SLICEBY= variable or CLASS effect X= CLASS variable or effect
SLICEFIT Displays a curve of predicted values versus a continuous variable grouped by the levels of a CLASS effect.	PLOTBY= variable or CLASS effect SLICEBY= variable or CLASS effect X= continuous variable

For full details about the syntax and options of the EFFECTPLOT statement, see the section "EFFECTPLOT Statement" on page 411 in Chapter 19, "Shared Concepts and Topics."

ESTIMATE Statement

ESTIMATE < *'label'* > *estimate-specification* < **(**divisor**=**n**)** >
         < **,** . . . < *'label'* > *estimate-specification* < **(**divisor**=**n**)** > >
         < **/** *options* > ;

The ESTIMATE statement provides a mechanism for obtaining custom hypothesis tests. Estimates are formed as linear estimable functions of the form $\mathbf{L}\boldsymbol{\beta}$. You can perform hypothesis tests for the estimable functions, construct confidence limits, and obtain specific nonlinear transformations.

Table 75.10 summarizes the options available in the ESTIMATE statement.

Table 75.10 ESTIMATE Statement Options

Option	Description
Construction and Computation of Estimable Functions	
DIVISOR=	Specifies a list of values to divide the coefficients
NOFILL	Suppresses the automatic fill-in of coefficients for higher-order effects
SINGULAR=	Tunes the estimability checking difference
Degrees of Freedom and *p*-values	
ADJUST=	Determines the method for multiple comparison adjustment of estimates
ALPHA=α	Determines the confidence level $(1 - \alpha)$
LOWER	Performs one-sided, lower-tailed inference
STEPDOWN	Adjusts multiplicity-corrected *p*-values further in a step-down fashion
TESTVALUE=	Specifies values under the null hypothesis for tests
UPPER	Performs one-sided, upper-tailed inference
Statistical Output	
CL	Constructs confidence limits
CORR	Displays the correlation matrix of estimates
COV	Displays the covariance matrix of estimates
E	Prints the **L** matrix
JOINT	Produces a joint F or chi-square test for the estimable functions
SEED=	Specifies the seed for computations that depend on random numbers

For details about the syntax of the ESTIMATE statement, see the section "ESTIMATE Statement" on page 437 in Chapter 19, "Shared Concepts and Topics."

INSET Statement

> **INSET** < *keyword-list* > < / *options* > ;

The box or table of summary information produced on plots made with the CDFPLOT, IPPPLOT, LPRED-PLOT, or PREDPPLOT statement is called an *inset*. You can use the INSET statement to customize both the information that is printed in the inset box and the appearance of the inset box. To supply the information that is displayed in the inset box, you specify *keywords* corresponding to the information you want shown. For example, the following statements produce a predicted probability plot with the number of trials, the number of events, the name of the distribution, and the estimated optimum natural threshold in the inset.

```
proc probit data=epidemic;
   model r/n = dose;
   predpplot;
   inset nobs ntrials nevents dist optc;
run;
```

By default, inset entries are identified with appropriate labels. However, you can provide a customized label by specifying the *keyword* for that entry followed by the equal sign (=) and the label in quotes. For example, the following INSET statement produces an inset containing the number of observations and the name of the distribution, labeled "Sample Size" and "Distribution" in the inset.

```
inset nobs='Sample Size' dist='Distribution';
```

If you specify a keyword that does not apply to the plot you are creating, then the keyword is ignored.

The *options* control the appearance of the box.

If you specify more than one INSET statement, only the first one is used.

Keywords Used in the INSET Statement

Table 75.11 and Table 75.12 list keywords available in the INSET statement to display summary statistics, distribution parameters, and distribution fitting information.

Table 75.11   Summary Statistics

NOBS	Number of observations
NTRIALS	Number of trials
NEVENTS	Number of events
C	User-input threshold
OPTC	Estimated natural threshold
NRESPLEV	Number of levels of the response variable

Table 75.12   General Information

CONFIDENCE	Confidence coefficient for all confidence intervals
DIST	Name of the distribution

Options Used in the INSET Statement

Table 75.13 and Table 75.14 list the options available in the INSET statement.

Table 75.13   Color and Pattern Options

CFILL=*color*	Specifies color for filling box
CFILLH=*color*	Specifies color for filling box header
CFRAME=*color*	Specifies color for frame
CHEADER=*color*	Specifies color for text in header
CTEXT=*color*	Specifies color for text

Table 75.14 General Appearance Options

FONT=*font*	Specifies software font for text
HEIGHT=*value*	Specifies height of text
HEADER=*'quoted string'*	Specifies text for header or box title
NOFRAME	Omits frame around box
POS= *value* <DATA | PERCENT>	Determines the position of the inset. The *value* can be a compass point (N, NE, E, SE, S, SW, W, NW) or a pair of coordinates (x, y) enclosed in parentheses. The coordinates can be specified in axis percentage units or axis data units.
REFPOINT= *name*	Specifies the reference point for an inset that is positioned by a pair of coordinates with the POS= option. You use the REFPOINT= option in conjunction with the POS= coordinates. The REFPOINT= option specifies which corner of the inset frame you have specified with coordinates (x, y), and it can take the value of BR (bottom right), BL (bottom left), TR (top right), or TL (top left). The default is REFPOINT=BL. If the inset position is specified as a compass point, then the REFPOINT= option is ignored.

IPPPLOT Statement

IPPPLOT < *variable* > < *options* > ;

The IPPPLOT statement plots the inverse of the predicted probability (IPP) against a single continuous variable (dose variable) in the MODEL statement for the binomial model. You can only use this statement after a binomial model statement. The confidence limits for the predicted values of the dose variable are the computed fiducial limits, not the inverse of the confidence limits of the predicted probabilities. See the section "Inverse Confidence Limits" on page 6410 for more details.

VAR= *variable*

specifies a single continuous variable (dose variable) in the independent variable list of the MODEL statement. If a VAR= variable is not specified, the first single continuous variable in the independent variable list of the MODEL statement is used. If such a variable does not exist in the independent variable list of the MODEL statement, an error is reported.

For the binomial model, the response variable is a probability. An estimate of the dose level $\hat{x}_1$ needed for a response of p is given by

$$\hat{x}_1 = (F^{-1}(p) - \mathbf{x}'_{-1}\hat{\mathbf{b}}_{-1})/\hat{b}_1$$

where F is the cumulative distribution function used to model the probability, $\mathbf{x}_{-1}$ is the vector of the rest of the covariates, $\hat{\mathbf{b}}_{-1}$ is the vector of the estimated parameters corresponding to $\mathbf{x}_{-1}$, and $\hat{b}_1$ is the estimated parameter for the dose variable of interest.

To plot $\hat{x}_1$ as a function of p, x_{-1} must be specified. You can use the XDATA= option to provide the values of x_{-1} (see the XDATA= option in the PROC PROBIT statement for details), or use the default values that follow the rules:

- If the effect contains a continuous variable (or variables), the overall mean of this effect is used.
- If the effect is a single classification variable, the highest level of the variable is used.

options

add features to the plot.

You can use options in the IPPPLOT statement to do the following:

- superimpose specification limits
- suppress or add the observed data points on the plot
- suppress or add the fiducial limits on the plot
- specify graphical enhancements (such as color or text height)

Summary of Options

Table 75.15 through Table 75.21 summarize the *options* available in the IPPPLOT statement. The "Dictionary of Options" on page 6374 describes each option in detail.

IPP Options

Table 75.15   Plot Layout Options for IPPPLOT

NOCONF	Suppresses fiducial limits
NODATA	Suppresses observed data points on the plot
NOTHRESH	Suppresses the threshold line
THRESHLABPOS=*value*	Specifies the position for the label of the threshold line

General Options

Table 75.16   Color Options

CAXIS=*color*	Specifies color for axis
CFIT=*color*	Specifies color for fitted curves
CFRAME=*color*	Specifies color for frame
CGRID=*color*	Specifies color for grid lines
CHREF=*color*	Specifies color for HREF= lines
CTEXT=*color*	Specifies color for text
CVREF=*color*	Specifies color for VREF= lines

Table 75.17   Options to Enhance Plots Produced on Graphics Devices

ANNOTATE= *SAS-data-set*	Specifies an Annotate data set
INBORDER	Requests a border around plot
LFIT=*linetype*	Specifies line style for fitted curves and confidence limits
LGRID=*linetype*	Specifies line style for grid lines
NOFRAME	Suppresses the frame around plotting areas
NOGRID	Suppresses grid lines
NOFIT	Suppresses fitted curves
NOHLABEL	Suppresses horizontal labels
NOHTICK	Suppresses horizontal ticks
NOVTICK	Suppresses vertical ticks
TURNVLABELS	Vertically strings out characters in vertical labels
WFIT=*n*	Specifies thickness for fitted curves
WGRID=*n*	Specifies thickness for grids
WREFL=*n*	Specifies thickness for reference lines

Table 75.18   Axis Options

HAXIS=*value1 to value2* *< by value3 >*	Specifies tick mark values for horizontal axis
HOFFSET=*value*	Specifies offset for horizontal axis
HLOWER=*value*	Specifies lower limit on horizontal axis scale
HUPPER=*value*	Specifies upper limit on horizontal axis scale
NHTICK=*n*	Specifies number of ticks for horizontal axis
NVTICK=*n*	Specifies number of ticks for vertical axis
VAXIS=*value1 to value2* *< by value3 >*	Specifies tick mark values for vertical axis
VAXISLABEL=*'label'*	Specifies label for vertical axis
VOFFSET=*value*	Specifies offset for vertical axis
VLOWER=*value*	Specifies lower limit on vertical axis scale
VUPPER=*value*	Specifies upper limit on vertical axis scale
WAXIS=*n*	Specifies thickness for axis

Table 75.19 Options for Reference Lines

HREF<(INTERSECT)> =*value-list*	Requests horizontal reference line
HREFLABELS= ('label1',...,'labeln')	Specifies labels for HREF= lines
HREFLABPOS=*n*	Specifies vertical position of labels for HREF= lines
LHREF=*linetype*	Specifies line style for HREF= lines
LVREF=*linetype*	Specifies line style for VREF= lines
VREF<(INTERSECT)> =*value-list*	Requests vertical reference line
VREFLABELS= ('label1',...,'labeln')	Specifies labels for VREF= lines
VREFLABPOS=*n*	Specifies horizontal position of labels for VREF= lines

Table 75.20 Graphics Catalog Options

DESCRIPTION=*'string'*	Specifies description for graphics catalog member
NAME=*'string'*	Specifies name for plot in graphics catalog

Table 75.21 Options for Text Enhancement

FONT=*font*	Specifies software font for text
HEIGHT=*value*	Specifies height of text used outside framed areas
INHEIGHT=*value*	Specifies height of text inside framed areas

Dictionary of Options

The following entries provide detailed descriptions of the *options* in the IPPPLOT statement.

ANNOTATE=*SAS-data-set*

ANNO=*SAS-data-set*

   specifies an Annotate data set, as described in *SAS/GRAPH: Reference*, that enables you to add features to the IPP plot. The ANNOTATE= data set you specify in the IPPPLOT statement is used for all plots created by the statement.

CAXIS=*color*

CAXES=*color*

   specifies the color used for the axes and tick marks. This option overrides any COLOR= specifications in an AXIS statement. The default is the first color in the device color list.

CFIT=*color*

   specifies the color for the fitted IPP curves. The default is the first color in the device color list.

CFRAME=*color*

CFR=*color*

   specifies the color for the area enclosed by the axes and frame. This area is not shaded by default.

CGRID=*color*

   specifies the color for grid lines. The default is the first color in the device color list.

CHREF=*color*

CH=*color*

   specifies the color for lines requested by the HREF= option. The default is the first color in the device color list.

CTEXT=*color*

   specifies the color for tick mark values and axis labels. The default is the color specified for the CTEXT= option in the most recent GOPTIONS statement.

CVREF=*color*

CV=*color*

   specifies the color for lines requested by the VREF= option. The default is the first color in the device color list.

DESCRIPTION=*'string'*

DES=*'string'*

   specifies a description, up to 40 characters, that appears in the PROC GREPLAY master menu. The default is the variable name.

FONT=*font*

   specifies a software font for reference line and axis labels. You can also specify fonts for axis labels in an AXIS statement. The FONT= font takes precedence over the FTEXT= font specified in the most recent GOPTIONS statement. Hardware characters are used by default.

HAXIS=*value1 to value2 < by value3 >*

   specifies tick mark values for the horizontal axis; *value1*, *value2*, and *value3* must be numeric, and *value1* must be less than *value2*. The lower tick mark is *value1*. Tick marks are drawn at increments of *value3*. The last tick mark is the greatest value that does not exceed *value2*. If *value3* is omitted, a value of 1 is used.

   Examples of HAXIS= lists follow:

```
haxis = 0 to 10
haxis = 2 to 10 by 2
haxis = 0 to 200 by 10
```

HEIGHT=*value*

   specifies the height of text used outside framed areas. The default value is 3.846 (in percentage).

HLOWER=*value*

specifies the lower limit on the horizontal axis scale. The HLOWER= option specifies *value* as the lower horizontal axis tick mark. The tick mark interval and the upper axis limit are determined automatically. This option has no effect if the HAXIS= option is used.

HOFFSET=*value*

specifies offset for horizontal axis. The default value is 1.

HUPPER=*value*

specifies *value* as the upper horizontal axis tick mark. The tick mark interval and the lower axis limit are determined automatically. This option has no effect if the HAXIS= option is used.

HREF < (INTERSECT) > =*value-list*

requests reference lines perpendicular to the horizontal axis. If (INTERSECT) is specified, a second reference line perpendicular to the vertical axis is drawn that intersects the fit line at the same point as the horizontal axis reference line. If a horizontal axis reference line label is specified, the intersecting vertical axis reference line is labeled with the vertical axis value. See also the CHREF=, HREFLABELS=, and LHREF= options.

HREFLABELS=*'label1',...,'labeln'*

HREFLABEL=*'label1',...,'labeln'*

HREFLAB=*'label1',...,'labeln'*

specifies labels for the lines requested by the HREF= option. The number of labels must equal the number of lines. Enclose each label in quotes. Labels can be up to 16 characters.

HREFLABPOS=*n*

specifies the vertical position of labels for HREF= lines. The following table shows valid values for *n* and the corresponding label placements.

n	**Label Placement**
1	Top
2	Staggered from top
3	Bottom
4	Staggered from bottom
5	Alternating from top
6	Alternating from bottom

INBORDER

requests a border around IPP plots.

INHEIGHT=*value*

Specifies height of text inside framed areas.

LFIT=*linetype*

specifies a line style for fitted curves and confidence limits. By default, fitted curves are drawn by connecting solid lines (*linetype* = 1) and confidence limits are drawn by connecting dashed lines (*linetype* = 3).

LGRID=_linetype_

> specifies a line style for all grid lines. The value for _linetype_ must be between 1 and 46. The default is 35.

LHREF=_linetype_

LH=_linetype_

> specifies the line type for lines requested by the HREF= option. The default is 2, which produces a dashed line.

LVREF=_linetype_

LV=_linetype_

> specifies the line type for lines requested by the VREF= option. The default is 2, which produces a dashed line.

NAME=_'string'_

> specifies a name for the plot, up to eight characters, that appears in the PROC GREPLAY master menu. The default is 'PROBIT'.

NHTICK=_n_

> Specifies number of ticks for horizontal axis.

NVTICK=_n_

> Specifies number of ticks for vertical axis.

NOCONF

> suppresses fiducial limits from the plot.

NODATA

> suppresses observed data points from the plot.

NOFIT

> suppresses the fitted IPP curves.

NOFRAME

> suppresses the frame around plotting areas.

NOGRID

> suppresses grid lines.

NOHLABEL

> suppresses horizontal labels.

NOHTICK

> suppresses horizontal tick marks.

NOTHRESH

> suppresses the threshold line.

NOVLABEL

> suppresses vertical labels.

TURNVLABELS

vertically strings out characters in vertical labels.

NOVTICK

suppresses vertical tick marks.

THRESHLABPOS=*n*

specifies the vertical position of labels for the threshold line. The following table shows valid values for *n* and the corresponding label placements.

n	**Label Placement**
1	Top
2	Bottom

VAXIS=*value1 to value2 < by value3 >*

specifies tick mark values for the vertical axis; *value1*, *value2*, and *value3* must be numeric, and *value1* must be less than *value2*. The lower tick mark is *value1*. Tick marks are drawn at increments of *value3*. The last tick mark is the greatest value that does not exceed *value2*. This method of specification of tick marks is not valid for logarithmic axes. If *value3* is omitted, a value of 1 is used.

Examples of VAXIS= lists follow:

```
vaxis = 0 to 10
vaxis = 0 to 2 by .1
```

VAXISLABEL=*'string'*

specifies a label for the vertical axis.

VLOWER=*value*

specifies the lower limit on the vertical axis scale. The VLOWER= option specifies *value* as the lower vertical axis tick mark. The tick mark interval and the upper axis limit are determined automatically. This option has no effect if the VAXIS= option is used.

VOFFSET=*value*

specifies the offset for the vertical axis.

VREF=*value-list*

requests reference lines perpendicular to the vertical axis. If (INTERSECT) is specified, a second reference line perpendicular to the horizontal axis is drawn that intersects the fit line at the same point as the vertical axis reference line. If a vertical axis reference line label is specified, the intersecting horizontal axis reference line is labeled with the horizontal axis value. See also the CVREF=, LVREF=, and VREFLABELS= options.

VREFLABELS=*'label1',...,'labeln'*

VREFLABEL=*'label1',...,'labeln'*

VREFLAB=*'label1',...,'labeln'*

specifies labels for the lines requested by the VREF= option. The number of labels must equal the number of lines. Enclose each label in quotes. Labels can be up to 16 characters.

VREFLABPOS=n

    specifies the horizontal position of labels for VREF= lines. The following table shows valid values for n and the corresponding label placements.

n	Label Placement
1	Left
2	Right

VUPPER=*value*

    specifies the upper limit on the vertical axis scale. The VUPPER= option specifies *value* as the upper vertical axis tick mark. The tick mark interval and the lower axis limit are determined automatically. This option has no effect if the VAXIS= option is used.

WAXIS=n

    specifies line thickness for axes and frame. The default value is 1.

WFIT=n

    specifies line thickness for fitted curves. The default value is 1.

WGRID=n

    specifies line thickness for grids. The default value is 1.

WREFL=n

    specifies line thickness for reference lines. The default value is 1.

LPREDPLOT Statement

    LPREDPLOT < **VAR=***variable* > < *options* > ;

The LPREDPLOT statement plots the linear predictor (LPRED) $\mathbf{x}'\mathbf{b}$ against a single continuous variable (dose variable) in the MODEL statement for either the binomial model or the multinomial model. The confidence limits for the predicted values are available only for the binomial model.

VAR= *variable*

    specifies a single continuous variable (dose variable) in the independent variable list of the MODEL statement for which the linear predictor plot is plotted. If a VAR= variable is not specified, the first single continuous variable in the independent variable list of the MODEL statement is used. If such a variable does not exist in the independent variable list of the MODEL statement, an error is reported.

    Let x_1 be the covariate of the dose variable, $\mathbf{x}_{-1}$ be the vector of the rest of the covariates, $\hat{\mathbf{b}}_{-1}$ be the vector of estimated parameters corresponding to $\mathbf{x}_{-1}$, and $\hat{b}_1$ be the estimated parameter for the dose variable of interest.

    To plot $\hat{\mathbf{x}}'\mathbf{b}$ as a function of x_1, $\mathbf{x}_{-1}$ must be specified. You can use the XDATA= option to provide the values of $\mathbf{x}_{-1}$ (see the XDATA= option in the PROC PROBIT statement for details), or use the default values that follow these rules:

- If the effect contains a continuous variable (or variables), the overall mean of this effect is used.
- If the effect is a single classification variable, the highest level of the variable is used.

options

add features to the plot.

For the multinomial model, you can use the LEVEL= option to specify the levels for which the linear predictor lines are plotted. The lines are labeled by the names of their levels in the middle.

You can use options in the LPREDPLOT statement to do the following:

- superimpose specification limits
- suppress or add the observed data points on the plot for the binomial model
- suppress or add the confidence limits for the binomial model
- specify the levels for which the linear predictor lines are requested for the multinomial model
- specify graphical enhancements (such as color or text height)

Summary of Options

Table 75.22 through Table 75.28 list all *options* by function. The "Dictionary of Options" on page 6382 describes each option in detail.

LPRED Options

Table 75.22   Plot Layout Options for LPREDPLOT

LEVEL=(*character-list*)	Specifies the names of the levels for which the linear predictor lines are requested (only for the multinomial model)
NOCONF	Suppresses fiducial limits (only for the binomial model)
NODATA	Suppresses observed data points on the plot (only for the binomial model)
NOTHRESH	Suppresses the threshold line
THRESHLABPOS=*value*	Specifies the position for the label of the threshold line

General Options

Table 75.23   Color Options

CAXIS=*color*	Specifies color for axis
CFIT=*color*	Specifies color for fitted curves
CFRAME=*color*	Specifies color for frame
CGRID=*color*	Specifies color for grid lines
CHREF=*color*	Specifies color for HREF= lines
CTEXT=*color*	Specifies color for text
CVREF=*color*	Specifies color for VREF= lines

Table 75.24 Options to Enhance Plots Produced on Graphics Devices

ANNOTATE=*SAS-data-set*	Specifies an Annotate data set
INBORDER	Requests a border around plot
LFIT=*linetype*	Specifies line style for fitted curves and confidence limits
LGRID=*linetype*	Specifies line style for grid lines
NOFRAME	Suppresses the frame around plotting areas
NOGRID	Suppresses grid lines
NOFIT	Suppresses fitted curves
NOHLABEL	Suppresses horizontal labels
NOHTICK	Suppresses horizontal ticks
NOVTICK	Suppresses vertical ticks
TURNVLABELS	Vertically strings out characters in vertical labels
WFIT=*n*	Specifies thickness for fitted curves
WGRID=*n*	Specifies thickness for grids
WREFL=*n*	Specifies thickness for reference lines

Table 75.25 Axis Options

HAXIS=*value1 to value2 < by value3 >*	Specifies tick mark values for horizontal axis
HOFFSET=*value*	Specifies offset for horizontal axis
HLOWER=*value*	Specifies lower limit on horizontal axis scale
HUPPER=*value*	Specifies upper limit on horizontal axis scale
NHTICK=*n*	Specifies number of ticks for horizontal axis
NVTICK=*n*	Specifies number of ticks for vertical axis
VAXIS=*value1 to value2 < by value3 >*	Specifies tick mark values for vertical axis
VAXISLABEL=*'label'*	Specifies label for vertical axis
VOFFSET=*value*	Specifies offset for vertical axis
VLOWER=*value*	Specifies lower limit on vertical axis scale
VUPPER=*value*	Specifies upper limit on vertical axis scale
WAXIS=*n*	Specifies thickness for axis

Table 75.26 Graphics Catalog Options

DESCRIPTION=*'string'*	Specifies description for graphics catalog member
NAME=*'string'*	Specifies name for plot in graphics catalog

Table 75.27 Options for Text Enhancement

FONT=*font*	Specifies software font for text
HEIGHT=*value*	Specifies height of text used outside framed areas
INHEIGHT=*value*	Specifies height of text inside framed areas

Table 75.28 Options for Reference Lines

HREF<(INTERSECT)> =*value-list*	Requests horizontal reference line
HREFLABELS= ('label1',...,'labeln')	Specifies labels for HREF= lines
HREFLABPOS=*n*	Specifies vertical position of labels for HREF= lines
LHREF=*linetype*	Specifies line style for HREF= lines
LVREF=*linetype*	Specifies line style for VREF= lines
VREF<(INTERSECT)> =*value-list*	Requests vertical reference line
VREFLABELS= ('label1',...,'labeln')	Specifies labels for VREF= lines
VREFLABPOS=*n*	Specifies horizontal position of labels for VREF= lines

Dictionary of Options

The following entries provide detailed descriptions of the *options* in the LPREDPLOT statement.

ANNOTATE=SAS-data-set

ANNO=SAS-data-set

specifies an Annotate data set, as described in *SAS/GRAPH: Reference*, that enables you to add features to the LPRED plot. The ANNOTATE= data set you specify in the LPREDPLOT statement is used for all plots created by the statement.

CAXIS=color

CAXES=color

specifies the color used for the axes and tick marks. This option overrides any COLOR= specifications in an AXIS statement. The default is the first color in the device color list.

CFIT=color

specifies the color for the fitted LPRED lines. The default is the first color in the device color list.

CFRAME=color

CFR=color

specifies the color for the area enclosed by the axes and frame. This area is not shaded by default.

CGRID=*color*

    specifies the color for grid lines. The default is the first color in the device color list.

CHREF=*color*

CH=*color*

    specifies the color for lines requested by the HREF= option. The default is the first color in the device color list.

CTEXT=*color*

    specifies the color for tick mark values and axis labels. The default is the color specified for the CTEXT= option in the most recent GOPTIONS statement.

CVREF=*color*

CV=*color*

    specifies the color for lines requested by the VREF= option. The default is the first color in the device color list.

DESCRIPTION=*'string'*

DES=*'string'*

    specifies a description, up to 40 characters, that appears in the PROC GREPLAY master menu. The default is the variable name.

FONT=*font*

    specifies a software font for reference line and axis labels. You can also specify fonts for axis labels in an AXIS statement. The FONT= font takes precedence over the FTEXT= font specified in the most recent GOPTIONS statement. Hardware characters are used by default.

HAXIS=*value1 to value2 < by value3 >*

    specifies tick mark values for the horizontal axis; *value1*, *value2*, and *value3* must be numeric, and *value1* must be less than *value2*. The lower tick mark is *value1*. Tick marks are drawn at increments of *value3*. The last tick mark is the greatest value that does not exceed *value2*. If *value3* is omitted, a value of 1 is used.

    Examples of HAXIS= lists follow:

```
haxis = 0 to 10
haxis = 2 to 10 by 2
haxis = 0 to 200 by 10
```

HEIGHT=*value*

    specifies the height of text used outside framed areas. The default value is 3.846 (in percentage).

HLOWER=*value*

    specifies the lower limit on the horizontal axis scale. The HLOWER= option specifies *value* as the lower horizontal axis tick mark. The tick mark interval and the upper axis limit are determined automatically. This option has no effect if the HAXIS= option is used.

HOFFSET=*value*

   specifies offset for horizontal axis. The default value is 1.

HUPPER=*value*

   specifies *value* as the upper horizontal axis tick mark. The tick mark interval and the lower axis limit are determined automatically. This option has no effect if the HAXIS= option is used.

HREF < (INTERSECT) > =*value-list*

   requests reference lines perpendicular to the horizontal axis. If (INTERSECT) is specified, a second reference line perpendicular to the vertical axis is drawn that intersects the fit line at the same point as the horizontal axis reference line. If a horizontal axis reference line label is specified, the intersecting vertical axis reference line is labeled with the vertical axis value. See also the CHREF=, HREFLABELS=, and LHREF= options.

HREFLABELS=*'label1',...,'labeln'*

HREFLABEL=*'label1',...,'labeln'*

HREFLAB=*'label1',...,'labeln'*

   specifies labels for the lines requested by the HREF= option. The number of labels must equal the number of lines. Enclose each label in quotes. Labels can be up to 16 characters.

HREFLABPOS=*n*

   specifies the vertical position of labels for HREF= lines. The following table shows valid values for *n* and the corresponding label placements.

n	**Label Placement**
1	Top
2	Staggered from top
3	Bottom
4	Staggered from bottom
5	Alternating from top
6	Alternating from bottom

INBORDER

   requests a border around LPRED plots.

INHEIGHT=*value*

   Specifies height of text inside framed areas.

LEVEL=(*character-list***)**

ORDINAL=(*character-list***)**

   specifies the names of the levels for which linear predictor lines are requested. Names should be quoted and separated by space. If there is no correct name provided, no LPRED line is plotted.

LFIT=*linetype*

   specifies a line style for fitted curves and confidence limits. By default, fitted curves are drawn by connecting solid lines (*linetype* = 1) and confidence limits are drawn by connecting dashed lines (*linetype* = 3).

LGRID=*linetype*

>   specifies a line style for all grid lines. The value for *linetype* is between 1 and 46. The default is 35.

LHREF=*linetype*

LH=*linetype*

>   specifies the line type for lines requested by the HREF= option. The default is 2, which produces a dashed line.

LVREF=*linetype*

LV=*linetype*

>   specifies the line type for lines requested by the VREF= option. The default is 2, which produces a dashed line.

NAME=*'string'*

>   specifies a name for the plot, up to eight characters, that appears in the PROC GREPLAY master menu. The default is 'PROBIT'.

NHTICK=*n*

>   Specifies number of ticks for horizontal axis.

NVTICK=*n*

>   Specifies number of ticks for vertical axis.

NOCONF

>   suppresses confidence limits from the plot. This works only for the binomial model. Confidence limits are not plotted for the multinomial model.

NODATA

>   suppresses observed data points from the plot. This works only for the binomial model. Data points are not plotted for the multinomial model.

NOFIT

>   suppresses the fitted LPRED lines.

NOFRAME

>   suppresses the frame around plotting areas.

NOGRID

>   suppresses grid lines.

NOHLABEL

>   suppresses horizontal labels.

NOHTICK

>   suppresses horizontal tick marks.

NOTHRESH

>   suppresses the threshold line.

NOVLABEL
   suppresses vertical labels.

NOVTICK
   suppresses vertical tick marks.

TURNVLABELS
   vertically strings out characters in vertical labels.

THRESHLABPOS=*n*
   specifies the horizontal position of labels for the threshold line. The following table shows valid
   values for *n* and the corresponding label placements.

n	**Label Placement**
1	Left
2	Right

VAXIS=*value1 to value2 < by value3 >*
   specifies tick mark values for the vertical axis; *value1*, *value2*, and *value3* must be numeric, and
   value1 must be less than *value2*. The lower tick mark is *value1*. Tick marks are drawn at increments
   of *value3*. The last tick mark is the greatest value that does not exceed *value2*. This method of
   specification of tick marks is not valid for logarithmic axes. If *value3* is omitted, a value of 1 is used.

   Examples of VAXIS= lists follow:

   ```
   vaxis = 0 to 10
   vaxis = 0 to 2 by .1
   ```

VAXISLABEL=*'string'*
   specifies a label for the vertical axis.

VLOWER=*value*
   specifies the lower limit on the vertical axis scale. The VLOWER= option specifies *value* as the lower
   vertical axis tick mark. The tick mark interval and the upper axis limit are determined automatically.
   This option has no effect if the VAXIS= option is used.

VOFFSET=*value*
   specifies the offset for the vertical axis.

VREF=*value-list*
   requests reference lines perpendicular to the vertical axis. If (INTERSECT) is specified, a second ref-
   erence line perpendicular to the horizontal axis is drawn that intersects the fit line at the same point as
   the vertical axis reference line. If a vertical axis reference line label is specified, the intersecting hor-
   izontal axis reference line is labeled with the horizontal axis value. See also the CVREF=, LVREF=,
   and VREFLABELS= options.

VREFLABELS=*'label1',...,'labeln'*

VREFLABEL=*'label1',...,'labeln'*

VREFLAB=*'label1',...,'labeln'*

> specifies labels for the lines requested by the VREF= option. The number of labels must equal the number of lines. Enclose each label in quotes. Labels can be up to 16 characters.

VREFLABPOS=*n*

> specifies the horizontal position of labels for VREF= lines. The following table shows valid values for *n* and the corresponding label placements.

n	**Label Placement**
1	Left
2	Right

VUPPER=*number*

> specifies the upper limit on the vertical axis scale. The VUPPER= option specifies *number* as the upper vertical axis tick mark. The tick mark interval and the lower axis limit are determined automatically. This option has no effect if the VAXIS= option is used.

WAXIS=*n*

> specifies line thickness for axes and frame. The default value is 1.

WFIT=*n*

> specifies line thickness for fitted lines. The default value is 1.

WGRID=*n*

> specifies line thickness for grids. The default value is 1.

WREFL=*n*

> specifies line thickness for reference lines. The default value is 1.

LSMEANS Statement

> **LSMEANS** < *model-effects* > < / *options* > **;**

The LSMEANS statement computes and compares least squares means (LS-means) of fixed effects. LS-means are *predicted population margins*—that is, they estimate the marginal means over a balanced population. In a sense, LS-means are to unbalanced designs as class and subclass arithmetic means are to balanced designs.

Table 75.29 summarizes the options available in the LSMEANS statement.

Table 75.29 LSMEANS Statement Options

Option	Description
Construction and Computation of LS-Means	
AT	Modifies the covariate value in computing LS-means
BYLEVEL	Computes separate margins
DIFF	Requests differences of LS-means
OM=	Specifies the weighting scheme for LS-means computation as determined by the input data set
SINGULAR=	Tunes estimability checking
Degrees of Freedom and *p*-values	
ADJUST=	Determines the method for multiple-comparison adjustment of LS-means differences
ALPHA=α	Determines the confidence level $(1 - \alpha)$
STEPDOWN	Adjusts multiple-comparison *p*-values further in a step-down fashion
Statistical Output	
CL	Constructs confidence limits for means and mean differences
CORR	Displays the correlation matrix of LS-means
COV	Displays the covariance matrix of LS-means
E	Prints the **L** matrix
LINES	Produces a "Lines" display for pairwise LS-means differences
MEANS	Prints the LS-means
PLOTS=	Requests graphs of means and mean comparisons
SEED=	Specifies the seed for computations that depend on random numbers

For details about the syntax of the LSMEANS statement, see the section "LSMEANS Statement" on page 453 in Chapter 19, "Shared Concepts and Topics."

LSMESTIMATE Statement

> **LSMESTIMATE** *model-effect* < *'label'* > *values* < *divisor=n* >
>   < , . . . < *'label'* > *values* < *divisor=n* > >
>   < / *options* > ;

The LSMESTIMATE statement provides a mechanism for obtaining custom hypothesis tests among least squares means.

Table 75.30 summarizes the options available in the LSMESTIMATE statement.

Table 75.30 LSMESTIMATE Statement Options

Option	Description
Construction and Computation of LS-Means	
AT	Modifies covariate values in computing LS-means
BYLEVEL	Computes separate margins
DIVISOR=	Specifies a list of values to divide the coefficients
OM=	Specifies the weighting scheme for LS-means computation as determined by a data set
SINGULAR=	Tunes estimability checking
Degrees of Freedom and *p*-values	
ADJUST=	Determines the method for multiple-comparison adjustment of LS-means differences
ALPHA=α	Determines the confidence level $(1 - \alpha)$
LOWER	Performs one-sided, lower-tailed inference
STEPDOWN	Adjusts multiple-comparison *p*-values further in a step-down fashion
TESTVALUE=	Specifies values under the null hypothesis for tests
UPPER	Performs one-sided, upper-tailed inference
Statistical Output	
CL	Constructs confidence limits for means and mean differences
CORR	Displays the correlation matrix of LS-means
COV	Displays the covariance matrix of LS-means
E	Prints the **L** matrix
ELSM	Prints the **K** matrix
JOINT	Produces a joint *F* or chi-square test for the LS-means and LS-means differences
SEED=	Specifies the seed for computations that depend on random numbers

For details about the syntax of the LSMESTIMATE statement, see the section "LSMESTIMATE Statement" on page 470 in Chapter 19, "Shared Concepts and Topics."

MODEL Statement

<*label*:> **MODEL** *response* = *effects* </ *options*> ;

<*label*:> **MODEL** *events/trials* = *effects* </ *options*> ;

The MODEL statement names the variables used as the response and the independent variables. Additionally, you can specify the distribution used to model the response, as well as other options. Only a single MODEL statement can be used with one invocation of the PROBIT procedure. If multiple MODEL statements are present, only the last is used. Main effects and interaction terms can be specified in the MODEL statement, as in the GLM procedure.

The optional *label*, which must be a valid SAS name, is used to label output from the matching MODEL statement.

The *response* can be a single variable with a value that is used to indicate the level of the observed response. For example, the response might be a variable called Symptoms that takes on the values 'None,' 'Mild,' or 'Severe.' Note that, for dichotomous response variables, the probability of the lower sorted value is modeled by default (see the section "Details: PROBIT Procedure" on page 6404). Because the model fit by the PROBIT procedure requires ordered response levels, you might need to use either the ORDER=DATA option in the PROC PROBIT statement or a numeric coding of the response to get the desired ordering of levels.

Alternatively, the response can be specified as a pair of variable names separated by a slash (/). The value of the first variable, *events*, is the number of positive responses (or events). The value of the second variable, *trials*, is the number of trials. Both variables must be numeric and nonnegative, and the ratio of the first variable value to the second variable value must be between 0 and 1, inclusive. For example, the variables might be hits, a variable containing the number of hits for a baseball player, and AtBats, a variable containing the number of times at bat. A model for hitting proportion (batting average) as a function of age could be specified as

```
model hits/AtBats=age;
```

The *effects* following the equal sign are the covariates in the model. Higher-order effects, such as interactions and nested terms, are allowed in the list, as in the GLM procedure. Variable names and combinations of variable names representing higher-order terms are allowed to appear in this list. Classification variables can be used as effects, and indicator variables are generated for the class levels. If you do not specify any covariates following the equal sign, an intercept-only model is fit.

Table 75.31 summarizes the options available in the MODEL statement.

Table 75.31 MODEL Statement Options

Option	Description
AGGREGATE	Specifies the subpopulations
ALPHA=	Sets the significance level
CONVERGE=	Specifies the convergence criterion
CORRB	Displays the estimated correlation matrix
COVB	Displays the estimated covariance matrix
DISTRIBUTION=	Specifies the cumulative distribution function
HPROB=	Specifies a minimum probability level
INITIAL=	Sets initial values for the parameters
INTERCEPT=	Initializes the intercept parameter
INVERSECL	Computes confidence limits
ITPRINT	Displays the iteration history, the final evaluation of the gradient, and the second derivative matrix
LACKFIT	Performs two goodness-of-fit tests
MAXITER=	Specifies the maximum number of iterations
NOINT	Fits a model with no intercept parameter
SCALE=	Specifies the method for estimating the dispersion parameter
SINGULAR=	Specifies the singularity criterion

The following options are available in the MODEL statement.

AGGREGATE

AGGREGATE=*variable-list*

    specifies the subpopulations on which the Pearson's chi-square test statistic and the log-likelihood ratio chi-square test statistic (deviance) are calculated if the LACKFIT option is specified. See the section "Rescaling the Covariance Matrix" on page 6409 for details of Pearson's chi-square and deviance calculations.

    Observations with common values in the given list of variables are regarded as coming from the same subpopulation. Variables in the list can be any variables in the input data set. Specifying the AGGREGATE option is equivalent to specifying the AGGREGATE= option with a variable list that includes all independent variables in the MODEL statement. The PROBIT procedure sorts the input data set according to the variables specified in this list. Information for the sorted data set is reported in the "Response-Covariate Profile" table.

    The deviance and Pearson's goodness-of-fit statistics are calculated if the LACKFIT option is specified in the MODEL statement. The calculated results are reported in the "Goodness-of-Fit" table. If the Pearson's chi-square test is significant with the test level specified by the HPROB= option, the fiducial limits, if required with the INVERSECL option in the MODEL statement, are modified (see the section "Inverse Confidence Limits" on page 6410 for details). Also, the covariance matrix is rescaled by the dispersion parameter when the SCALE= option is specified.

ALPHA=*value*

    sets the significance level for the confidence intervals for regression parameters, fiducial limits for the predicted values, and confidence intervals for the predicted probabilities. The value must be between 0 and 1. The default value is ALPHA=0.05.

CONVERGE=*value*

    specifies the convergence criterion. Convergence is declared when the maximum change in the parameter estimates between Newton-Raphson steps is less than the value specified. The change is a relative change if the parameter is greater than 0.01 in absolute value; otherwise, it is an absolute change.

    By default, CONVERGE=1.0E–8.

CORRB

    displays the estimated correlation matrix of the parameter estimates.

COVB

    displays the estimated covariance matrix of the parameter estimates.

DISTRIBUTION=*distribution-type*

DIST=*distribution-type*

D=*distribution-type*

    specifies the cumulative distribution function used to model the response probabilities. The distributions are described in the section "Details: PROBIT Procedure" on page 6404. Valid values for *distribution-type* are as follows:

NORMAL   the normal distribution for the probit model

LOGISTIC   the logistic distribution for the logit model

EXTREMEVALUE | EXTREME | GOMPERTZ   the extreme value, or Gompertz distribution for the gompit model

By default, DISTRIBUTION=NORMAL.

HPROB=p

specifies a minimum probability level for the Pearson's chi-square to indicate a good fit. The default value is 0.10. The LACKFIT option must also be specified for this option to have any effect. For Pearson's goodness-of-fit chi-square values with probability greater than the HPROB= value, the fiducial limits, if requested with the INVERSECL option, are computed by using a critical value of 1.96. For chi-square values with probability less than the value of the HPROB= option, the critical value is a 0.95 two-sided quantile value taken from the t distribution with degrees of freedom equal to $(k - 1) \times m - q$, where k is the number of levels for the response variable, m is the number of different sets of independent variable values, and q is the number of parameters fit in the model. If you specify the HPROB= option in both the PROC PROBIT and MODEL statements, the MODEL statement option takes precedence.

INITIAL=$values$

sets initial values for the parameters in the model other than the intercept. The values must be given in the order in which the variables are listed in the MODEL statement. If some of the independent variables listed in the MODEL statement are classification variables, then there must be as many values given for that variable as there are classification levels minus 1. The INITIAL option can be specified as follows.

Type of List	Specification
List separated by blanks	`initial=3 4 5`
List separated by commas	`initial=3,4,5`

By default, all parameters have initial estimates of zero.

NOTE: The INITIAL= option is overwritten by the INEST= option in the PROC PROBIT statement.

INTERCEPT=$value$

initializes the intercept parameter to *value*. By default, INTERCEPT=0.

INVERSECL< (PROB=$rates$**) >**

computes confidence limits for the values of the first continuous independent variable (such as dose) that yield selected response rates. You can optionally specify a list of response rates as *rates*. The response rates must be between zero and one; they can be a list separated by blanks, commas, or in the form of a DO list. For example, the following expressions are all valid lists of response rates:

```
PROB = .1 TO .9 by .1
PROB = .1 .2 .3 .4
PROB = .01, .25, .75, .9
```

If the algorithm fails to converge (this can happen when C is nonzero), missing values are reported for the confidence limits. See the section "Inverse Confidence Limits" on page 6410 for details.

ITPRINT

> displays the iteration history, the final evaluation of the gradient, and the second derivative matrix (Hessian).

LACKFIT

> performs two goodness-of-fit tests (a Pearson's chi-square test and a log-likelihood ratio chi-square test) for the fitted model.

> To compute the test statistics, proper grouping of the observations into subpopulations is needed. You can use the AGGREGATE or AGGREGATE= option for this purpose. See the entry for the AGGREGATE and AGGREGATE= options under the MODEL statement. If neither AGGREGATE nor AGGREGATE= is specified, PROC PROBIT assumes each observation is from a separate subpopulation and computes the goodness-of-fit test statistics only for the *events/trials* syntax.

> **NOTE:** This test is not appropriate if the data are very sparse, with only a few values at each set of the independent variable values.

> If the Pearson's chi-square test statistic is significant, then the covariance estimates and standard error estimates are adjusted. See the section "Lack-of-Fit Tests" on page 6408 for a description of the tests. Note that the LACKFIT option can also appear in the PROC PROBIT statement. See the section "PROC PROBIT Statement" on page 6352 for details.

MAXITER=_value_

MAXIT=_value_

> specifies the maximum number of iterations to be performed in estimating the parameters. By default, MAXITER=50.

NOINT

> fits a model with no intercept parameter. If the INTERCEPT= option is also specified, the intercept is fixed at the specified value; otherwise, it is set to zero. This is most useful when the response is binary. When the response has k levels, then $k - 1$ intercept parameters are fit. The NOINT option sets the intercept parameter corresponding to the lowest response level equal to zero. A Lagrange multiplier, or score, test for the restricted model is computed when the NOINT option is specified.

SCALE=_scale_

> enables you to specify the method for estimating the dispersion parameter. To correct for overdispersion or underdispersion, the covariance matrix is multiplied by the estimate of the dispersion parameter. Valid values for **scale** are as follows:

> | | |
> |---|---|
> | D ∣ DEVIANCE | specifies that the dispersion parameter be estimated by the deviance divided by its degrees of freedom. |
> | P ∣ PEARSON | specifies that the dispersion parameter be estimated by the Pearson's chi-square statistic divided by its degrees of freedom. This is set as the default method for estimating the dispersion parameter. |

> You can use the AGGREGATE= option to define the subpopulations for calculating the Pearson's chi-square statistic and the deviance.

> The "Goodness-of-Fit" table includes the Pearson's chi-square statistic, the deviance, their degrees of freedom, the ratio of each statistic divided by its degrees of freedom, and the corresponding *p*-value.

SINGULAR=*value*

> specifies the singularity criterion for determining linear dependencies in the set of independent variables. The sum of squares and crossproducts matrix of the independent variables is formed and swept. If the relative size of a pivot becomes less than the value specified, then the variable corresponding to the pivot is considered to be linearly dependent on the previous set of variables considered. By default, *value*=1E–12.

OUTPUT Statement

> **OUTPUT** < **OUT=***SAS-data-set keyword=name . . . keyword=name* > ;

The OUTPUT statement creates a new SAS data set containing all variables in the input data set and, optionally, the fitted probabilities, the estimate of $\mathbf{x}'\boldsymbol{\beta}$, and the estimate of its standard error. Estimates of the probabilities, $\mathbf{x}'\boldsymbol{\beta}$, and the standard errors are computed for observations with missing response values as long as the values of all the explanatory variables are nonmissing. This enables you to compute these statistics for additional settings of the explanatory variables that are of interest but for which responses are not observed.

You can specify multiple OUTPUT statements. Each OUTPUT statement creates a new data set and applies only to the preceding MODEL statement. If you want to create a SAS data set in a permanent library, you must specify a two-level name. For more information about permanent libraries and SAS data sets, see *SAS Language Reference: Concepts*.

Details on the specifications in the OUTPUT statement are as follows:

keyword=name   specifies the statistics to include in the output data set and assigns names to the new variables that contain the statistics. Specify a keyword for each desired statistic (see the following list of keywords), an equal sign, and the variable to contain the statistic.

The keywords allowed and the statistics they represent are as follows:

> PROB | P   cumulative probability estimates
>
> $$p = C + (1 - C)F(a_j + \mathbf{x}'\boldsymbol{\beta})$$
>
> STD   standard error estimates of $a_j + \mathbf{x}'\mathbf{b}$
>
> XBETA   estimates of $a_j + \mathbf{x}'\boldsymbol{\beta}$

OUT=*SAS-data-set*   names the output data set. By default, the new data set is named by using the DATA*n* convention.

When the *single variable response* syntax is used, the _LEVEL_ variable is added to the output data set, and there are $k - 1$ output observations for each input observation, where k is the number of response levels. There is no observation output corresponding to the highest response level. For each of the $k - 1$ observations, the PROB variable contains the fitted probability of obtaining a response level up to the level indicated by the _LEVEL_ variable, the XBETA variable contains $a_j + \mathbf{x}'\mathbf{b}$, where j references the levels ($a_1 = 0$), and the STD variable contains the standard error estimate of the XBETA variable. See the section "Details: PROBIT Procedure" on page 6404 for the formulas for the parameterizations.

PREDPPLOT Statement

> **PREDPPLOT** < **VAR=***variable* > < *options* > **;**

The PREDPPLOT statement plots the predicted probability against a single continuous variable (dose variable) in the MODEL statement for both the binomial model and the multinomial model. Confidence limits are available only for the binomial model. An attached box on the right side of the plot is used to label predicted probability curves with the names of their levels for the multinomial model. You can specify the color of this box by using the CLABBOX= option.

VAR=*variable*
>   specifies a single continuous variable (dose variable) in the independent variable list of the MODEL statement. If a VAR= variable is not specified, the first single continuous variable in the independent variable list of the MODEL statement is used. If such a variable does not exist in the independent variable list of the MODEL statement, an error is reported.
>
>   The predicted probability is
>
> $$\hat{p} = C + (1 - C)F(\mathbf{x}'\hat{\mathbf{b}})$$
>
>   for the binomial model and
>
> $$\begin{aligned} \hat{p}_1 &= C + (1 - C)F(\mathbf{x}'\hat{\mathbf{b}}) \\ \hat{p}_j &= (1 - C)(F(\hat{a}_j + \mathbf{x}'\hat{\mathbf{b}}) - F(\hat{a}_{j-1} + \mathbf{x}'\hat{\mathbf{b}})) \quad j = 2, \ldots, k - 1 \\ \hat{p}_k &= (1 - C)(1 - F(\hat{a}_{k-1} + \mathbf{x}'\hat{\mathbf{b}})) \end{aligned}$$
>
>   for the multinomial model with k response levels, where F is the cumulative distribution function used to model the probability, $\mathbf{x}'$ is the vector of the covariates, $\hat{a}_j$ are the estimated ordinal intercepts with $\hat{a}_1 = 0$, C is the threshold parameter, either known or estimated from the model, and $\hat{\mathbf{b}}'$ is the vector of estimated parameters.
>
>   To plot $\hat{p}$ (or $\hat{p}_j$) as a function of a continuous variable x_1, the remaining covariates $\mathbf{x}_{-1}$ must be specified. You can use the XDATA= option to provide the values of $\mathbf{x}_{-1}$ (see the XDATA= option in the PROC PROBIT statement for details), or use the default values that follow these rules:
>
>   - If the effect contains a continuous variable (or variables), the overall mean of this effect is used.
>   - If the effect is a single classification variable, the highest level of the variable is used.

options
>   enable you to plot the observed data and add features to the plot.
>
>   You can use options in the PREDPPLOT statement to do the following:
>
>   - superimpose specification limits
>   - suppress or add observed data points for the binomial model
>   - suppress or add confidence limits for the binomial model
>   - specify the levels for which predicted probability curves are requested for the multinomial model
>   - specify graphical enhancements (such as color or text height)

Summary of Options

Table 75.32 through Table 75.38 list all *options* by function. The "Dictionary of Options" on page 6398 describes each option in detail.

PREDPPLOT Options

Table 75.32 Plot Layout Options for PREDPPLOT

LEVEL=(*character-list*)	Specifies the names of the levels for which the predicted probability curves are requested (only for the multinomial model)
NOCONF	Suppresses confidence limits
NODATA	Suppresses observed data points on the plot
NOTHRESH	Suppresses the threshold line
THRESHLABPOS=*value*	Specifies the position for the label of the threshold line

General Options

Table 75.33 Color Options

CAXIS=*color*	Specifies color for the axes
CFIT=*color*	Specifies color for fitted curves
CFRAME=*color*	Specifies color for frame
CGRID=*color*	Specifies color for grid lines
CHREF=*color*	Specifies color for HREF= lines
CLABBOX=*color*	Specifies color for label box
CTEXT=*color*	Specifies color for text
CVREF=*color*	Specifies color for VREF= lines

Table 75.34 Options to Enhance Plots Produced on Graphics Devices

ANNOTATE= *SAS-data-set*	Specifies an Annotate data set
INBORDER	Requests a border around plot
LFIT=*linetype*	Specifies line style for fitted curves and confidence limits
LGRID=*linetype*	Specifies line style for grid lines
NOFRAME	Suppresses the frame around plotting areas
NOGRID	Suppresses grid lines
NOFIT	Suppresses fitted curves
NOHLABEL	Suppresses horizontal labels
NOHTICK	Suppresses horizontal ticks
NOVTICK	Suppresses vertical ticks
TURNVLABELS	Vertically strings out characters in vertical labels
WFIT=*n*	Specifies thickness for fitted curves
WGRID=*n*	Specifies thickness for grids
WREFL=*n*	Specifies thickness for reference lines

Table 75.35 Axis Options

HAXIS=*value1 to value2* *< by value3 >*	Specifies tick mark values for horizontal axis
HOFFSET=*value*	Specifies offset for horizontal axis
HLOWER=*value*	Specifies lower limit on horizontal axis scale
HUPPER=*value*	Specifies upper limit on horizontal axis scale
NHTICK=*n*	Specifies number of ticks for horizontal axis
NVTICK=*n*	Specifies number of ticks for vertical axis
VAXIS=*value1 to value2* *< by value3 >*	Specifies tick mark values for vertical axis
VAXISLABEL=*'label'*	Specifies label for vertical axis
VOFFSET=*value*	Specifies offset for vertical axis
VLOWER=*value*	Specifies lower limit on vertical axis scale
VUPPER=*value*	Specifies upper limit on vertical axis scale
WAXIS=*n*	Specifies thickness for axis

Table 75.36 Graphics Catalog Options

DESCRIPTION=*'string'*	Specifies description for graphics catalog member
NAME=*'string'*	Specifies name for plot in graphics catalog

Table 75.37 Options for Text Enhancement

FONT=*font*	Specifies software font for text
HEIGHT=*value*	Specifies height of text used outside framed areas
INHEIGHT=*value*	Specifies height of text inside framed areas

Table 75.38 Options for Reference Lines

HREF<(INTERSECT)> =*value-list*	Requests horizontal reference line
HREFLABELS= ('label1',...,'labeln')	Specifies labels for HREF= lines
HREFLABPOS=*n*	Specifies vertical position of labels for HREF= lines
LHREF=*linetype*	Specifies line style for HREF= lines
LVREF=*linetype*	Specifies line style for VREF= lines
VREF<(INTERSECT)> =*value-list*	Requests vertical reference line
VREFLABELS= ('label1',...,'labeln')	Specifies labels for VREF= lines
VREFLABPOS=*n*	Specifies horizontal position of labels for VREF= lines

Dictionary of Options

The following entries provide detailed descriptions of the *options* in the PREDPPLOT statement.

ANNOTATE=_SAS-data-set_

ANNO=_SAS-data-set_

> specifies an Annotate data set, as described in *SAS/GRAPH: Reference*, that enables you to add features to the predicted probability plot. The ANNOTATE= data set you specify in the PREDPPLOT statement is used for all plots created by the statement.

CAXIS=_color_

CAXES=_color_

> specifies the color used for the axes and tick marks. This option overrides any COLOR= specifications in an AXIS statement. The default is the first color in the device color list.

CFIT=_color_

> specifies the color for the fitted predicted probability curves. The default is the first color in the device color list.

CFRAME=_color_

CFR=_color_

> specifies the color for the area enclosed by the axes and frame. This area is not shaded by default.

CGRID=*color*
> specifies the color for grid lines. The default is the first color in the device color list.

CHREF=*color*

CH=*color*
> specifies the color for lines requested by the HREF= option. The default is the first color in the device color list.

CTEXT=*color*
> specifies the color for tick mark values and axis labels. The default is the color specified for the CTEXT= option in the most recent GOPTIONS statement.

CVREF=*color*

CV=*color*
> specifies the color for lines requested by the VREF= option. The default is the first color in the device color list.

DESCRIPTION=*'string'*

DES=*'string'*
> specifies a description, up to 40 characters, that appears in the PROC GREPLAY master menu. The default is the variable name.

FONT=*font*
> specifies a software font for reference line and axis labels. You can also specify fonts for axis labels in an AXIS statement. The FONT= font takes precedence over the FTEXT= font specified in the most recent GOPTIONS statement. Hardware characters are used by default.

HAXIS=*value1 to value2* **<** *by value3* **>**
> specifies tick mark values for the horizontal axis; *value1*, *value2*, and *value3* must be numeric, and *value1* must be less than *value2*. The lower tick mark is *value1*. Tick marks are drawn at increments of *value3*. The last tick mark is the greatest value that does not exceed *value2*. If *value3* is omitted, a value of 1 is used.
>
> Examples of HAXIS= lists follow:

```
haxis = 0 to 10
haxis = 2 to 10 by 2
haxis = 0 to 200 by 10
```

HEIGHT=*value*
> specifies the height of text used outside framed areas.

HLOWER=*value*
> specifies the lower limit on the horizontal axis scale. The HLOWER= option specifies *value* as the lower horizontal axis tick mark. The tick mark interval and the upper axis limit are determined automatically. This option has no effect if the HAXIS= option is used.

HOFFSET=*value*

    specifies the offset for the horizontal axis. The default value is 1.

HUPPER=*value*

    specifies *value* as the upper horizontal axis tick mark. The tick mark interval and the lower axis limit are determined automatically. This option has no effect if the HAXIS= option is used.

HREF < (INTERSECT) > =*value-list*

    requests reference lines perpendicular to the horizontal axis. If (INTERSECT) is specified, a second reference line perpendicular to the vertical axis is drawn that intersects the fit line at the same point as the horizontal axis reference line. If a horizontal axis reference line label is specified, the intersecting vertical axis reference line is labeled with the vertical axis value. See also the CHREF=, HREFLABELS=, and LHREF= options.

HREFLABELS=*'label1',...,'labeln'*

HREFLABEL=*'label1',...,'labeln'*

HREFLAB=*'label1',...,'labeln'*

    specifies labels for the lines requested by the HREF= option. The number of labels must equal the number of lines. Enclose each label in quotes. Labels can be up to 16 characters.

HREFLABPOS=*n*

    specifies the vertical position of labels for HREF= lines. The following table shows valid values for *n* and the corresponding label placements.

n	**Label Placement**
1	Top
2	Staggered from top
3	Bottom
4	Staggered from bottom
5	Alternating from top
6	Alternating from bottom

INBORDER

    requests a border around predicted probability plots.

INHEIGHT=*value*

    Specifies height of text inside framed areas.

LEVEL=(*character-list***)**

ORDINAL= (*character-list***)**

    specifies the names of the levels for which predicted probability curves are requested. Names should be quoted and separated by space. If there is no correct name provided, no fitted probability curve is plotted.

LFIT=*linetype*

    specifies a line style for fitted curves and confidence limits. By default, fitted curves are drawn by connecting solid lines (*linetype* = 1) and confidence limits are drawn by connecting dashed lines (*linetype* = 3).

LGRID=*linetype*

> specifies a line style for all grid lines. The value for *linetype* is between 1 and 46. The default is 35.

LHREF=*linetype*

LH=*linetype*

> specifies the line type for lines requested by the HREF= option. The default is 2, which produces a dashed line.

LVREF=*linetype*

LV=*linetype*

> specifies the line type for lines requested by the VREF= option. The default is 2, which produces a dashed line.

NAME=*'string'*

> specifies a name for the plot, up to eight characters, that appears in the PROC GREPLAY master menu. The default is 'PROBIT'.

NHTICK=*n*

> Specifies number of ticks for horizontal axis.

NVTICK=*n*

> Specifies number of ticks for vertical axis.

NOCONF

> suppresses confidence limits from the plot. This works only for the binomial model. Confidence limits are not plotted for the multinomial model.

NODATA

> suppresses observed data points from the plot. This works only for the binomial model. The data points are not plotted for the multinomial model.

NOFIT

> suppresses the fitted predicted probability curves.

NOFRAME

> suppresses the frame around plotting areas.

NOGRID

> suppresses grid lines.

NOHLABEL

> suppresses horizontal labels.

NOHTICK

> suppresses horizontal tick marks.

NOTHRESH

> suppresses the threshold line.

NOVLABEL

> suppresses vertical labels.

NOVTICK

    suppresses vertical tick marks.

TURNVLABELS

    vertically strings out characters in vertical labels.

THRESHLABPOS=n

    specifies the horizontal position of labels for the threshold line. The following table shows valid values for n and the corresponding label placements.

n	Label Placement
1	Left
2	Right

VAXIS=$value1$ to $value2$ < by $value3$ >

    specifies tick mark values for the vertical axis; $value1$, $value2$, and $value3$ must be numeric, and $value1$ must be less than $value2$. The lower tick mark is $value1$. Tick marks are drawn at increments of $value3$. The last tick mark is the greatest value that does not exceed $value2$. This method of specification of tick marks is not valid for logarithmic axes. If $value3$ is omitted, a value of 1 is used.

    Examples of VAXIS= lists follow:

```
vaxis = 0 to 10
vaxis = 0 to 2 by .1
```

VAXISLABEL='string'

    specifies a label for the vertical axis.

VLOWER=$value$

    specifies the lower limit on the vertical axis scale. The VLOWER= option specifies $value$ as the lower vertical axis tick mark. The tick mark interval and the upper axis limit are determined automatically. This option has no effect if the VAXIS= option is used.

VOFFSET=$value$

    specifies the offset for the vertical axis.

VREF=$value\text{-}list$

    requests reference lines perpendicular to the vertical axis. If (INTERSECT) is specified, a second reference line perpendicular to the horizontal axis is drawn that intersects the fit line at the same point as the vertical axis reference line. If a vertical axis reference line label is specified, the intersecting horizontal axis reference line is labeled with the horizontal axis value. See also the CVREF=, LVREF=, and VREFLABELS= options.

VREFLABELS='label1',...,'labeln'

VREFLABEL='label1',...,'labeln'

VREFLAB='label1',...,'labeln'

    specifies labels for the lines requested by the VREF= option. The number of labels must equal the number of lines. Enclose each label in quotes. Labels can be up to 16 characters.

VREFLABPOS=*n*

    specifies the horizontal position of labels for VREF= lines. The following table shows valid values for *n* and the corresponding label placements.

n	**Label Placement**
1	Left
2	Right

VUPPER=*value*

    specifies the upper limit on the vertical axis scale. The VUPPER= option specifies *value* as the upper vertical axis tick mark. The tick mark interval and the lower axis limit are determined automatically. This option has no effect if the VAXIS= option is used.

WAXIS=*n*

    specifies line thickness for axes and frame. The default value is 1.

WFIT=*n*

    specifies line thickness for fitted curves. The default value is 1.

WGRID=*n*

    specifies line thickness for grids. The default value is 1.

WREFL=*n*

    specifies line thickness for reference lines. The default value is 1.

SLICE Statement

    SLICE *model-effect* < / *options* > **;**

The SLICE statement provides a general mechanism for performing a partitioned analysis of the LS-means for an interaction. This analysis is also known as an analysis of simple effects.

The SLICE statement uses the same options as the LSMEANS statement, which are summarized in Table 19.21. For details about the syntax of the SLICE statement, see the section "SLICE Statement" on page 498 in Chapter 19, "Shared Concepts and Topics."

STORE Statement

    STORE < **OUT=** > *item-store-name* < / **LABEL=***'label'* > **;**

The STORE statement requests that the procedure save the context and results of the statistical analysis. The resulting item store has a binary file format that cannot be modified. The contents of the item store can be processed with the PLM procedure.

For details about the syntax of the STORE statement, see the section "STORE Statement" on page 501 in Chapter 19, "Shared Concepts and Topics."

TEST Statement

> **TEST** < *model-effects* > < / *options* > ;

The TEST statement enables you to perform chi-square tests for model effects that test Type I, Type II, or Type III hypotheses. By default, the Type III tests are performed. For more information, see Chapter 19, "Shared Concepts and Topics."

WEIGHT Statement

> **WEIGHT** *variable* ;

A WEIGHT statement can be used with PROC PROBIT to weight each observation by the value of the variable specified. The contribution of each observation to the likelihood function is multiplied by the value of the weight variable. Observations with zero, negative, or missing weights are not used in model estimation.

Details: PROBIT Procedure

Missing Values

PROC PROBIT does not use any observations having missing values for any of the independent variables, the response variables, or the weight variable. If only the response variables are missing, statistics requested in the OUTPUT statement are computed.

Response Level Ordering

For binary response data, PROC PROBIT fits the following model by default:

$$\Phi^{-1}\left(\frac{p - C}{1 - C}\right) = x'\beta$$

where p is the probability of the response level identified as the first level in the "Response Profile" table in the output and Φ is the normal cumulative distribution function. By default, the covariate vector $\mathbf{x}$ contains an intercept term. This is sometimes called Abbot's formula.

Because of the symmetry of the normal (and logistic) distribution, the effect of reversing the order of the two response values is to change the signs of β in the preceding equation.

By default, response levels appear in ascending, sorted order (that is, the lowest level appears first, and then the next lowest, and so on). There are a number of ways that you can control the sort order of the response categories and, therefore, which level is assigned the first ordered level. One of the most common sets of response levels is {0,1}, with 1 representing the event with the probability that is to be modeled.

Consider the example where Y takes the values 1 and 0 for event and nonevent, respectively, and EXPO-SURE is the explanatory variable. By default, PROC PROBIT assigns the first ordered level to response level 0, causing the probability of the nonevent to be modeled. There are several ways to change this.

Besides recoding the variable Y, you can do the following:

- assign a format to Y such that the first formatted value (when the formatted values are put in sorted order) corresponds to the event. For the following example, Y=0 could be assigned formatted value 'nonevent' and Y=1 could be assigned formatted value 'event.' Since ORDER=FORMATTED by default, Y=1 becomes the first ordered level. See Example 75.3 for an illustration of this method.

```
proc format;
    value disease 1='event' 0='nonevent';
run;
proc probit;
    model y=exposure;
    format y disease.;
run;
```

- arrange the input data set so that Y=1 appears first and use the ORDER=DATA option in the PROC PROBIT statement. Since ORDER=DATA sorts levels in order of their appearance in the data set, Y=1 becomes the first ordered level. Note that this option causes classification variables to be sorted by their order of appearance in the data set, also.

Computational Method

The log-likelihood function is maximized by means of a ridge-stabilized Newton-Raphson algorithm. Initial regression parameter estimates are set to zero. The INITIAL= and INTERCEPT= options in the MODEL statement can be used to give nonzero initial estimates.

The log-likelihood function, L, is computed as

$$L = \sum_i w_i \ln(p_i)$$

where the sum is over the observations in the data set, w_i is the weight for the ith observation, and p_i is the modeled probability of the observed response. In the case of the events/trials syntax in the MODEL statement, each observation contributes two terms corresponding to the probability of the event and the probability of its complement:

$$L = \sum_i w_i [r_i \ln(p_i) + (n_i - r_i) \ln(1 - p_i)]$$

where r_i is the number of events and n_i is the number of trials for observation i. This log-likelihood function differs from the log-likelihood function for a binomial or multinomial distribution by additive terms consisting of the log of binomial or multinomial coefficients. These terms are parameter-independent and do not affect the model estimation or the standard errors and tests.

The estimated covariance matrix, $\mathbf{V}$, of the parameter estimates is computed as the negative inverse of the information matrix of second derivatives of L with respect to the parameters evaluated at the final parameter estimates. Thus, the estimated covariance matrix is derived from the observed information matrix rather than the expected information matrix (these are generally not the same). The standard error estimates for the parameter estimates are taken as the square roots of the corresponding diagonal elements of $\mathbf{V}$.

If convergence of the maximum likelihood estimates is attained, a Type III chi-square test statistic is computed for each effect, testing whether there is any contribution from any of the levels of the effect. This statistic is computed as a quadratic form in the appropriate parameter estimates by using the corresponding submatrix of the asymptotic covariance matrix estimate. See Chapter 42, "The GLM Procedure," and Chapter 15, "The Four Types of Estimable Functions," for more information about Type III estimable functions.

The asymptotic covariance matrix is computed as the inverse of the observed information matrix. Note that if the NOINT option is specified and classification variables are used, the first classification variable contains a contribution from an intercept term. The results are displayed in an ODS table named "Type3Analysis".

Chi-square tests for individual parameters are Wald tests based on the observed information matrix and the parameter estimates. If an effect has a single degree of freedom in the parameter estimates table, the chi-square test for this parameter is equivalent to the Type III test for this effect.

Prior to SAS 8.2, a multiple-degrees-of-freedom statistic was computed for each effect to test for contribution from any level of the effect. In general, the Type III test statistic in a main-effect-only model (no interaction terms) will be equal to the previously computed effect statistic, unless there are collinearities among the effects. If there are collinearities, the Type III statistic will adjust for them, and the value of the Type III statistic and the number of degrees of freedom might not be equal to those of the previous effect statistic.

The theory behind these tests assumes large samples. If the samples are not large, it might be better to base the tests on log-likelihood ratios. These changes in log likelihood can be obtained by fitting the model twice, once with all the parameters of interest and once leaving out the parameters to be tested. See Cox and Oakes (1984) for a discussion of the merits of some possible test methods.

If some of the independent variables are perfectly correlated with the response pattern, then the theoretical parameter estimates can be infinite. Although fitted probabilities of 0 and 1 are not especially pathological, infinite parameter estimates are required to yield these probabilities. Due to the finite precision of computer arithmetic, the actual parameter estimates are not infinite. Indeed, since the tails of the distributions allowed in the PROBIT procedure become small rapidly, an argument to the cumulative distribution function of around 20 becomes effectively infinite. In the case of such parameter estimates, the standard error estimates and the corresponding chi-square tests are not trustworthy.

Distributions

The distributions, $F(x)$, allowed in the PROBIT procedure are specified with the DISTRIBUTION= option in the MODEL statement. The cumulative distribution functions for the available distributions are

Cumulative Distribution Function	Distribution
$\int_{-\infty}^{x} \frac{1}{\sqrt{2\pi}} \exp\left(-\frac{z^2}{2}\right) \, dz$	Normal
$\frac{1}{1+e^{-x}}$	Logistic
$1 - e^{-e^{x}}$	Extreme value or Gompertz

The variances of these three distributions are not all equal to 1, and their means are not all equal to zero. Their means and variances are shown in the following table, where γ is the Euler constant.

Distribution	Mean	Variance
Normal	0	1
Logistic	0	$\pi^2/3$
Extreme value or Gompertz	$-\gamma$	$\pi^2/6$

When comparing parameter estimates by using different distributions, you need to take into account the different scalings and, for the extreme value (or Gompertz) distribution, a possible shift in location. For example, if the fitted probabilities are in the neighborhood of 0.1 to 0.9, then the parameter estimates from the logistic model should be about $\pi/\sqrt{3}$ larger than the estimates from the probit model.

INEST= *SAS-data-set*

The INEST= data set names a SAS data set that specifies initial estimates for all the parameters in the model.

The INEST= data set must contain the intercept variables (named Intercept for binary response model and Intercept, Intercept2, Intercept3, and so forth, for multinomial response models) and all independent variables in the MODEL statement.

If BY processing is used, the INEST= data set should also include the BY variables, and there must be at least one observation for each BY group. If there is more than one observation in a BY group, the first one read is used for that BY group.

If the INEST= data set also contains the _TYPE_ variable, only observations with the _TYPE_ value "PARMS" are used as starting values. Combining the INEST= data set and the option MAXIT= in the MODEL statement, partial scoring can be done, such as predicting on a validation data set by using the model built from a training data set.

You can specify starting values for the iterative algorithm in the INEST= data set. This data set overwrites the INITIAL= option in the MODEL statement, which is a little difficult to use for models with multilevel interaction effects. The INEST= data set has the same structure as the "OUTEST= *SAS-data-set*" on page 6411, but it is not required to have all the variables or observations that appear in the OUTEST= data set. One simple use of the INEST= option is passing the previous OUTEST= data set directly to the next model as an INEST= data set, assuming that the two models have the same parameterization.

Model Specification

For a two-level response, the probability that the lesser response occurs is modeled by the probit equation as

$$p = C + (1 - C)F(\mathbf{x}'\mathbf{b})$$

The probability of the other (complementary) event is $1 - p$.

For a multilevel response with outcomes labeled l_i for $i = 1, 2, \ldots, k$, the probability, p_j, of observing level l_j is as follows:

$$
\begin{aligned}
p_1 &= C + (1 - C)F(\mathbf{x}'\mathbf{b}) \\
p_2 &= (1 - C)\left(F(a_2 + \mathbf{x}'\mathbf{b}) - F(\mathbf{x}'\mathbf{b})\right) \\
&\vdots \\
p_j &= (1 - C)\left(F(a_j + \mathbf{x}'\mathbf{b}) - F(a_{j-1} + \mathbf{x}'\mathbf{b})\right) \\
&\vdots \\
p_k &= (1 - C)(1 - F(a_{k-1} + \mathbf{x}'\mathbf{b}))
\end{aligned}
$$

Thus, for a k-level response, there are $k - 2$ additional parameters, $a_2, a_3, \ldots, a_{k-1}$, estimated. These parameters are denoted by Intercept$j$, $j = 2, 3, \ldots, k - 1$, in the output.

An intercept parameter is always added to the set of independent variables as the first term in the model unless the NOINT option is specified in the MODEL statement. If a classification variable taking on k levels is used as one of the independent variables, a set of k indicator variables is generated to model the effect of this variable. Because of the presence of the intercept term, there are at most $k - 1$ degrees of freedom for this effect in the model.

Lack-of-Fit Tests

Two goodness-of-fit tests can be requested from the PROBIT procedure: a Pearson's chi-square test and a log-likelihood ratio chi-square test.

To compute the test statistics, you can use the AGGREGATE or AGGREGATE= option grouping the observations into subpopulations. If neither AGGREGATE nor AGGREGATE= is specified, PROC PROBIT assumes that each observation is from a separate subpopulation and computes the goodness-of-fit test statistics only for the *events/trials* syntax.

If the Pearson's goodness-of-fit chi-square test is requested and the p-value for the test is too small, variances and covariances are adjusted by a heterogeneity factor (the goodness-of-fit chi-square divided by its degrees of freedom) and a critical value from the t distribution is used to compute the fiducial limits. The Pearson's chi-square test statistic is computed as

$$
\chi_P^2 = \sum_{i=1}^{m} \sum_{j=1}^{k} \frac{(r_{ij} - n_i \hat{p}_{ij})^2}{n_i \hat{p}_{ij}}
$$

where the sum on i is over grouping, the sum on j is over levels of response, r_{ij} is the frequency of response level j for the ith grouping, n_i is the total frequency for the ith grouping, and $\hat{p}_{ij}$ is the fitted probability for the jth level at the ith grouping.

The likelihood ratio chi-square test statistic is computed as

$$
\chi_D^2 = 2 \sum_{i=1}^{m} \sum_{j=1}^{k} r_{ij} \ln\left(\frac{r_{ij}}{n_i \hat{p}_{ij}}\right)
$$

This quantity is sometimes called the deviance. If the modeled probabilities fit the data, these statistics should be approximately distributed as chi-square with degrees of freedom equal to $(k - 1) \times m - q$, where k is the number of levels of the multinomial or binomial response, m is the number of sets of independent variable values (covariate patterns), and q is the number of parameters fit in the model.

In order for the Pearson's statistic and the deviance to be distributed as chi-square, there must be sufficient replication within the groupings. When this is not true, the data are sparse, and the p-values for these statistics are not valid and should be ignored. Similarly, these statistics, divided by their degrees of freedom, cannot serve as indicators of overdispersion. A large difference between the Pearson's statistic and the deviance provides some evidence that the data are too sparse to use either statistic.

Rescaling the Covariance Matrix

One way of correcting overdispersion is to multiply the covariance matrix by a dispersion parameter. You can supply the value of the dispersion parameter directly, or you can estimate the dispersion parameter based on either the Pearson's chi-square statistic or the deviance for the fitted model.

The Pearson's chi-square statistic χ_P^2 and the deviance χ_D^2 are defined in the section "Lack-of-Fit Tests" on page 6408. If the SCALE= option is specified in the MODEL statement, the dispersion parameter is estimated by

$$
\widehat{\sigma}^2 = \begin{cases}
\chi_P^2/(m(k-1) - q) & \text{SCALE=PEARSON} \\
\chi_D^2/(m(k-1) - q) & \text{SCALE=DEVIANCE} \\
(constant)^2 & \text{SCALE=}constant
\end{cases}
$$

In order for the Pearson's statistic and the deviance to be distributed as chi-square, there must be sufficient replication within the subpopulations. When this is not true, the data are sparse, and the p-values for these statistics are not valid and should be ignored. Similarly, these statistics, divided by their degrees of freedom, cannot serve as indicators of overdispersion. A large difference between the Pearson's statistic and the deviance provides some evidence that the data are too sparse to use either statistic.

You can use the AGGREGATE (or AGGREGATE=) option to define the subpopulation profiles. If you do not specify this option, each observation is regarded as coming from a separate subpopulation. For *events/trials* syntax, each observation represents n Bernoulli trials, where n is the value of the *trials* variable; for *single-trial* syntax, each observation represents a single trial. Without the AGGREGATE (or AGGREGATE=) option, the Pearson's chi-square statistic and the deviance are calculated only for *events/trials* syntax.

Note that the parameter estimates are not changed by this method. However, their standard errors are adjusted for overdispersion, affecting their significance tests.

Tolerance Distribution

For a single independent variable, such as a dosage level, the models for the probabilities can be justified on the basis of a population with mean μ and scale parameter σ of tolerances for the subjects. Then, given a dose x, the probability, P, of observing a response in a particular subject is the probability that the subject's tolerance is less than the dose or

$$
P = F\left(\frac{x - \mu}{\sigma}\right)
$$

Thus, in this case, the intercept parameter, b_0, and the regression parameter, b_1, are related to μ and σ by

$$b_0 = -\frac{\mu}{\sigma}, \ b_1 = \frac{1}{\sigma}$$

NOTE: The parameter σ is not equal to the standard deviation of the population of tolerances for the logistic and extreme value distributions.

Inverse Confidence Limits

In bioassay problems, estimates of the values of the independent variables that yield a desired response are often needed. For instance, the value yielding a 50% response rate (called the ED50 or LD50) is often used. The INVERSECL option requests that confidence limits be computed for the value of the independent variable that yields a specified response. These limits are computed only for the first continuous variable effect in the model. The other variables are set either at their mean values if they are continuous or at the reference (last) level if they are discrete variables. For a discussion of inverse confidence limits, see Hubert, Bohidar, and Peace (1988).

For the PROBIT procedure, the response variable is a probability. An estimate of the first continuous variable value needed to achieve a response of p is given by

$$\hat{x}_1 = \frac{1}{b_1} \left(F^{-1}(p) - \mathbf{x}^{*\prime} \mathbf{b}^* \right)$$

where F is the cumulative distribution function used to model the probability, $\mathbf{x}^*$ is the vector of independent variables excluding the first one, which can be specified by the XDATA= option described in the section "XDATA= *SAS-data-set*" on page 6411, $\mathbf{b}^*$ is the vector of parameter estimates excluding the first one, and b_1 is the estimated regression coefficient for the independent variable of interest. This estimate assumes that there is no natural response rate ($C = 0$). When C is nonzero, the quantiles and confidence limits for the independent variable correspond to the adjusted probability $C + (1 - C)p$, rather than to p. As a result, an estimate of the value yielding response rate p is associated with the $(p-C)/(1-C)$ quantile. For example, if $C = 0.1$ then an estimate of the LD50 is found corresponding to the 0.44 quantile. This value can be thought of as yielding 50% of the variable's effect, but a 44% response rate. For both binary and ordinal models, the INVERSECL option provides estimates of the value of x_1, which yields $\Pr(\text{first response level}) = p$, for various values of p.

This estimator is given as a ratio of random variables, such as $r = a/b$. Confidence limits for this ratio can be computed by using Fieller's theorem. A brief description of this theorem follows. See Finney (1971) for a more complete description of Fieller's theorem.

If the random variables a and b are thought to be distributed as jointly normal, then for any fixed value r the following probability statement holds if z is an $\alpha/2$ quantile from the standard normal distribution and $\mathbf{V}$ is the variance-covariance matrix of a and b:

$$\Pr \left((a - rb)^2 > z^2 (V_{aa} - 2rV_{ab} + r^2 V_{bb}) \right) = \alpha$$

Usually the inequality can be solved for r to yield a confidence interval. The PROBIT procedure uses a value of 1.96 for z, corresponding to an α value of 0.05, unless the goodness-of-fit p-value is less than the specified value of the HPROB= option. When this happens, the covariance matrix is scaled by the heterogeneity factor, and a t distribution quantile is used for z.

It is possible for the roots of the equation for *r* to be imaginary or for the confidence interval to be all points outside of an interval. In these cases, the limits are set to missing by the PROBIT procedure.

Although the normal and logistic distribution give comparable fitted values of *p* if the empirically observed proportions are not too extreme, they can give appreciably different values when extrapolated into the tails. Correspondingly, the estimates of the confidence limits and dose values can be different for the two distributions even when they agree quite well in the body of the data. Extrapolation outside of the range of the actual data is often sensitive to model assumptions, and caution is advised if extrapolation is necessary.

OUTEST= *SAS-data-set*

The OUTEST= data set contains parameter estimates and the log likelihood for the model. You can specify a label in the MODEL statement to distinguish between the estimates for different models used by the PROBIT procedure. If you specify the COVOUT option, the OUTEST= data set also contains the estimated covariance matrix of the parameter estimates.

The OUTEST= data set contains each variable used as a dependent or independent variable in any MODEL statement. One observation consists of parameter values for the model with the dependent variable having the value -1. If you specify the COVOUT option, there are additional observations containing the rows of the estimated covariance matrix. For these observations, the dependent variable contains the parameter estimate for the corresponding row variable. The following variables are also added to the data set:

MODEL	a character variable containing the label of the MODEL statement, if present, or blank otherwise
NAME	a character variable containing the name of the dependent variable for the parameter estimates observations or the name of the row for the covariance matrix estimates
TYPE	a character variable containing the type of the observation, either PARMS for parameter estimates or COV for covariance estimates
DIST	a character variable containing the name of the distribution modeled
LNLIKE	a numeric variable containing the last computed value of the log likelihood
_C_	a numeric variable containing the estimated threshold parameter
INTERCEPT	a numeric variable containing the intercept parameter estimates and covariances

Any BY variables specified are also added to the OUTEST= data set.

XDATA= *SAS-data-set*

The XDATA= data set is used for specifying values for the effects in the MODEL statement when predicted values and/or fiducial limits for a single continuous variable (dose variable) are required. It is also used for plots specified by the CDFPLOT, IPPPLOT, LPREDPLOT, and PREDPPLOT statement.

The XDATA= data names a SAS data set that contains user input values for all the independent variables in the MODEL statement and the variables in the CLASS statement. The XDATA= data set has the same structure as the DATA= data set but is not required to have all the variables or observations that appear in the DATA= data set.

The XDATA= data set must contain all the independent variables in the MODEL statement and variables in the CLASS statement. Even though variables in the CLASS statement are not used in the MODEL statement, valid values are required for these variables in the XDATA= data set. Missing values are not allowed. For independent variables in the MODEL statement, although the dose variable's value is not used in the computing of predicted values and/or fiducial limits for the dose variable, missing values are not allowed in the XDATA= data set for any of the independent variables. Missing values are allowed for the dependent variables and other variables if they are included in the XDATA= data set and not listed in the CLASS statement.

If BY processing is used, the XDATA= data set should also include the BY variables, and there must be at least one valid observation for each BY group. If there is more than one valid observation in one BY group, the last one read is used for that BY group.

If there is no XDATA= data set in the PROC PROBIT statement, by default, the PROBIT procedure will use overall mean for effects containing continuous variable (or variables) and the highest level of a single classification variable as reference level. The rules are summarized as follows:

- If the effect contains a continuous variable (or variables), the overall mean of this effect is used.

- If the effect is a single classification variable, the highest level of the variable is used.

Traditional High-Resolution Graphics

This section provides examples of using syntax available with the traditional high-resolution plots. A more modern alternative is to use ODS Graphics. See the section "ODS Graphics" on page 6415 for details.

There are four plot statements that you can use to request traditional high-resolution plots: CDFPLOT, IPPPLOT, LPREDPLOT, and PREDPPLOT. Some of these statements apply only to either the binomial model or the multinomial model. Table 75.39 shows the availability of these statements for different models.

Table 75.39 Plot Statement Availability

Statement	Binomial	Multinomial
CDFPLOT	No	Yes
IPPPLOT	Yes	No
LPREDPLOT	Yes	Yes
PREDPPLOT	Yes	Yes

The following example uses the data set study in the section "Estimating the Natural Response Threshold Parameter" on page 6347 to illustrate how to create high-resolution plots for the binomial model:

```
proc probit data=study log10 optc;
   model respond/number=dose;
   predpplot var=dose cfit=blue; inset;
   lpredplot var=dose cfit=blue; inset;
   ippplot   var=dose cfit=blue; inset/pos=se;
run;
```

All plot statements must follow the MODEL statement. The VAR= option specifies a continuous independent variable (dose variable) against which the predicted probability or the linear predictor is plotted. The INSET statement requests the inset box with summary information. See the section "INSET Statement" on page 6369 for more details.

The PREDPPLOT statement creates a plot that shows the relationship between dosage level, observed response proportions, and estimated probability values. See the section "PREDPPLOT Statement" on page 6395 for more details. The IPPPLOT statement creates a similar plot. See the section "IPPPLOT Statement" on page 6371 for details about this plot. The LPREDPLOT statement creates a linear predictor plot, which is described in the section "LPREDPLOT Statement" on page 6379.

The following example uses the data set multi from Example 75.2 to illustrate how to create high-resolution plots for the multinomial model:

```
proc probit data=multi order=data;
   class prep symptoms;
   model symptoms=prep ldose;
   cdfplot var=ldose  level=("None" "Mild" "Severe")
            cfit=blue cframe=ligr noconf;
   lpredplot var=ldose  level=("None" "Mild" "Severe")
            cfit=blue cframe=ligr;
   predpplot var=ldose  level=("None" "Mild" "Severe")
            cfit=blue cframe=ligr;
   weight n;
run;
```

The CDFPLOT statement creates a plot that shows the relationship between the cumulative response probabilities and the dose levels. The multinomial model plots are similar to those with the binomial model.

Displayed Output

If you request the iteration history (ITPRINT), PROC PROBIT displays the following:

- the current value of the log likelihood

- the ridging parameter for the modified Newton-Raphson optimization process

- the current estimate of the parameters

- the current estimate of the parameter C for a natural (threshold) model

- the values of the gradient and the Hessian on the last iteration

If you include classification variables, PROC PROBIT displays the following:

- the numbers of levels for each classification variable

- the (ordered) values of the levels

- the number of observations used

After the model is fit, PROC PROBIT displays the following:

- the name of the input data set

- the name of the dependent variables

- the number of observations used

- the number of events and the number of trials

- the final value of the log-likelihood function

- the parameter estimates

- the standard error estimates of the parameter estimates

- approximate chi-square test statistics for the test

If you specify the COVB or CORRB options, PROC PROBIT displays the following:

- the estimated covariance matrix for the parameter estimates

- the estimated correlation matrix for the parameter estimates

If you specify the LACKFIT option, PROC PROBIT displays the following:

- a count of the number of levels of the response and the number of distinct sets of independent variables

- a goodness-of-fit test based on the Pearson's chi-square

- a goodness-of-fit test based on the likelihood-ratio chi-square

If you specify only one independent variable, the normal distribution is used to model the probabilities, and the response is binary, then PROC PROBIT displays the following:

- the mean MU of the stimulus tolerance

- the scale parameter SIGMA of the stimulus tolerance

- the covariance matrix for MU, SIGMA, and the natural response parameter *C*

If you specify the INVERSECL options, PROC PROBIT also displays the following:

- the estimated dose along with the 95% fiducial limits for probability levels 0.01 to 0.10, 0.15 to 0.85 by 0.05, and 0.90 to 0.99

ODS Table Names

PROC PROBIT assigns a name to each table it creates. You can use these names to reference the table when using the Output Delivery System (ODS) to select tables and create output data sets. These names are listed in the following table. For more information about ODS, see Chapter 20, "Using the Output Delivery System."

Table 75.40   ODS Tables Produced by PROC PROBIT

ODS Table Name	Description	Statement	Option
ClassLevels	Classification variable levels	CLASS	Default
ConvergenceStatus	Convergence status	MODEL	Default
CorrB	Parameter estimate correlation matrix	MODEL	CORRB
CovB	Parameter estimate covariance matrix	MODEL	COVB
CovTolerance	Covariance matrix for location and scale	MODEL	Default
GoodnessOfFit	Goodness-of-fit tests	MODEL	LACKFIT
Heterogeneity	Heterogeneity correction	MODEL	LACKFIT
IterHistory	Iteration history	MODEL	ITPRINT
LagrangeStatistics	Lagrange statistics	MODEL	NOINT
LastGrad	Last evaluation of the gradient	MODEL	ITPRINT
LastHess	Last evaluation of the Hessian	MODEL	ITPRINT
LogProbitAnalysis	Probit analysis for log dose	MODEL	INVERSECL
ModelInfo	Model information	MODEL	Default
MuSigma	Location and scale	MODEL	Default
NObs	Observations summary	PROC	Default
ParameterEstimates	Parameter estimates	MODEL	Default
ParmInfo	Parameter indices	MODEL	Default
ProbitAnalysis	Probit analysis for linear dose	MODEL	INVERSECL
ResponseLevels	Response-covariate profile	MODEL	LACKFIT
ResponseProfiles	Counts for ordinal data	MODEL	Default
Type3Analysis	Type III tests	MODEL	Default

ODS Graphics

Statistical procedures use ODS Graphics to create graphs as part of their output. ODS Graphics is described in detail in Chapter 21, "Statistical Graphics Using ODS."

Before you create graphs, ODS Graphics must be enabled (for example, by specifying the ODS GRAPHICS ON statement). For more information about enabling and disabling ODS Graphics, see the section "Enabling and Disabling ODS Graphics" on page 600 in Chapter 21, "Statistical Graphics Using ODS."

The overall appearance of graphs is controlled by ODS styles. Styles and other aspects of using ODS Graphics are discussed in the section "A Primer on ODS Statistical Graphics" on page 599 in Chapter 21, "Statistical Graphics Using ODS."

These ODS graphs are controlled by the PLOTS= option in the PROC PROBIT statement. You can specify more than one graph request with the PLOTS= option. Table 75.41 summarizes these requests.

Table 75.41 Options for Plots

Option	Plot
ALL	All appropriate plots
CDFPLOT	Estimated cumulative probability
IPPPLOT	Inverse predicted probability
LPREDPLOT	Linear predictor
NONE	No plot
PREDPPLOT	Predicted probability

The following subsections provide information about these graphs.

ODS Graph Names

PROC PROBIT assigns a name to each graph it creates using ODS. You can use these names to reference the graphs when using ODS. The names are listed in Table 75.42.

Table 75.42 Graphs Produced by PROC PROBIT

ODS Graph Name	Plot Description	Statement	PLOTS= Option
CDFPlot	Estimated cumulative probability	PROC	CDFPLOT
IPPPlot	Inverse predicted probability	PROC	IPPPLOT
LPredPlot	Linear predictor	PROC	LPREDPLOT
PredPPlot	Predicted probability	PROC	PREDPPLOT

CDF Plot

For a multinomial model, the predicted cumulative distribution function is defined as

$$\hat{F}_j(\mathbf{x}) = C + (1 - C)F(\hat{a}_j + \mathbf{x}'\hat{\mathbf{b}})$$

where $j = 1, \ldots, k$ are the indexes of the k levels of the multinomial response variable, F is the CDF of the distribution used to model the cumulative probabilities, $\hat{\mathbf{b}}$ is the vector of estimated parameters, $\mathbf{x}$ is the covariate vector, $\hat{a}_j$ are estimated ordinal intercepts with $\hat{a}_1 = 0$, and C is the threshold parameter, either known or estimated from the model. Let x_1 be the covariate corresponding to the dose variable and $\mathbf{x}_{-1}$ be the vector of the rest of the covariates. Let the corresponding estimated parameters be $\hat{b}_1$ and $\hat{\mathbf{b}}_{-1}$. Then

$$\hat{F}_j(\mathbf{x}) = C + (1 - C)F(\hat{a}_j + x_1\hat{b}_1 + \mathbf{x}'_{-1}\hat{\mathbf{b}}_{-1})$$

To plot $\hat{F}_j$ as a function of x_1, $\mathbf{x}_{-1}$ must be specified. You can use the XDATA= option to provide the values of $\mathbf{x}_{-1}$ (see the XDATA= option in the PROC PROBIT statement for details), or use the default values that follow these rules:

- If the effect contains a continuous variable (or variables), the overall mean of this effect is used.

- If the effect is a single classification variable, the highest level of the variable is used.

The LEVEL= suboption specify the levels of the multinomial response variable for which the CDF curves are requested. There are $k - 1$ curves for a k-level multinomial response variable (for the highest level, it is the constant line 1). You can specify any of them to be plotted by the LEVEL= suboption. See the plot in Output 75.2.6 for an example.

Inverse Predicted Probability Plot

For the binomial model, the response variable is a probability. An estimate of the dose level $\hat{x}_1$ needed for a response of p is given by

$$\hat{x}_1 = (F^{-1}(p) - \mathbf{x}'_{-1}\hat{\mathbf{b}}_{-1})/\hat{b}_1$$

where F is the cumulative distribution function used to model the probability, $\mathbf{x}_{-1}$ is the vector of the rest of the covariates, $\hat{\mathbf{b}}_{-1}$ is the vector of the estimated parameters corresponding to $\mathbf{x}_{-1}$, and $\hat{b}_1$ is the estimated parameter for the dose variable of interest.

To plot $\hat{x}_1$ as a function of p, $\mathbf{x}_{-1}$ must be specified. You can use the XDATA= option to provide the values of $\mathbf{x}_{-1}$ (see the XDATA= option in the PROC PROBIT statement for details), or use the default values that follow these rules:

- If the effect contains a continuous variable (or variables), the overall mean of this effect is used.

- If the effect is a single classification variable, the highest level of the variable is used.

Output 75.4.12 in Example 75.4 shows an inverse predicted probability plot.

Linear Predictor Plot

For both binomial models and multinomial models, the linear predictor $\mathbf{x}'\mathbf{b}$ can be plotted against the first single continuous variable (dose variable) in the MODEL statement.

Let x_1 be the covariate of the dose variable, $\mathbf{x}_{-1}$ be the vector of the rest of the covariates, $\hat{\mathbf{b}}_{-1}$ be the vector of estimated parameters corresponding to $\mathbf{x}_{-1}$, and $\hat{b}_1$ be the estimated parameter for the dose variable of interest.

To plot $\hat{\mathbf{x}}'\mathbf{b}$ as a function of x_1, $\mathbf{x}_{-1}$ must be specified. You can use the XDATA= option to provide the values of $\mathbf{x}_{-1}$ (see the XDATA= option in the PROC PROBIT statement for details), or use the default values that follow these rules:

- If the effect contains a continuous variable (or variables), the overall mean of this effect is used.

- If the effect is a single classification variable, the highest level of the variable is used.

For the multinomial model, you can use the LEVEL= suboption to specify the levels for which the linear predictor lines are plotted.

The confidence limits for the predicted values are only available for the binomial model. Output 75.4.13 in Example 75.4 shows a linear predictor plot for a binomial model.

Predicted Probability Plot

The predicted probability is

$$\hat{p} = C + (1 - C)F(\mathbf{x}'\hat{\mathbf{b}})$$

for the binomial model and

$$
\begin{aligned}
\hat{p}_1 &= C + (1 - C)F(\mathbf{x}'\hat{\mathbf{b}}) \\
\hat{p}_j &= (1 - C)(F(\hat{a}_j + \mathbf{x}'\hat{\mathbf{b}}) - F(\hat{a}_{j-1} + \mathbf{x}'\hat{\mathbf{b}})), \quad j = 2, \ldots, k-1 \\
\hat{p}_k &= (1 - C)(1 - F(\hat{a}_{k-1} + \mathbf{x}'\hat{\mathbf{b}}))
\end{aligned}
$$

for the multinomial model with k response levels, where F is the cumulative distribution function used to model the probability, $\mathbf{x}'$ is the vector of the covariates, $\hat{a}_j$ are the estimated ordinal intercepts with $\hat{a}_1 = 0$, C is the threshold parameter, either known or estimated from the model, and $\hat{\mathbf{b}}'$ is the vector of estimated parameters.

To plot $\hat{p}$ (or $\hat{p}_j$) as a function of a continuous variable x_1, the remaining covariates $\mathbf{x}_{-1}$ must be specified. You can use the XDATA= option to provide the values of $\mathbf{x}_{-1}$ (see the XDATA= option in the PROC PROBIT statement for details), or use the default values that follow these rules:

- If the effect contains a continuous variable (or variables), the overall mean of this effect is used.

- If the effect is a single classification variable, the highest level of the variable is used.

For the multinomial model, you can use the LEVEL= suboption to specify the levels for which the linear predictor lines are plotted.

Confidence limits are plotted only for the binomial model. Output 75.1.7 in Example 75.1 shows a predicted probability plot for a binomial model; and Output 75.2.3 in Example 75.2 shows a predicted probability plot for a multinomial model.

Examples: PROBIT Procedure

Example 75.1: Dosage Levels

In this example, Dose is a variable representing the level of a stimulus, N represents the number of subjects tested at each level of the stimulus, and Response is the number of subjects responding to that level of the stimulus. Both probit and logit response models are fit to the data. The LOG10 option in the PROC PROBIT statement requests that the log base 10 of Dose is used as the independent variable. Specifically, for a given level of Dose, the probability p of a positive response is modeled as

$$p = \text{Pr}(\text{Response}) = F\left(b_0 + b_1 \times \log_{10}(\text{Dose})\right)$$

The probabilities are estimated first by using the normal distribution function (the default) and then by using the logistic distribution function. Note that, in this model specification, the natural rate is assumed to be zero.

The LACKFIT option specifies lack-of-fit tests and the INVERSECL option specifies inverse confidence limits.

In the DATA step that reads the data, a number of observations are generated that have a missing value for the response. Although the PROBIT procedure does not use the observations with the missing values to fit the model, it does give predicted values for all nonmissing sets of independent variables. These data points fill in the plot of fitted and observed values in the logistic model displayed in Output 75.1.7. The plot, requested with the PLOT=PREDPPLOT option, displays the estimated logistic cumulative distribution function and the observed response rates.

The following statements produce Output 75.1.1:

```
data a;
   infile cards eof=eof;
   input Dose N Response @@;
   Observed= Response/N;
   output;
   return;
eof: do Dose=0.5 to 7.5 by 0.25;
      output;
   end;
   datalines;
1 10 1  2 12 2  3 10 4  4 10 5
5 12 8  6 10 8  7 10 10
;


ods graphics on;


proc probit log10;
   model Response/N=Dose / lackfit inversecl itprint;
   output out=B p=Prob std=std xbeta=xbeta;
run;
```

Output 75.1.1 Probit Analysis with Normal Distribution

```
                         The Probit Procedure

                  Iteration History for Parameter Estimates

       Iter    Ridge    Loglikelihood      Intercept      Log10(Dose)

          0       0        -51.292891               0               0
          1       0        -37.881166    -1.355817008     2.635206083
          2       0        -37.286169    -1.764939171    3.3408954936
          3       0        -37.280389    -1.812147863    3.4172391614
          4       0        -37.280388    -1.812704962     3.418117919
          5       0        -37.280388    -1.812704962     3.418117919
```

Output 75.1.1 *continued*

```
                          Model Information

            Data Set                       WORK.A
            Events Variable                Response
            Trials Variable                       N
            Number of Observations                7
            Number of Events                     38
            Number of Trials                     74
            Name of Distribution             Normal
            Log Likelihood            -37.28038802


        Last Evaluation of the Negative of the Gradient

                       Intercept     Log10(Dose)

                    3.4349069E-7    -2.09809E-8


        Last Evaluation of the Negative of the Hessian

                             Intercept     Log10(Dose)

          Intercept        36.005280383   20.152675982
          Log10(Dose)      20.152675982   13.078826305


                      Goodness-of-Fit Tests

    Statistic                    Value      DF    Value/DF    Pr > ChiSq

    Pearson Chi-Square          3.6497       5     0.7299        0.6009
    L.R.    Chi-Square          4.6381       5     0.9276        0.4616


                    Response-Covariate Profile

                 Response Levels                2
                 Number of Covariate Values     7


        Analysis of Maximum Likelihood Parameter Estimates

                         Standard    95% Confidence       Chi-
      Parameter   DF Estimate   Error       Limits       Square Pr > ChiSq

      Intercept    1  -1.8127   0.4493 -2.6934  -0.9320    16.27    <.0001
      Log10(Dose)  1   3.4181   0.7455  1.9569   4.8794    21.02    <.0001


            Probit Model in Terms of Tolerance Distribution

                          MU           SIGMA

                     0.53032254     0.29255866
```

Output 75.1.1 *continued*

```
                Estimated Covariance Matrix
                  for Tolerance Parameters


                           MU                SIGMA


       MU              0.002418           -0.000409
       SIGMA          -0.000409            0.004072
```

The *p*-values in the goodness-of-fit table of 0.6009 for the Pearson's chi-square and 0.4616 for the likelihood ratio chi-square indicate an adequate fit for the model fit with the normal distribution.

Tolerance distribution parameter estimates for the normal distribution indicate a mean tolerance for the population of 0.5303.

Output 75.1.2 displays probit analysis with the logarithm of dose levels. The LD50 (ED50 for log dose) is 0.5303, the dose corresponding to a probability of 0.5. This is the same as the mean tolerance for the normal distribution.

Output 75.1.2 Probit Analysis with Normal Distribution

```
                    The Probit Procedure

               Probit Analysis on Log10(Dose)

    Probability          Log10(Dose)       95% Fiducial Limits

        0.01             -0.15027       -0.69518        0.07710
        0.02             -0.07052       -0.55766        0.13475
        0.03             -0.01992       -0.47064        0.17156
        0.04              0.01814       -0.40534        0.19941
        0.05              0.04911       -0.35233        0.22218
        0.06              0.07546       -0.30731        0.24165
        0.07              0.09857       -0.26793        0.25881
        0.08              0.11926       -0.23273        0.27425
        0.09              0.13807       -0.20080        0.28837
        0.10              0.15539       -0.17147        0.30142
        0.15              0.22710       -0.05086        0.35631
        0.20              0.28410        0.04369        0.40124
        0.25              0.33299        0.12343        0.44116
        0.30              0.37690        0.19348        0.47857
        0.35              0.41759        0.25658        0.51504
        0.40              0.45620        0.31429        0.55182
        0.45              0.49356        0.36754        0.58999
        0.50              0.53032        0.41693        0.63057
        0.55              0.56709        0.46296        0.67451
        0.60              0.60444        0.50618        0.72271
        0.65              0.64305        0.54734        0.77603
        0.70              0.68374        0.58745        0.83550
        0.75              0.72765        0.62776        0.90265
        0.80              0.77655        0.66999        0.98008
        0.85              0.83354        0.71675        1.07279
        0.90              0.90525        0.77313        1.19191
        0.91              0.92257        0.78646        1.22098
        0.92              0.94139        0.80083        1.25265
        0.93              0.96208        0.81653        1.28759
        0.94              0.98519        0.83394        1.32672
        0.95              1.01154        0.85367        1.37149
        0.96              1.04250        0.87669        1.42424
        0.97              1.08056        0.90480        1.48928
        0.98              1.13116        0.94189        1.57602
        0.99              1.21092        0.99987        1.71321
```

Output 75.1.3 displays probit analysis with dose levels. The ED50 for dose is 3.39 with a 95% confidence interval of (2.61, 4.27).

Output 75.1.3 Probit Analysis with Normal Distribution

```
                      The Probit Procedure

                    Probit Analysis on Dose

        Probability        Dose       95% Fiducial Limits

            0.01         0.70750      0.20175       1.19427
            0.02         0.85012      0.27691       1.36380
            0.03         0.95517      0.33834       1.48444
            0.04         1.04266      0.39324       1.58274
            0.05         1.11971      0.44429       1.66793
            0.06         1.18976      0.49282       1.74443
            0.07         1.25478      0.53960       1.81473
            0.08         1.31600      0.58515       1.88042
            0.09         1.37427      0.62980       1.94252
            0.10         1.43019      0.67380       2.00181
            0.15         1.68696      0.88950       2.27147
            0.20         1.92353      1.10584       2.51906
            0.25         2.15276      1.32870       2.76161
            0.30         2.38180      1.56128       3.01000
            0.35         2.61573      1.80543       3.27374
            0.40         2.85893      2.06200       3.56306
            0.45         3.11573      2.33098       3.89038
            0.50         3.39096      2.61175       4.27138
            0.55         3.69051      2.90374       4.72619
            0.60         4.02199      3.20759       5.28090
            0.65         4.39594      3.52651       5.97077
            0.70         4.82770      3.86765       6.84706
            0.75         5.34134      4.24385       7.99189
            0.80         5.97787      4.67724       9.55169
            0.85         6.81617      5.20900      11.82480
            0.90         8.03992      5.93105      15.55653
            0.91         8.36704      6.11584      16.63320
            0.92         8.73752      6.32165      17.89163
            0.93         9.16385      6.55431      19.39034
            0.94         9.66463      6.82245      21.21881
            0.95        10.26925      7.13949      23.52275
            0.96        11.02811      7.52816      26.56066
            0.97        12.03830      8.03149      30.85201
            0.98        13.52585      8.74763      37.67206
            0.99        16.25233      9.99709      51.66627
```

The following statements request probit analysis of dosage levels with the logistic distribution:

```
proc probit log10 plot=predpplot;
   model Response/N=Dose / d=logistic inversecl;
   output out=B p=Prob std=std xbeta=xbeta;
run;
```

The regression parameter estimates in Output 75.1.4 for the logistic model of –3.22 and 5.97 are approximately $\pi/\sqrt{3}$ times as large as those for the normal model.

Output 75.1.4 Probit Analysis with Logistic Distribution

```
                    The Probit Procedure

                    Model Information

           Data Set                      WORK.B
           Events Variable              Response
           Trials Variable                     N
           Number of Observations              7
           Number of Events                   38
           Number of Trials                   74
           Name of Distribution         Logistic
           Log Likelihood          -37.11065336


       Analysis of Maximum Likelihood Parameter Estimates

                      Standard   95% Confidence     Chi-
   Parameter    DF Estimate   Error       Limits      Square Pr > ChiSq

   Intercept     1  -3.2246   0.8861  -4.9613  -1.4880   13.24    0.0003
   Log10(Dose)   1   5.9702   1.4492   3.1299   8.8105   16.97    <.0001
```

Output 75.1.5 and Output 75.1.6 show that both the ED50 and the LD50 are similar to those for the normal model.

Output 75.1.5 Probit Analysis with Logistic Distribution

```
                    The Probit Procedure

                 Probit Analysis on Log10(Dose)

      Probability        Log10(Dose)        95% Fiducial Limits

         0.01             -0.22955        -0.97441       0.04234
         0.02             -0.11175        -0.75158       0.12404
         0.03             -0.04212        -0.62018       0.17265
         0.04              0.00780        -0.52618       0.20771
         0.05              0.04693        -0.45265       0.23533
         0.06              0.07925        -0.39205       0.25826
         0.07              0.10686        -0.34037       0.27796
         0.08              0.13103        -0.29521       0.29530
         0.09              0.15259        -0.25502       0.31085
         0.10              0.17209        -0.21875       0.32498
         0.15              0.24958        -0.07552       0.38207
         0.20              0.30792         0.03092       0.42645
         0.25              0.35611         0.11742       0.46451
         0.30              0.39820         0.19143       0.49932
         0.35              0.43644         0.25684       0.53275
         0.40              0.47221         0.31588       0.56619
         0.45              0.50651         0.36986       0.60089
         0.50              0.54013         0.41957       0.63807
         0.55              0.57374         0.46559       0.67894
         0.60              0.60804         0.50846       0.72474
         0.65              0.64381         0.54896       0.77673
         0.70              0.68205         0.58815       0.83637
         0.75              0.72414         0.62752       0.90582
         0.80              0.77233         0.66915       0.98876
         0.85              0.83067         0.71631       1.09242
         0.90              0.90816         0.77562       1.23343
         0.91              0.92766         0.79014       1.26931
         0.92              0.94922         0.80607       1.30912
         0.93              0.97339         0.82378       1.35391
         0.94              1.00100         0.84384       1.40523
         0.95              1.03332         0.86713       1.46546
         0.96              1.07245         0.89511       1.53864
         0.97              1.12237         0.93053       1.63228
         0.98              1.19200         0.97952       1.76329
         0.99              1.30980         1.06166       1.98569
```

Output 75.1.6 Probit Analysis with Logistic Distribution

The Probit Procedure

Probit Analysis on Dose

Probability	Dose	95% Fiducial Limits	
0.01	0.58945	0.10607	1.10241
0.02	0.77312	0.17718	1.33058
0.03	0.90757	0.23978	1.48817
0.04	1.01813	0.29773	1.61327
0.05	1.11413	0.35266	1.71922
0.06	1.20018	0.40546	1.81244
0.07	1.27896	0.45670	1.89654
0.08	1.35218	0.50675	1.97379
0.09	1.42100	0.55588	2.04572
0.10	1.48625	0.60430	2.11339
0.15	1.77656	0.84038	2.41030
0.20	2.03199	1.07379	2.66961
0.25	2.27043	1.31046	2.91416
0.30	2.50152	1.55393	3.15736
0.35	2.73172	1.80652	3.40996
0.40	2.96627	2.06957	3.68292
0.45	3.21006	2.34345	3.98927
0.50	3.46837	2.62768	4.34578
0.55	3.74746	2.92138	4.77466
0.60	4.05546	3.22451	5.30573
0.65	4.40366	3.53961	5.98041
0.70	4.80891	3.87391	6.86079
0.75	5.29836	4.24155	8.05044
0.80	5.92009	4.66820	9.74455
0.85	6.77126	5.20365	12.37149
0.90	8.09391	5.96508	17.11715
0.91	8.46559	6.16800	18.59129
0.92	8.89644	6.39837	20.37592
0.93	9.40575	6.66469	22.58957
0.94	10.02317	6.97977	25.42292
0.95	10.79732	7.36428	29.20549
0.96	11.81534	7.85438	34.56521
0.97	13.25466	8.52173	42.88232
0.98	15.55972	9.53941	57.98207
0.99	20.40815	11.52549	96.75820

The PLOT=PREDPPLOT option together with the ODS GRAPHICS statement creates the plot of observed and fitted probabilities in Output 75.1.7. The dashed line represent pointwise confidence bands for the probabilities.

Output 75.1.7 Plot of Observed and Fitted Probabilities
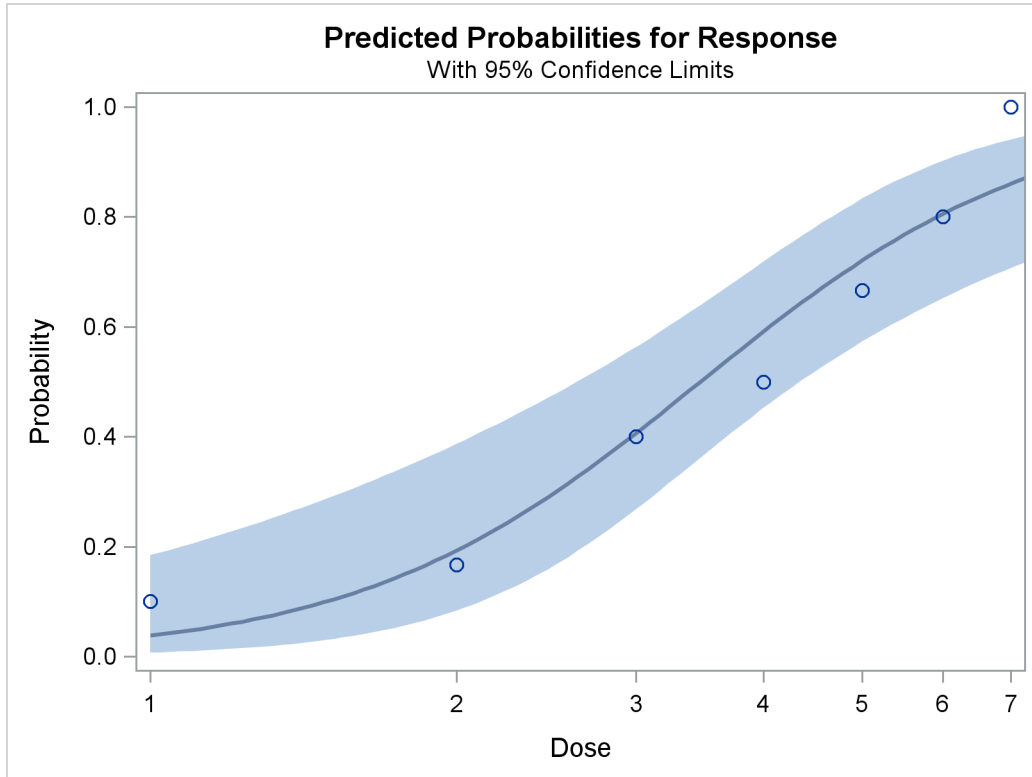
Example 75.2: Multilevel Response

In this example, two preparations, a standard preparation and a test preparation, are each given at several dose levels to groups of insects. The symptoms are recorded for each insect within each group, and two multilevel probit models are fit. Because the natural sort order of the three levels is not the same as the response order, the ORDER=DATA option is specified in the PROC PROBIT statement to get the desired order. The following statements fit two models:

```
data multi;
   input Prep $ Dose Symptoms $ N;
   LDose=log10(Dose);
   if Prep='test' then PrepDose=LDose;
   else PrepDose=0;
   datalines;
stand     10      None       33
stand     10      Mild        7
stand     10      Severe     10
stand     20      None       17
stand     20      Mild       13
stand     20      Severe     17
stand     30      None       14
stand     30      Mild        3
stand     30      Severe     28
stand     40      None        9
stand     40      Mild        8
stand     40      Severe     32
test      10      None       44
test      10      Mild        6
test      10      Severe      0
test      20      None       32
test      20      Mild       10
test      20      Severe     12
test      30      None       23
test      30      Mild        7
test      30      Severe     21
test      40      None       16
test      40      Mild        6
test      40      Severe     19
;

proc probit order=data data=multi;
   class Prep Symptoms;
   nonpara: model Symptoms=Prep LDose PrepDose / lackfit;
   weight N;
run;

proc probit order=data data=multi;
   class Prep Symptoms;
   parallel: model Symptoms=Prep LDose / lackfit;
   weight N;
run;
```

Results of these two models are shown in Output 75.2.1 and Output 75.2.2. The first model allows for nonparallelism between the dose response curves for the two preparations by inclusion of an interaction between Prep and LDose. The interaction term is labeled PrepDose in the "Analysis of Parameter Estimates" table. The results of this first model indicate that the parameter for the interaction term is not significant, having a Wald chi-square of 0.73. Also, since the first model is a generalization of the second, a likelihood ratio test statistic for this same parameter can be obtained by multiplying the difference in log likelihoods between the two models by 2. The value obtained, $2 \times (-345.94 - (-346.31))$, is 0.73. This is in close agreement with the Wald chi-square from the first model. The lack-of-fit test statistics for the two models do not indicate a problem with either fit.

Output 75.2.1 Multilevel Response: Nonparallel Analysis

```
                        The Probit Procedure

                        Model Information

                Data Set                    WORK.MULTI
                Dependent Variable           Symptoms
                Weight Variable                     N
                Number of Observations             23
                Name of Distribution           Normal
                Log Likelihood          -345.9401767


                    Class Level Information

            Name            Levels     Values

            Prep               2       stand test
            Symptoms           3       None Mild Severe


        Analysis of Maximum Likelihood Parameter Estimates
```

Parameter		DF	Estimate	Standard Error	95% Confidence Limits		Chi-Square	Pr > ChiSq
Intercept		1	3.8080	0.6252	2.5827	5.0333	37.10	<.0001
Intercept2		1	0.4684	0.0559	0.3589	0.5780	70.19	<.0001
Prep	stand	1	-1.2573	0.8190	-2.8624	0.3479	2.36	0.1247
Prep	test	0	0.0000	.	.	.	.	.
LDose		1	-2.1512	0.3909	-2.9173	-1.3851	30.29	<.0001
PrepDose		1	-0.5072	0.5945	-1.6724	0.6580	0.73	0.3935

Output 75.2.2 Multilevel Response: Parallel Analysis

```
                        The Probit Procedure

                          Model Information

              Data Set                        WORK.MULTI
              Dependent Variable                 Symptoms
              Weight Variable                            N
              Number of Observations                    23
              Name of Distribution                  Normal
              Log Likelihood                  -346.306141


                       Class Level Information

           Name              Levels     Values

           Prep                  2      stand test
           Symptoms              3      None Mild Severe


            Analysis of Maximum Likelihood Parameter Estimates

                               Standard   95% Confidence    Chi-
      Parameter          DF Estimate   Error       Limits       Square Pr > ChiSq

      Intercept           1   3.4148   0.4126   2.6061    4.2235   68.50    <.0001
      Intercept2          1   0.4678   0.0558   0.3584    0.5772   70.19    <.0001
      Prep       stand    1  -0.5675   0.1259  -0.8142   -0.3208   20.33    <.0001
      Prep       test     0   0.0000      .        .         .        .         .
      LDose               1  -2.3721   0.2949  -2.9502   -1.7940   64.68    <.0001
```

The negative coefficient associated with LDose indicates that the probability of having no symptoms (Symptoms='None') or no or mild symptoms (Symptoms='None' or Symptoms='Mild') decreases as LDose increases; that is, the probability of a severe symptom increases with LDose. This association is apparent for both treatment groups.

The negative coefficient associated with the standard treatment group (Prep = stand) indicates that the standard treatment is associated with more severe symptoms across all Ldose values.

The following statements use the PLOTS= option to create the plot shown in Output 75.2.3 and Output 75.2.4. Output 75.2.3 is the plot of the probabilities of the response taking on individual levels as a function of LDose. Since there are two covariates, LDose and Prep, the value of the classification variable Prep is fixed at the highest level, test. Instead of individual response level probabilities, the CDFPLOT option creates the plot of the cumulative response probabilities with confidence limits shown in Output 75.2.4.

```
proc probit data=multi order=data
            plots=(predpplot(level=("None" "Mild" "Severe"))
                   cdfplot(level=("None" "Mild" "Severe")));
   class Prep Symptoms;
   parallel: model Symptoms=Prep LDose / lackfit;
   weight N;
run;
```

Output 75.2.3 Plot of Predicted Probabilities for the Test Preparation Group
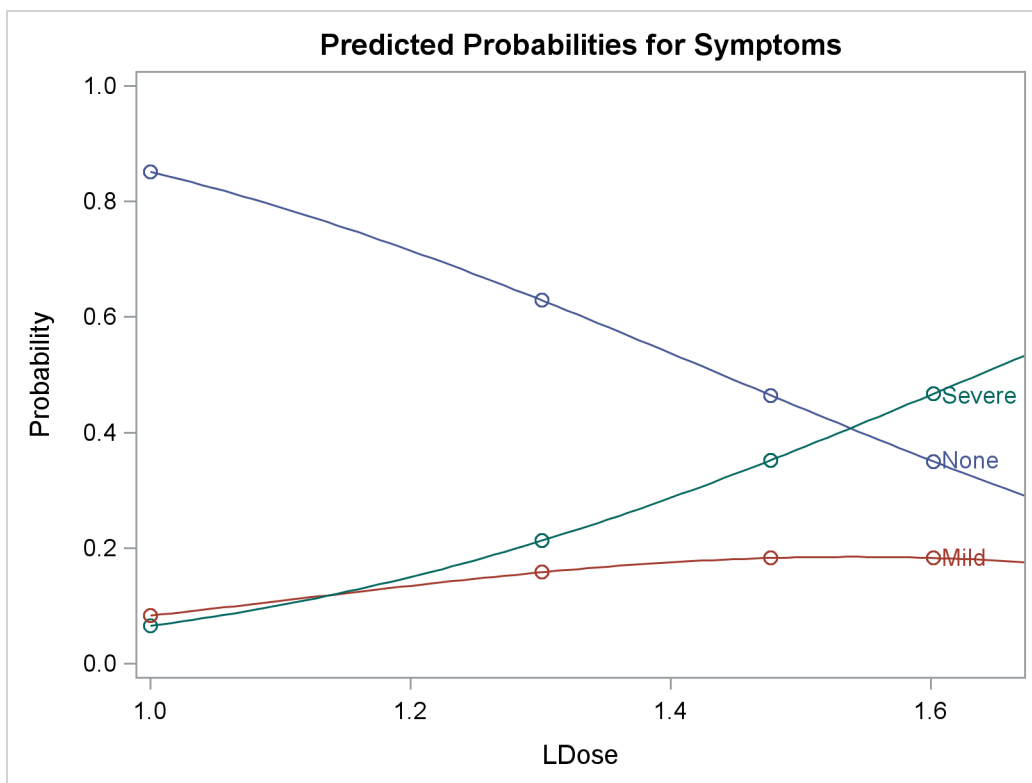
Output 75.2.4 Plot of Predicted Cumulative Probabilities for the Test Preparation Group
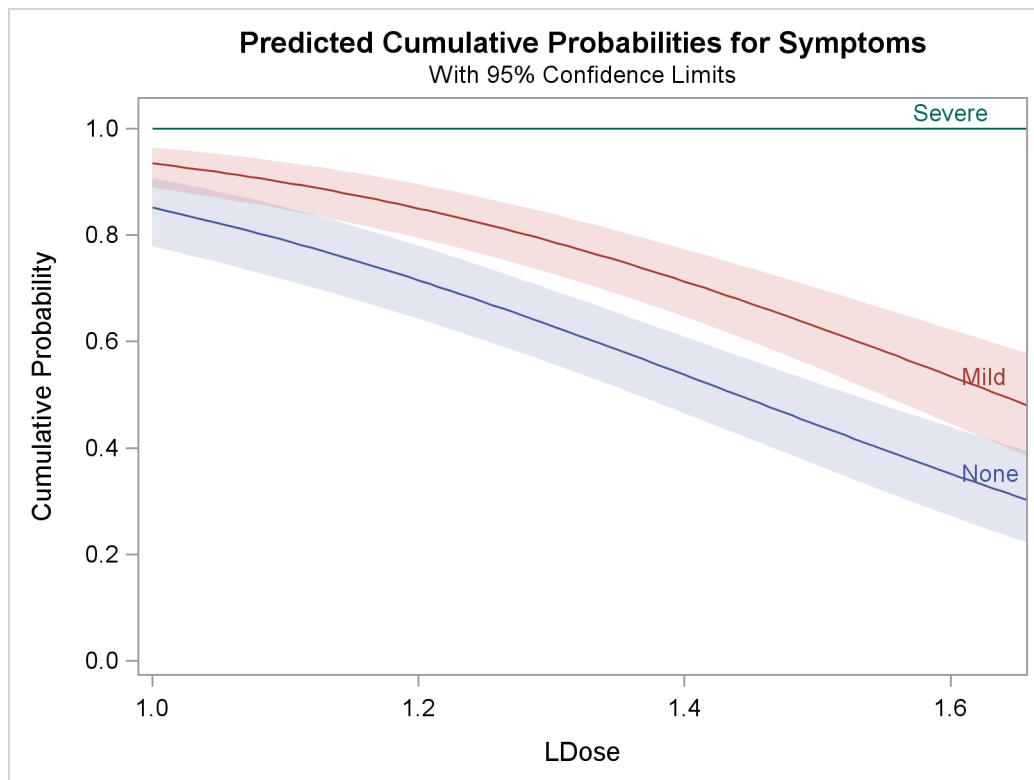
The following statements use the XDATA= data set to create plots of predicted probabilities and cumulative probabilities with Prep set to the stand level. The resulting plots are shown in Output 75.2.5 and Output 75.2.6.

```
data xrow;
   input Prep $ Dose Symptoms $ N;
   LDose=log10(Dose);
   datalines;
stand     40      Severe      32
run;

proc probit data=multi order=data xdata=xrow
            plots=(predpplot(level=("None" "Mild" "Severe"))
                   cdfplot(level=("None" "Mild" "Severe")));
   class Prep Symptoms;
   parallel: model Symptoms=Prep LDose / lackfit;
   weight N;
run;
```

Output 75.2.5 Plot of Predicted Probabilities for the Standard Preparation Group
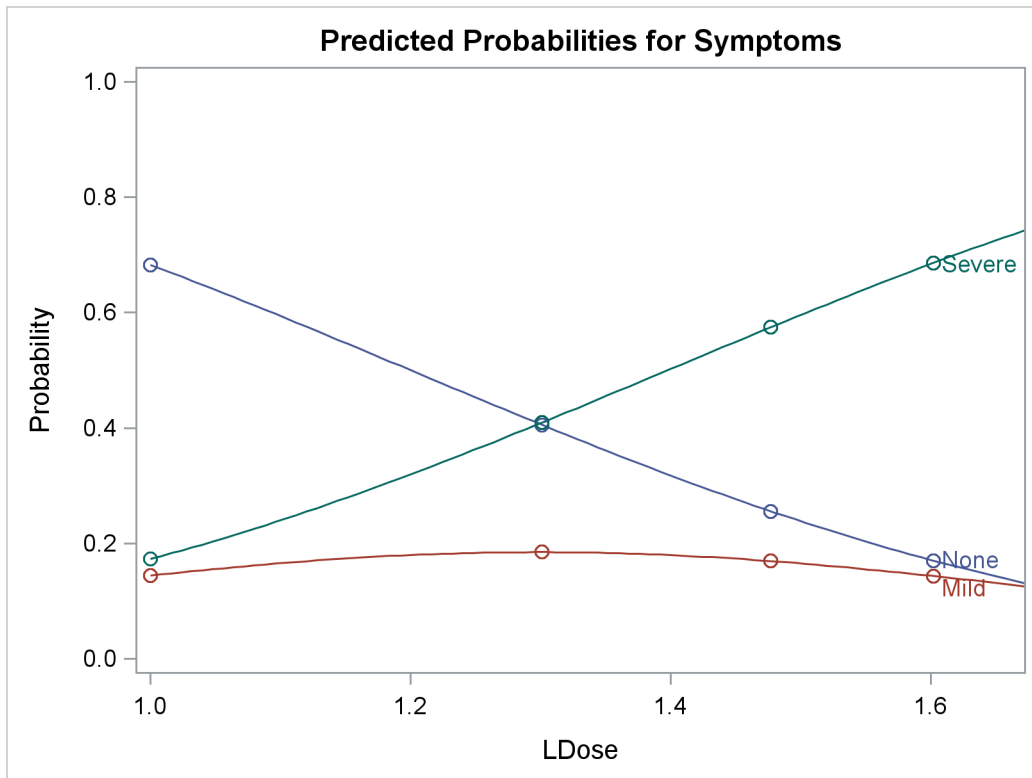
Output 75.2.6 Plot of Predicted Cumulative Probabilities for the Standard Preparation Group
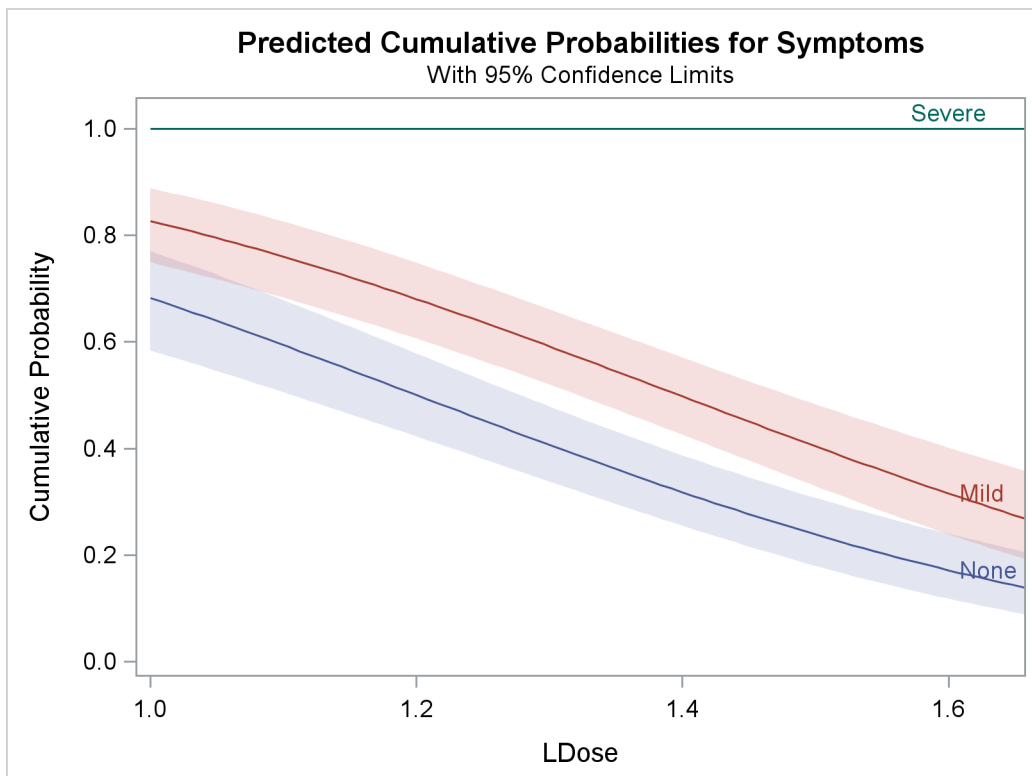
Example 75.3: Logistic Regression

In this example, a series of people are asked whether or not they would subscribe to a new newspaper. For each person, the variables sex (Female, Male), age, and subs (1=yes,0=no) are recorded. The PROBIT procedure is used to fit a logistic regression model to the probability of a positive response (subscribing) as a function of the variables sex and age. Specifically, the probability of subscribing is modeled as

$$p = \Pr(\text{subs} = 1) = F\left(b_0 + b_1 \times \text{sex} + b_2 \times \text{age}\right)$$

where F is the cumulative logistic distribution function.

By default, the PROBIT procedure models the probability of the lower response level for binary data. One way to model $\Pr(\text{subs} = 1)$ is to format the response variable so that the formatted value corresponding to subs=1 is the lower level. The following statements format the values of subs as 1 = 'accept' and 0 = 'reject', so that PROBIT models $\Pr(\text{accept}) = \Pr(\text{subs} = 1)$. They produce Output 75.3.1.

```
data news;
   input sex $ age subs @@;
   datalines;
Female     35   0   Male       44   0
Male       45   1   Female     47   1
Female     51   0   Female     47   0
Male       54   1   Male       47   1
Female     35   0   Female     34   0
Female     48   0   Female     56   1
Male       46   1   Female     59   1
Female     46   1   Male       59   1
Male       38   1   Female     39   0
Male       49   1   Male       42   1
Male       50   1   Female     45   0
Female     47   0   Female     30   1
Female     39   0   Female     51   0
Female     45   0   Female     43   1
Male       39   1   Male       31   0
Female     39   0   Male       34   0
Female     52   1   Female     46   0
Male       58   1   Female     50   1
Female     32   0   Female     52   1
Female     35   0   Female     51   0
;

proc format;
   value subscrib 1 = 'accept' 0 = 'reject';
run;

proc probit data=news;
   class subs sex;
   model subs=sex age / d=logistic itprint;
   format subs subscrib.;
run;
```

Output 75.3.1 Logistic Regression of Subscription Status

```
                        The Probit Procedure

                 Iteration History for Parameter Estimates

  Iter    Ridge    Loglikelihood       Intercept       sexFemale              age

    0        0      -27.725887               0               0                  0
    1        0      -20.142659     -3.634567629    -1.648455751      0.1051634384
    2        0       -19.52245     -5.254865196    -2.234724956      0.1506493473
    3        0      -19.490439     -5.728485385    -2.409827238      0.1639621828
    4        0      -19.490303      -5.76187293    -2.422349862      0.1649007124
    5        0      -19.490303       -5.7620267    -2.422407743      0.1649050312
    6        0      -19.490303       -5.7620267    -2.422407743      0.1649050312


                          Model Information

            Data Set                       WORK.NEWS
            Dependent Variable                  subs
            Number of Observations                40
            Name of Distribution            Logistic
            Log Likelihood               -19.49030281


                      Class Level Information

             Name       Levels     Values

             subs           2       accept reject
             sex            2       Female Male


         Last Evaluation of the Negative of the Gradient

              Intercept       sexFemale              age

         -5.95557E-12   8.768324E-10     -1.6367E-8


         Last Evaluation of the Negative of the Hessian

                    Intercept       sexFemale              age

      Intercept   6.4597397447   4.6042218284   292.04051848
      sexFemale   4.6042218284   4.6042218284   216.20829515
      age          292.04051848   216.20829515   13487.329973
```

Output 75.3.1 *continued*

```
        Analysis of Maximum Likelihood Parameter Estimates

                              Standard    95% Confidence    Chi-
   Parameter          DF Estimate   Error       Limits      Square Pr > ChiSq

   Intercept           1  -5.7620   2.7635 -11.1783  -0.3458   4.35    0.0371
   sex       Female    1  -2.4224   0.9559  -4.2959  -0.5489   6.42    0.0113
   sex       Male      0   0.0000      .        .        .       .        .
   age                 1   0.1649   0.0652   0.0371   0.2927   6.40    0.0114
```

Output 75.3.1 shows that there appears to be an effect due to both the variables sex and age. The positive coefficient for age indicates that older people are more likely to subscribe than younger people. The negative coefficient for sex indicates that females are less likely to subscribe than males.

Example 75.4: An Epidemiology Study

The data in this example, which are from an epidemiology study, consist of five variables: the number, r, of individuals surviving after an epidemic, out of n treated, for combinations of medicine dosage (dose), treatment (treat = A, B), and sex (sex = 0(Female), 1(Male)).

To see whether the two treatments have different effects on male and female individual survival rates, the interaction term between the two variables treat and sex is included in the model.

The following invocation of PROC PROBIT fits the binary probit model to the grouped data:

```
data epidemic;
   input treat$ dose n r sex @@;
   label dose = Dose;
   datalines;
A  2.17 142 142  0   A    .57 132   47  1
A  1.68 128 105  1   A   1.08 126 100   0
A  1.79 125 118  0   B   1.66 117 115   1
B  1.49 127 114  0   B   1.17  51   44  1
B  2.00 127 126  0   B    .80 129 100   1
;

data xval;
   input treat $ dose sex;
   datalines;
B  2.  1
;
```

```
proc probit optc lackfit covout data=epidemic
          outest = out1 xdata = xval
          Plots=(predpplot ippplot lpredplot);
   class treat sex;
   model r/n = dose treat sex sex*treat/corrb covb inversecl;
   output out = out2 p =p;
run;
```

The results of this analysis are shown in the outputs that follow.

Output 75.4.1 displays the table of level information for *all* classification variables in the CLASS statement.

Output 75.4.1 Class Level Information

```
              The Probit Procedure

            Class Level Information

      Name        Levels    Values

      treat          2      A B
      sex            2      0 1
```

Output 75.4.2 displays the table of parameter information for the effects in the MODEL statement.

Output 75.4.2 Parameter Information

```
                    Parameter Information

         Parameter      Effect        treat     sex

         Intercept      Intercept
         dose           dose
         treatA         treat         A
         treatB         treat         B
         sex0           sex                       0
         sex1           sex                       1
         treatAsex0     treat*sex     A           0
         treatAsex1     treat*sex     A           1
         treatBsex0     treat*sex     B           0
         treatBsex1     treat*sex     B           1
```

Output 75.4.3 displays background information about the model fit. Included are the name of the input data set, the response variables used, the numbers of observations, events, and trials, the type of distribution, and the final value of the log-likelihood function.

Output 75.4.3 Model Information

```
                    The Probit Procedure

                    Model Information

         Data Set                    WORK.EPIDEMIC
         Events Variable                         r
         Trials Variable                         n
         Number of Observations                 10
         Number of Events                     1011
         Number of Trials                     1204
         Name of Distribution               Normal
         Log Likelihood               -387.2467391
```

Output 75.4.4 displays the table of goodness-of-fit tests requested with the LACKFIT option in the PROC PROBIT statement. Two goodness-of-fit statistics, the Pearson's chi-square statistic and the likelihood ratio chi-square statistic, are computed. The grouping method for computing these statistics can be specified by the AGGREGATE= option. The details can be found in the AGGREGATE= option, and an example can be found in the second part of this example. By default, the PROBIT procedure uses the covariates in the MODEL statement to do grouping. Observations with the same values of the covariates in the MODEL statement are grouped into cells and the two statistics are computed according to these cells. The total number of cells and the number of levels for the response variable are reported next in the "Response-Covariate Profile."

In this example, neither the Pearson's chi-square nor the log-likelihood ratio chi-square tests are significant at the 0.1 level, which is the default test level used by the PROBIT procedure. That means that the model, which includes the interaction of treat and sex, is suitable for this epidemiology data set. (Further investigation shows that models without the interaction of treat and sex are not acceptable by either test.)

Output 75.4.4 Goodness-of-Fit Tests and Response-Covariate Profile

```
                      Goodness-of-Fit Tests

  Statistic                      Value      DF    Value/DF    Pr > ChiSq

  Pearson Chi-Square            4.9317       4     1.2329       0.2944
  L.R.    Chi-Square            5.7079       4     1.4270       0.2220


                   Response-Covariate Profile

              Response Levels                2
              Number of Covariate Values    10
```

Output 75.4.5 displays the Type III test results for all effects specified in the MODEL statement, which include the degrees of freedom for the effect, the Wald Chi-Square test statistic, and the *p*-value.

Output 75.4.5 Type III Tests

```
                  Type III Analysis of Effects

                                  Wald
          Effect          DF    Chi-Square    Pr > ChiSq

          dose             1      42.1691       <.0001
          treat            1      16.1421       <.0001
          sex              1       1.7710       0.1833
          treat*sex        1      13.9343       0.0002
```

Output 75.4.6 displays the table of parameter estimates for the model. The PROBIT procedure displays information for all the parameters of an effect. Degenerate parameters are indicated by 0 degree of freedom. Confidence intervals are computed for all parameters with nonzero degrees of freedom, including the natural threshold C if the OPTC option is specified in the PROC PROBIT statement. The confidence level can be specified by the ALPHA= option in the MODEL statement. The default confidence level is 95%.

Output 75.4.6 Analysis of Parameter Estimates

```
          Analysis of Maximum Likelihood Parameter Estimates

                              Standard    95% Confidence      Chi-
   Parameter         DF Estimate   Error       Limits       Square Pr > ChiSq

   Intercept          1  -0.8871   0.3632  -1.5991  -0.1752    5.96    0.0146
   dose               1   1.6774   0.2583   1.1711   2.1837   42.17   <.0001
   treat      A       1  -1.2537   0.2616  -1.7664  -0.7410   22.97   <.0001
   treat      B       0   0.0000      .        .        .        .       .
   sex        0       1  -0.4633   0.2289  -0.9119  -0.0147    4.10    0.0429
   sex        1       0   0.0000      .        .        .        .       .
   treat*sex A 0      1   1.2899   0.3456   0.6126   1.9672   13.93    0.0002
   treat*sex A 1      0   0.0000      .        .        .        .       .
   treat*sex B 0      0   0.0000      .        .        .        .       .
   treat*sex B 1      0   0.0000      .        .        .        .       .
   _C_                1   0.2735   0.0946   0.0881   0.4589
```

From Table 75.4.6, you can see the following results:

- The variable dose has a significant positive effect on the survival rate.

- Individuals under treatment A have a lower survival rate.

- Male individuals have a higher survival rate.

- Female individuals under treatment A have a higher survival rate.

Output 75.4.7 and Output 75.4.8 display tables of estimated covariance matrix and estimated correlation matrix for estimated parameters with a nonzero degree of freedom, respectively. They are computed by the inverse of the Hessian matrix of the estimated parameters.

Output 75.4.7 Estimated Covariance Matrix

```
                    Estimated Covariance Matrix

               Intercept         dose        treatA          sex0     treatAsex0

Intercept       0.131944    -0.087353      0.053551      0.030285     -0.067056
dose           -0.087353     0.066723     -0.047506     -0.034081      0.058620
treatA          0.053551    -0.047506      0.068425      0.036063     -0.075323
sex0            0.030285    -0.034081      0.036063      0.052383     -0.063599
treatAsex0     -0.067056     0.058620     -0.075323     -0.063599      0.119408
_C_            -0.028073     0.018196     -0.017084     -0.008088      0.019134


                    Estimated Covariance Matrix

                                   _C_

                   Intercept     -0.028073
                   dose           0.018196
                   treatA        -0.017084
                   sex0          -0.008088
                   treatAsex0     0.019134
                   _C_            0.008948
```

Output 75.4.8 Estimated Correlation Matrix

```
                    Estimated Correlation Matrix

               Intercept         dose        treatA          sex0     treatAsex0

Intercept       1.000000    -0.930998      0.563595      0.364284     -0.534227
dose           -0.930998     1.000000     -0.703083     -0.576477      0.656744
treatA          0.563595    -0.703083      1.000000      0.602359     -0.833299
sex0            0.364284    -0.576477      0.602359      1.000000     -0.804154
treatAsex0     -0.534227     0.656744     -0.833299     -0.804154      1.000000
_C_            -0.817027     0.744699     -0.690420     -0.373565      0.585364


                    Estimated Correlation Matrix

                                   _C_

                   Intercept     -0.817027
                   dose           0.744699
                   treatA        -0.690420
                   sex0          -0.373565
                   treatAsex0     0.585364
                   _C_            1.000000
```

Output 75.4.9 displays the computed values and fiducial limits for the first single continuous variable dose in the MODEL statement, given the probability levels, without the effect of the natural threshold, and when the option INSERSECL in the MODEL statement is specified. If there is no single continuous variable in the MODEL specification but the INVERSECL option is specified, an error is reported.

Output 75.4.9 Probit Analysis on Dose

```
                        The Probit Procedure

                       Probit Analysis on dose

        Probability         dose         95% Fiducial Limits

             0.01        -0.85801       -1.81301       -0.33743
             0.02        -0.69549       -1.58167       -0.21116
             0.03        -0.59238       -1.43501       -0.13093
             0.04        -0.51482       -1.32476       -0.07050
             0.05        -0.45172       -1.23513       -0.02130
             0.06        -0.39802       -1.15888        0.02063
             0.07        -0.35093       -1.09206        0.05742
             0.08        -0.30877       -1.03226        0.09039
             0.09        -0.27043       -0.97790        0.12040
             0.10        -0.23513       -0.92788        0.14805
             0.15        -0.08900       -0.72107        0.26278
             0.20         0.02714       -0.55706        0.35434
             0.25         0.12678       -0.41669        0.43322
             0.30         0.21625       -0.29095        0.50437
             0.35         0.29917       -0.17477        0.57064
             0.40         0.37785       -0.06487        0.63387
             0.45         0.45397        0.04104        0.69546
             0.50         0.52888        0.14481        0.75654
             0.55         0.60380        0.24800        0.81819
             0.60         0.67992        0.35213        0.88157
             0.65         0.75860        0.45879        0.94803
             0.70         0.84151        0.56985        1.01942
             0.75         0.93099        0.68770        1.09847
             0.80         1.03063        0.81571        1.18970
             0.85         1.14677        0.95926        1.30171
             0.90         1.29290        1.12867        1.45386
             0.91         1.32819        1.16747        1.49273
             0.92         1.36654        1.20867        1.53590
             0.93         1.40870        1.25284        1.58450
             0.94         1.45579        1.30084        1.64012
             0.95         1.50949        1.35397        1.70515
             0.96         1.57258        1.41443        1.78353
             0.97         1.65015        1.48626        1.88238
             0.98         1.75326        1.57833        2.01720
             0.99         1.91577        1.71776        2.23537
```

If the XDATA= option is used to input a data set for the independent variables in the MODEL statement, the PROBIT procedure uses these values for the independent variables other than the single continuous variable. Missing values are not permitted in the XDATA= data set for the independent variables, although the value for the single continuous variable is not used in the computing of the fiducial limits. A suitable valid value should be given. In the data set xval created by the SAS statements on page 6436, dose = 2.

Only one observation from the XDATA= data set is used to produce a probit analysis table for a combination of classification variable levels. If more than one observation is present in the XDATA= data set, only the last observation is used.

See the section "XDATA= *SAS-data-set*" on page 6411 for the default values for those effects other than the single continuous variable, for which the fiducial limits are computed.

In this example, there are two classification variables, treat and sex. Fiducial limits for the dose variable are computed for the highest level of the classification variables, treat = B and sex = 1, which is the default specification. Since these are the default values, you would get the same values and fiducial limits if you did not specify the XDATA= option in this example. The confidence level for the fiducial limits can be specified by the ALPHA= option in the MODEL statement. The default level is 95%.

If a LOG10 or LOG option is used in the PROC PROBIT statement, the values and the fiducial limits are computed for both the single continuous variable and its logarithm.

Output 75.4.10 displays the OUTEST= data set. All parameters for an effect are included. The name of a parameter is generated by combining the variable names and levels in the effect. The maximum length of a parameter name is 32.

Output 75.4.10 Outest Data Set for Epidemiology Study

Obs	_MODEL_	_NAME_	_TYPE_	_DIST_	_STATUS_	_LNLIKE_	r	Intercept
1		r	PARMS	Normal 0	Converged	-387.247	-1.00000	-0.88714
2		Intercept	COV	Normal 0	Converged	-387.247	-0.88714	0.13194
3		dose	COV	Normal 0	Converged	-387.247	1.67739	-0.08735
4		treatA	COV	Normal 0	Converged	-387.247	-1.25367	0.05355
5		treatB	COV	Normal 0	Converged	-387.247	0.00000	0.00000
6		sex0	COV	Normal 0	Converged	-387.247	-0.46329	0.03029
7		sex1	COV	Normal 0	Converged	-387.247	0.00000	0.00000
8		treatAsex0	COV	Normal 0	Converged	-387.247	1.28991	-0.06706
9		treatAsex1	COV	Normal 0	Converged	-387.247	0.00000	0.00000
10		treatBsex0	COV	Normal 0	Converged	-387.247	0.00000	0.00000
11		treatBsex1	COV	Normal 0	Converged	-387.247	0.00000	0.00000
12		_C_	COV	Normal 0	Converged	-387.247	0.27347	-0.02807

Obs	dose	treatA	treatB	sex0	sex1	treatAsex0	treatAsex1	treatBsex0	treatBsex1	_C_
1	1.67739	-1.25367	0	-0.46329	0	1.28991	0	0	0	0.27347
2	-0.08735	0.05355	0	0.03029	0	-0.06706	0	0	0	-0.02807
3	0.06672	-0.04751	0	-0.03408	0	0.05862	0	0	0	0.01820
4	-0.04751	0.06843	0	0.03606	0	-0.07532	0	0	0	-0.01708
5	0.00000	0.00000	0	0.00000	0	0.00000	0	0	0	0.00000
6	-0.03408	0.03606	0	0.05238	0	-0.06360	0	0	0	-0.00809
7	0.00000	0.00000	0	0.00000	0	0.00000	0	0	0	0.00000
8	0.05862	-0.07532	0	-0.06360	0	0.11941	0	0	0	0.01913
9	0.00000	0.00000	0	0.00000	0	0.00000	0	0	0	0.00000
10	0.00000	0.00000	0	0.00000	0	0.00000	0	0	0	0.00000
11	0.00000	0.00000	0	0.00000	0	0.00000	0	0	0	0.00000
12	0.01820	-0.01708	0	-0.00809	0	0.01913	0	0	0	0.00895

The plots in the following three outputs, Output 75.4.11, Output 75.4.12, and Output 75.4.13, are generated by the PLOTS= option. The first plot, specified with the PREDPPLOT option, is the plot of the predicted probability against the first single continuous variable dose in the MODEL statement. You can specify values of other independent variables in the MODEL statement by using an XDATA= data set or by using the default values.

The second plot, specified with the IPPPLOT option, is the inverse of the predicted probability plot with the fiducial limits. It should be pointed out that the fiducial limits are *not* just the inverse of the confidence limits in the predicted probability plot; see the section "Inverse Confidence Limits" on page 6410 for the computation of these limits. The third plot, specified with the LPREDPLOT option, is the plot of the linear predictor x'_c against the first single continuous variable with the Wald confidence intervals.

Output 75.4.11 Predicted Probability Plot
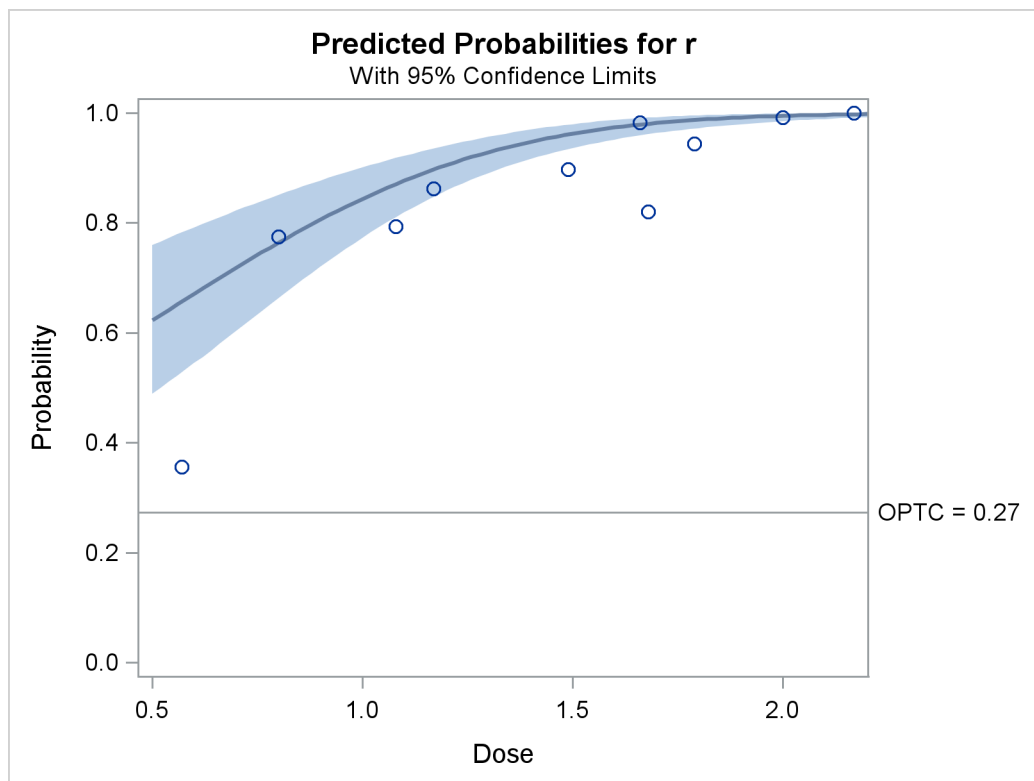
Output 75.4.12 Inverse Predicted Probability Plot
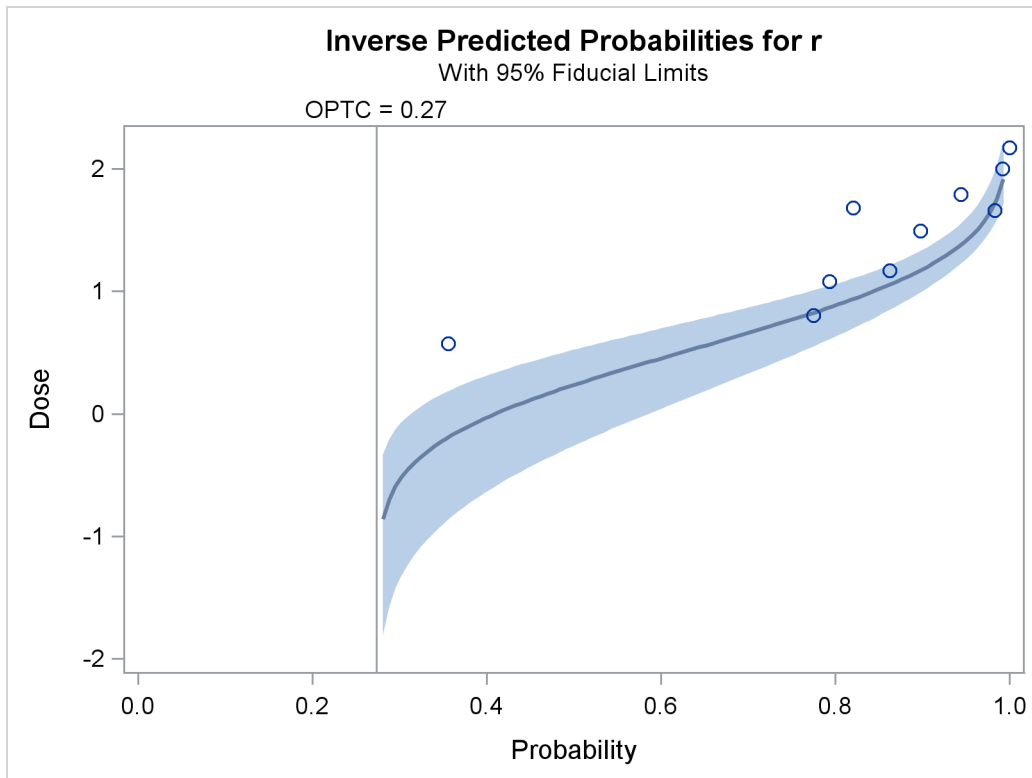
Output 75.4.13 Linear Predictor Plot
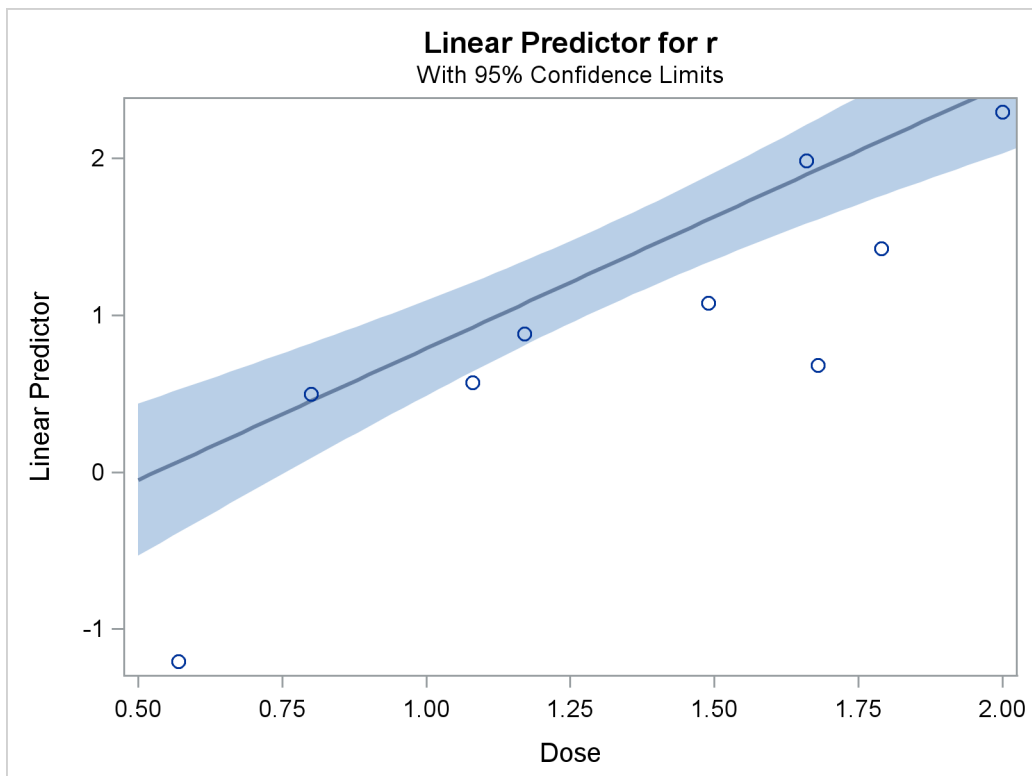
When you combine the INEST= data set and the MAXIT= option in the MODEL statement, the PROBIT procedure can do prediction, if the parameterizations for the models used for the training data and the validation data are exactly the same. The following SAS statements show an example:

```
data validate;
   input treat $ dose  sex  n r group @@;
   datalines;
B  2.0  0  44 43  1   B  2.0  1   54 52  2
B  1.5  1  36 32  3   B  1.5  0   45 40  4
A  2.0  0  66 64  5   A  2.0  1   89 89  6
A  1.5  1  45 39  7   A  1.5  0   66 60  8
B  2.0  0  44 44  1   B  2.0  1   54 54  2
B  1.5  1  36 30  3   B  1.5  0   45 41  4
A  2.0  0  66 65  5   A  2.0  1   89 88  6
A  1.5  1  45 38  7   A  1.5  0   66 59  8
;

proc probit optc data=validate inest=out1;
   class treat sex;
   model r/n = dose treat sex sex*treat  / maxit = 0;
   output out = out3 p =p;
run;

proc probit optc lackfit data=validate inest=out1;
   class treat sex;
   model r/n = dose treat sex sex*treat / aggregate = group;
   output out = out4 p =p;
run;
```

After the first invocation of PROC PROBIT, you have the estimated parameters and their covariance matrix in the data set OUTEST = Out1, and the fitted probabilities for the training data set epidemic in the data set OUTPUT = Out2. See Output 75.4.10 for the data set Out1 and Output 75.4.14 for the data set Out2.

The validation data are collected in data set validate. The second invocation of PROC PROBIT simply passes the estimated parameters from the training data set epidemic to the validation data set validate for prediction. The predicted probabilities are stored in the data set OUTPUT = Out3 (see Output 75.4.15). The third invocation of PROC PROBIT passes the estimated parameters as initial values for a new fit of the validation data set with the same model. Predicted probabilities are stored in the data set OUTPUT = Out4 (see Output 75.4.16). Goodness-of-fit tests are computed based on the cells grouped by the AGGREGATE= group variable. Results are shown in Output 75.4.17.

Output 75.4.14 Out2

Obs	treat	dose	n	r	sex	p
1	A	2.17	142	142	0	0.99272
2	A	0.57	132	47	1	0.35925
3	A	1.68	128	105	1	0.81899
4	A	1.08	126	100	0	0.77517
5	A	1.79	125	118	0	0.96682
6	B	1.66	117	115	1	0.97901
7	B	1.49	127	114	0	0.90896
8	B	1.17	51	44	1	0.89749
9	B	2.00	127	126	0	0.98364
10	B	0.80	129	100	1	0.76414

Output 75.4.15 Out3

Obs	treat	dose	sex	n	r	group	p
1	B	2.0	0	44	43	1	0.98364
2	B	2.0	1	54	52	2	0.99506
3	B	1.5	1	36	32	3	0.96247
4	B	1.5	0	45	40	4	0.91145
5	A	2.0	0	66	64	5	0.98500
6	A	2.0	1	89	89	6	0.91835
7	A	1.5	1	45	39	7	0.74300
8	A	1.5	0	66	60	8	0.91666
9	B	2.0	0	44	44	1	0.98364
10	B	2.0	1	54	54	2	0.99506
11	B	1.5	1	36	30	3	0.96247
12	B	1.5	0	45	41	4	0.91145
13	A	2.0	0	66	65	5	0.98500
14	A	2.0	1	89	88	6	0.91835
15	A	1.5	1	45	38	7	0.74300
16	A	1.5	0	66	59	8	0.91666

Output 75.4.16 Out4

Obs	treat	dose	sex	n	r	group	p
1	B	2.0	0	44	43	1	0.98954
2	B	2.0	1	54	52	2	0.98262
3	B	1.5	1	36	32	3	0.86187
4	B	1.5	0	45	40	4	0.90095
5	A	2.0	0	66	64	5	0.98768
6	A	2.0	1	89	89	6	0.98614
7	A	1.5	1	45	39	7	0.88075
8	A	1.5	0	66	60	8	0.88964
9	B	2.0	0	44	44	1	0.98954
10	B	2.0	1	54	54	2	0.98262
11	B	1.5	1	36	30	3	0.86187
12	B	1.5	0	45	41	4	0.90095
13	A	2.0	0	66	65	5	0.98768
14	A	2.0	1	89	88	6	0.98614
15	A	1.5	1	45	38	7	0.88075
16	A	1.5	0	66	59	8	0.88964

Output 75.4.17 Goodness-of-Fit Table

The Probit Procedure

Goodness-of-Fit Tests

Statistic	Value	DF	Value/DF	Pr > ChiSq
Pearson Chi-Square	2.8101	2	1.4050	0.2454
L.R. Chi-Square	2.8080	2	1.4040	0.2456

Example 75.5: Model Postfitting Analysis

Recall the previous example of an epidemic study, in which the treat*sex interaction is statistically significant. Suppose you want to know whether such an effect is the same at different levels of the two categorical variables.

The following SAS statements fit a probit model and use the SLICE statement to request analysis of the two-way interaction term treat*sex:

```
proc probit data=epidemic;
   class treat sex;
   model r/n = dose treat sex treat*sex;
   slice treat*sex / diff;
   effectplot;
run;
```

Output 75.5.1 displays the test results for the interaction effect. As you can see, the difference between the two treatments is not significant among females.

Output 75.5.1 Tests Conditional on treat*sex

```
                      The Probit Procedure

                 Chi-Square Test for treat*sex
                   Least Squares Means Slice

                       Num
            Slice       DF      Chi-Square      Pr > ChiSq

            treat A      1        18.52           <.0001


                 Chi-Square Test for treat*sex
                   Least Squares Means Slice

                       Num
            Slice       DF      Chi-Square      Pr > ChiSq

            treat B      1         2.65           0.1035


                 Chi-Square Test for treat*sex
                   Least Squares Means Slice

                       Num
            Slice       DF      Chi-Square      Pr > ChiSq

            sex 0        1         0.00           0.9579


                 Chi-Square Test for treat*sex
                   Least Squares Means Slice

                       Num
            Slice       DF      Chi-Square      Pr > ChiSq

            sex 1        1        47.43           <.0001
```

The DIFF option computes effect differences between groups within the same slice. Results are displayed in Output 75.5.2.

Output 75.5.2 Effect Differences Conditional on treat*sex

```
            Simple Differences of treat*sex Least Squares Means

                                          Standard
          Slice      sex    _sex    Estimate    Error    z Value    Pr > |z|

          treat A     0      1       0.5957     0.1384      4.30      <.0001
```

Output 75.5.2 *continued*

Simple Differences of treat*sex Least Squares Means

Slice	sex	_sex	Estimate	Standard Error	z Value	Pr > |z|
treat B	0	1	−0.2956	0.1816	−1.63	0.1035

Simple Differences of treat*sex Least Squares Means

Slice	treat	_treat	Estimate	Standard Error	z Value	Pr > |z|
sex 0	A	B	−0.00899	0.1702	−0.05	0.9579

Simple Differences of treat*sex Least Squares Means

Slice	treat	_treat	Estimate	Standard Error	z Value	Pr > |z|
sex 1	A	B	−0.9003	0.1307	−6.89	<.0001

The EFFECTPLOT statement produces a predicted probability plot for dose by the four groups that are formed by the treat*sex interaction. The plot is displayed in Output 75.5.3. The two overlapping curves represent the two treatment groups for females, suggesting no treatment effect. It appears that males tend to respond to the two treatments differently: those on treatment B have a better survival rate, and those on treatment A have a worse chance of survival.

Output 75.5.3 Predicted Probability versus Dose Level by treat*sex
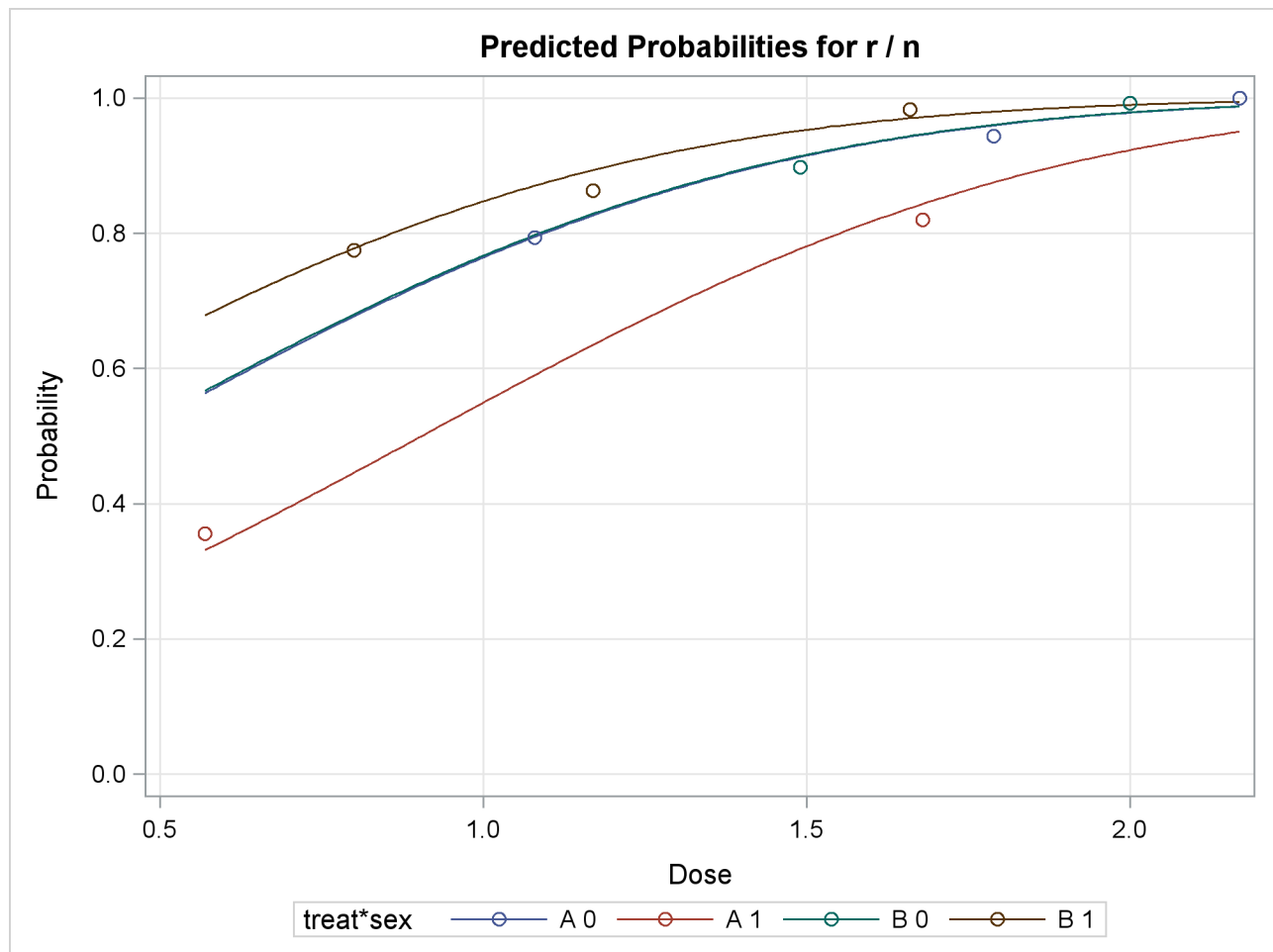
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Subject Index

Syntax Index

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