Understanding The Influence Of Day Of The Week On Reviews Written

April 8 - 11 | Denver, CO **#SASGF**

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USERS PROGRAM



Understanding The Influence Of Day Of The Week On Reviews Written Sujal Reddy Alugubelli

ABSTRACT

- Today most companies utilize customer feedback to understand customer desires.
- It is very crucial for a firm to know the key feedback and thereby providing synthesized services to their customers.
- Analysis on the effect of day of the week on the way a user reacts.
- Human beings over react when they are under pressures whereas more accepting in nature when they are calm.
- Stressed on weekdays \rightarrow React more ; Relaxed on weekends \rightarrow Accept more.
- Real reason of a review \rightarrow Is it just because of a users' bad mood that day or genuine defect in product?
- This way a company can not only attain excellent customer satisfaction but also be efficient in providing the services.

METHODOLOGY

In this research, nearly 10,000 reviews written on Amazon.com for two products from electronics product category are explored.

The process that is followed in this analysis is broadly classified into following steps:

- **Data Extraction using Python.**
- Data Preparation and initial analysis using Base SAS.
- Initial Text analysis and filtration of the text.
- Review analysis using SAS Enterprise Miner.
- User sentiment analysis using SAS Sentiment Analysis Studio.



Graduate Student, Oklahoma State University | Data Scientist Intern, INVISTA

LG Tone HBS-730 Wireless Stereo Headset – Black

For product LG Tone HBS-730 wireless stereo headset, negative reviews were written about earbuds on weekdays but positive feedback is written on weekend.

Cluster ID	Descriptive Terms
1 2 3 4 5 6 7 8 9	+volume +loud +'low s +pocket +foot +phone audio +support +issue +charge +charger +fu +model 'previous mod +box +package real + +hear +noise people +month +bud +'ear bu +quality +'sound qualit

Cluster ID	Descriptive Terms
1	+headphone +wire +r
2	+noise +volume +vol +battery +life +'batter
4	+phone +music +love +battery +day +last +
6	+love +great +recomi lg +tone +'lg tone' +h
8	+ear +bud +'ear bud'

I LOVE these headphones. The first pair I received was defective - the battery got so hot it burned my neck and it would die after 40 minutes of play. I ordered a replacement and it works beautifully. They are so comfortable I forget I am wearing them - which is exactly how I want wireless headphones to feel. The earbuds are comfortable and delivery great sound quality. The controls are easy to memorize and respond quickly. I am thrilled with this purchase.

Examples of reviews on Weekend

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RESULTS

set' +'volume control' +low +control +ios +set +iphone +music +comfortable +listen +device +button +read +signal +walk +drop +cut +bluetooth +connection +range +lose +listen +pair +keep +headphone +hbs-700 +connection +music +iphone +battery +ios +device +connect +comfortable different +second +know II +minute +replacement first +day +unit +turn +keep +hold right +work +package amazon el' previous +version +old +battery +love +range +second +'sound guality' +year +hbs-700 back +cut +issue +fake +open amazon +order +original +charger +receive +cheap +look +item different +warranty +talk money +problem +work well +item +want +stop +know +button +set back ud' +ear +warranty +break +wire +stop +last +year +piece +long +design +cheap +work ' +sound +range +horrible +bluetooth +hope +'bluetooth headset' +review +read +device +good lg +headset +tone

Clusters formed for negative reviews on Weekday

review +issue +range +problem +set +want +pair +little +fit +neck +order +buy well ce +hear +connect +device well +little +thing 'sound quality' +range +time +excellent +quality +problem y life' +good +long +guality +sound +great +comfortable 'sound guality' +amaze +range +easy +excellent +far +listen +hand +drive +talk +free cell +product able +work +great +'great product' +clear +charge +hour 'long time' +long +week +work +time +listen +wear +comfortable +'great product' +music mend +product +buy +purchase +husband +friend +gift family +order +price +'great product' +item +awesome neadset +bluetooth +look +review cell +recommend +friend +long +talk +highly +item +voice +piece +stay +neck +fit +little +hold +wear +happy +item +comfortable +thing +hear

Clusters formed for positive reviews on Weekend

5

Was satisfied with my headset until this morning I look up when my music kept going in and out to see the left side ear bud cord has broken near the connection of the headset. The earbud cords are stiff and this could have come from placing the bud in my ear. Not happy at all. I haven't even had this item 6 months.

Examples of reviews on Weekday



Frequency
476
 423
 176
 221
 339
 92
 177



Understanding The Influence Of Day Of The Week On Reviews Written Sujal Reddy Alugubelli

RESULTS CONTINUED (CLICK TO EDIT)

Motorola S305 Bluetooth Stereo Headset w/ Microphone (Black)

Volume seems to be a significant topic in negative weekday reviews. Whereas it is positive on weekends.



Terms connected to Volume and Examples on Weekend

Effect of Length of Review

- Neutral Reviews more number of words
 - Explain both positive side and negative side.
- Negative reviews more lengthy than positive reviews.

Analysis Variable : no_words						
Rate	N Obs	Mean	Std Dev	Minimum	Maximum	N
Bad	1100	94.2209091	100.2060993	2.0000000	884.0000000	1100
Good	5339	74.0513205	98.0537850	1.0000000	1460.00	5339
Neutral	620	101.7822581	124.6959313	1.0000000	1559.00	620

Com Cor Bad Bad Good Good

CONCLUSIONS

This research illustrates the influence of day of the week on the way a customer writes a review. Specifically, for the two examples considered, data is extracted, prepared and text mining is performed. Manufacturers can leverage this analysis or elements thereof, in their own analysis.

In this research behavior changes are observed using text analysis. Text cluster node and text topic node helped in finding out the significant reviews written by the customer. Text filter node helped in finding out the terms that are related to those significant topics. With this information a company can greatly synthesize its services and optimize the efforts spent on the betterment of the product.

Graduate Student, Oklahoma State University | Data Scientist Intern, INVISTA



Terms connected to Volume and Examples on Weekday

parisons significant at the 0.05 level are indicated by ***.							
Rate nparison	Difference Between Means	Simultaneous 9 Lim	5% Confidence hits				
tral - Bad	7.561	-4.329	19.452				
ral - Good	27.731	17.685	37.777	***			
- Neutral	-7.561	-19.452	4.329				
- Good	20.170	12.330	28.010	***			
- Neutral	-27.731	-37.777	-17.685	***			
od - Bad	-20.170	-28.010	-12.330	***			

- Sunday.
- Does not consider Age factor of the customer.

- and Aerospace Engineering".
- Sentiment Mining".

Sujal Reddy Alugubelli **Data Scientist Intern – INVISTA** (405)385-2039 sujalreddyaugubelli@gmail.com Sujal reddy.alugubelli@okstate.edu

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LIMITATIONS and FUTURE SCOPE

Does not take into consideration the nature of the customer.

• Applies to a working customer who has work Monday through Friday and has off from work on Saturday and

Time of the day is one more important factor, which needs to be tracked.

REFERENCES

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CONTACT INFORMATION

MS Business Analytics | Spears School of Business | Oklahoma State University

https://www.linkedin.com/in/sujal-reddy-alugubelli-08651093







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2654 Understanding the influence of the day of the week in the reviews written using SAS® Enterprise Miner [™] and SAS® Sentiment Analysis Studio

Sujal Reddy Alugubelli, Oklahoma State University

ABSTRACT

Understanding customer needs is the most critical aspect in every business. Today most companies utilize customer feedback to understand customer desires. Analyzing what factors affect customer feedback is important for any company. This paper analyses the effect day of the week has on the way a user reacts. SAS® Enterprise Miner [™] and SAS® Sentiment Analysis Studio is used to analyze reviews written on weekdays and weekends. By understanding the differences in the opinions expressed on weekdays versus weekends, a company would have better understanding of their customer reviews. In this paper nearly 10,000 reviews written on Amazon.com for two products from an electronics product category are explored. The ratings given, the date on which a review is written and the actual written review are the key fields on which the analysis is carried. The paper systematically analyses the customer reviews and shows a few scenarios where terms that are considered as negative on weekdays are considered as positive on weekends. In one such scenario, "Ear buds" seem to be an issue for one of the products that is analyzed during the weekday where as "Ear buds" are received positively during the weekend for the same product. Likewise, after carefully analyzing the weekend and weekday reviews we can clearly understand the impact of weekdays on the negative reviews.

INTRODUCTION

In the recent times most manufacturers give utmost importance to the reviews written by users on social networking sites and ecommerce websites. It is very crucial for a firm to know the key feedback and thereby providing synthesized services to their customers ^[1]. With the huge number of reviews and comments available it is an arduous job to narrow down to the most significant reviews for the betterment of the business. One of the most important factor that a manufacturer could keep in mind while eliminating non-useful reviews is the effect of day of the week on the way a user writes a review.

According to research ^[4], human beings react differently at different times. They tend to over react when they have some pressures and are more accepting in nature when they are relaxed and calm. As most individuals are stressed on weekdays because of their work load and the things that happen in a day they are bit more arrogant whereas they are more peaceful and calm on weekends. Going by this concept, in this paper analysis is done on weekday and weekend reviews so as to understand the contributing factors of user comment. This can greatly aid a company in understanding if a review on a particular specification is written out of a real reason or was it just because of a user's bad mood that day.

This way of analyzing the user reviews and thereby providing synthesized services would greatly aid a company in multiple ways. First valuable point is that companies gets to know the true voice of their customers. The company would also save a lot of money and effort that would have been spent on non-significant requirements. This way a company can not only attain excellent customer satisfaction but also be efficient in providing the services.

METHODOLOGY

The process that is followed in this analysis is broadly classified into following steps:

- Data Extraction using Python.
- Data Preparation and initial analysis using Base SAS.
- Initial Text analysis and filtration of the text.
- Review analysis using SAS Enterprise Miner.
- User sentiment analysis using SAS Sentiment Studio.



Figure 1. Process flow diagram

DATA EXTRACTION USING PYTHON

For the purpose of this paper, the first step is to collect user ratings and written reviews. Python web scraper is used to extract the reviews from Amazon.com [5] and are parsed. Reviews written for two products pertaining to the cell phones accessories product category are extracted. The data has the information regarding the rating given, written review, the date on which review was written and a few other fields.

DATA PREPARATION AND INITIAL ANALYSIS USING SAS FOUNDATION

After extracting the data from the website, the second step is to format the data in a proper way suitable for the analysis. The day of the week needed to be obtained from the posted date and the reviews written on weekdays and weekends are to be sorted and should be classified as either positive review or a negative review.

Reviews that had a rating of 5 and 4 are classified as positive reviews and reviews with a rating of 1 and 2 are classified as negative reviews. As an initial step, a ANOVA test is performed so as to see the differences in the ratings given on weekdays and weekends. The results are visualized as shown below. We can clearly see from the plots below that the percentage of lower ratings (1 and 2) on weekdays is more when compared to the percentage on a weekend. This clearly shows numerically that users tend to given comparatively lower ratings on a weekday than on a weekend.



Figure 2. Product -1 Percentage of ratings on different weekdays.



Figure 3. Product -2 Percentage of ratings on different weekdays.

TEXT MINING AND REVIEW ANALYSIS USING SAS ENTERPRISE MINER

After preparing the data suitably for text mining, data is brought into SAS Enterprise Miner. In this step initially basic text analysis is performed using text filter and text parsing nodes [3] and the frequencies of terms and documents are observed. The initial results are as follows



Figure 4. Results of Number of Documents by weight using Text filtering node SAS EM

Term	Role	Altribute	Freq	# Docs	Keep
+ be	Verb	Alpha	2225	64	2N
+ have	Verb	Aloha	1329	48	BN
not	Adv	Aloha	1249	47	BN
+ use	Verb	Aloha	676	40	DN
+ do	Verb	Alpha	743	35	2N
+ headphone	Noun	Aloha	741	32	7Y
+ phone	Noun	Aloha	567	32	5Y
+ work	Verb	Alpha	454	31	1Y
+ mod	Adi	Alpha	423	30	9Y
+ quality	Noun	Aloha	304	20	0Y
Verv	Adv	Alpha	444	28	7N
+ hondent	Noun	Aloha	629	29	5Y
+ net	Verb	Aloha	449	20	201
+ cound	Noun	Alpha	260	27	2V
T SUGIA	Nour	Aloha	300	20	ev.
hoor	Moun	Aloha	410	20	
+ ear	NOUN	Alpha	919	20	91
sound		Alpha	209	23	31
comortable	Adj	Alpha	202	20	
+ DUY	Verb	Alpha	201	20	31
no	Adv	Alpha	2/0	20	311
+ great	Adj	Alpha	246	19	5Y
5	Noun	Alpha	355	18	6N
+ price	Noun	Alpha	220	18	3Y
+ listen	Verb	Alpha	228	18	1Y
bluetooth	Noun	Alpha	253	17.	3Y
+ call	Noun	Alpha	243	17.	2N
great	Adv	Alpha	186	17	OY
also	Adv	Alpha	236	16	9N
+ wear	Verb	Alpha	209	16	OY
+ go	Verb	Alpha	193	15	5N
battery	Noun	Alpha	162	14	9Y
sound quality	Noun Group	Alpha	170	14	BY
SO	Adv	Alpha	181	14	7N
good	Noun	Alpha	173	14	6Y
+ recommend	Verb	Alpha	154	14	6Y
rust	Adv	Alpha	183	13	9N
+ love	Verb	Alpha	157	13	BY
+ hour	Noun	Alpha	178	13	79
+ make	Verb	Alpha	183	13	7N
Ef.a.	Mauro	Alaba	4.47	40	ev.

Figure 5. Results of Terms using Text parsing node SAS EM

REVIEW ANALYSIS USING SAS ENTERPRISE MINER

Once the initial filtration and analysis is done the following steps are done so as to analyze the written reviews and extract the information that would be useful for eliminating the non-significant reviews and also determine the effect of day of the week on the reviews. This step is the most crucial and the core process of the analysis.

- As the reviews are already classified as positive and negative within weekdays and weekends, the negative reviews written on a weekday are of utmost concern. This dataset is analyzed using the text cluster, text topic and concept links.
- By performing the above mentioned process few significant and moderately important topics are picked.
- Next the weekend positive reviews are analyzed using the text cluster, text topic and concept links. The previous picked topics from weekday negative reviews are looked for in the positive weekend reviews.
- After comparing, if the significant negative comments present in the weekday reviews are also present in the positive reviews of the weekend then most likely those reviews are because of the mood of the customer.

By following the above steps, it yields the results and helps us better understand the true voice of the customers. In addition to the analysis performed we could also perform sentimental analysis using SAS Sentiment Analysis Studio. This can clearly depict what a customer feels on a weekday and what they have in mind on a weekend.



Figure 6. SAS Enterprise Miner model for the analysis

USER SENTIMENT ANALYSIS USING SAS SENTIMENT ANALYSIS STUDIO

In order to see what a user in general believes on a weekday and determine if it is different from the weekend, sentiment analysis is performed using SAS Sentiment Analysis. By this one can get fair understanding if the reviews written are different on both the cases or are the same.

Using the current data the results of sentiment analysis is as below.

Positive Negative Overall

BEST MODEL is Smoothed Relative Frequency and Chi Square





Figure 7. Best model selected by the sentimental analysis studio

Posit	ve Negative	Neut	tral	
	Type	∇	Body	Weight
1	CLASSIFIER		great	0.046423
2	CLASSIFIER		price	0.039659
3	CLASSIFIER		music	0.024507
4	CLASSIFIER		sound	0.024349
5	CLASSIFIER		phone	0.023998
6	CLASSIFIER		comfortable	0.023585
7	CLASSIFIER		love	0.021847
8	CLASSIFIER		cal	0.018949
9	CLASSIFIER		listen	0.016861
10	CLASSIFIER		easy	0.015936
11	CLASSIFIER		clear	0.013647
12	CLASSIFIER		Bluetooth	0.013484
13	CLASSIFIER		best	0.013289
14	CLASSIFIER		always	0.013184
15	CLASSIFIER		far	0.013174
16	CLASSIFIER		Great	0.011991
17	CLASSIFIER		cel	0.011923
18	CLASSIFIER		happy	0.011455
19	CLASSIFIER		life	0.011339
20	CLASSIFIER		wife	0.010981
21	CLASSIFIER		value	0.010981

Figure 8. Statistical rules formed to classify a review as positive

Posi	ive Negative Neu	tral	
	Type $ abla$	Body	Weight
1	CLASSIFIER	stop	0.067178
2	CLASSIFIER	return	0.064698
3	CLASSIFIER	poor	0.039243
4	CLASSIFIER	disappoint	0.029197
5	CLASSIFIER	week	0.027773
6	CLASSIFIER	send	0.025152
7	CLASSIFIER	waste	0.024742
8	CLASSIFIER	not	0.023669
9	CLASSIFIER	waranty	0.023093
10	CLASSIFIER	Then	0.022898
11	CLASSIFIER	minute	0.022034
12	CLASSIFIER	month	0.020863
13	CLASSIFIER	red	0.020676
14	CLASSIFIER	working	0.017632
15	CLASSIFIER	completely	0.017486
16	CLASSIFIER	stopped working	0.017265
17	CLASSIFIER	replacement	0.016953
18	CLASSIFIER	useless	0.01647
19	CLASSIFIER	plan	0.016339
20	CLASSIFIER	refund	0.014933

Figure 9. Statistical rules formed to classify a review as positive

RESULTS

- In the current scenario, for the product Motorola Stereo Headset, the term volume seems to be one of the significant topics in negative weekday reviews. The same term appears in the weekend significant positive reviews list.
- This clearly raised the flag of uncertainty of the proper feedback. For this purpose, the reviews that consisted of that term were pulled and analyzed.
- After analyzing the reviews, it is evident that there is actually no problem with the volume aspect of the device. So the negative reviews written on weekday regarding volume are only because of bad mood of the customer and not really a problem of the product.
- Similarly, with the second product LG Tone HBS-730 wireless stereo headset. There were negative reviews written about earbuds on weekdays but a lot of positive feedback is written on weekend. Thus it clearly shows the presence of day of the week effect the way people write a review.

So, it seems that there is an effect of the day of the week on the way people write a review. By performing such analysis, companies can better understand the nature of the reviews.

MODEL: LG TONE HBS-730 WIRELESS STEREO HEADSET – BLACK

Cluster ID	Descriptive Terms	F	Frequency
	1+volume +loud +'low set' +'volume control' +low +control +ios +set +iphone +music +comfortable +listen +device +button +read		45
	2+pocket +foot +phone +signal +walk +drop +cut +bluetooth +connection +range +lose +listen +pair +keep +headphone		104
	3audio +support +issue +hbs-700 +connection +music +iphone +battery +ios +device +connect +comfortable different +second +know		116
	4+charge +charger +full +minute +replacement first +day +unit +turn +keep +hold right +work +package amazon		63
	5+model 'previous model' previous +version +old +battery +love +range +second +'sound quality' +year +hbs-700 back +cut +issue		33
	6+box +package real +fake +open amazon +order +original +charger +receive +cheap +look +item different +warranty		75
	7+hear +noise people +talk money +problem +work well +item +want +stop +know +button +set back		115
	8+month +bud +'ear bud' +ear +warranty +break +wire +stop +last +year +piece +long +design +cheap +work		90
1	9+quality +'sound quality' +sound +range +horrible +bluetooth +hope +'bluetooth headset' +review +read +device +good lg +headset +tone		62



Cluster ID	Descriptive Terms	Frequency
	1+headphone +wire +review +issue +range +problem +set +want +pair +little +fit +neck +order +buy well	. 476
2	2+noise +volume +voice +hear +connect +device well +little +thing 'sound quality' +range +time +excellent +quality +problem	. 423
:	3+battery +life +'battery life' +good +long +quality +sound +great +comfortable 'sound quality' +amaze +range +easy +excellent +far	. 176
4	4+phone +music +love +listen +hand +drive +talk +free cell +product able +work +great +'great product' +clear	. 221
	5+battery +day +last +charge +hour 'long time' +long +week +work +time +listen +wear +comfortable +'great product' +music	. 148
(6+love +great +recommend +product +buy +purchase +husband +friend +gift family +order +price +'great product' +item +awesome	. 339
	7lg +tone +l'g tone' +headset +bluetooth +look +review cell +recommend +friend +long +talk +highly +item +voice	. 92
8	8+ear +bud +'ear bud' +piece +stay +neck +fit +little +hold +wear +happy +item +comfortable +thing +hear	. 177

Figure 11. Clusters formed for positive reviews on Weekend



Figure 12. Terms connected to "Bad" on weekday



Figure 13. Terms connected to "Earbud" on weekend



Figure 14. Examples of positive reviews about earbud on Weekend



2

Figure 15. Examples of negative reviews about earbud on Weekday

MODEL: MOTOROLA S305 BLUETOOTH STEREO HEADSET W/ MICROPHONE (BLACK)



Figure 16. Terms connected to Volume on Weekend



Figure 17. Terms connected to Volume on Weekday

Cluster ID	Descriptive Terms	Frequency
	1+listen +music +'battery life' able +pair +battery +purchase back first amazon +motorola +life +device +problem +feel	 37
	2+device bluetooth price +recommend +issue +keep +headset +phone +great motorola quality +bluetooth +bad +find +connect	 56
:	3+arrive +mac +iphone +pretty amazon sound +connect +'sound quality' +day quality +easy return +stop +connection +feel	 29
	4+loud +hear +ipod +noise +laptop +finally +sound +little +phone +comfortable +connect +fit +connection +bluetooth +return	 32
	5+volume +button +'volume button' +hit +break +control +low +look +useless +replace +easy +month +drop +hard +long	 28
	δ+item +waste money +fail return +window amazon +purchase +week +replacement +time +device +connection +second working …	 31
	7red 'turn on' +charge +last +running +stop +sweat +month working +light +workout great gym few +week	 92
1	8+wear +head +uncomfortable +ear +minute +fit +hour +hard +comfortable +hear +phone +design +music +noise +feel	 52



Cluster ID	Descriptive Terms	Frequency
	1+touch +ipod +'ipod touch' ipod +pause +play +audio +button +volume +control +wire back +turn well +computer	. 8
	2+hear +headphone +'bluetooth headphone' +price +first +work +pair great +set +music +bluetooth +little +great +sound +listen	10
	3+headset +'bluetooth headset' +listen +music +look +excellent +want +hear +bluetooth +s305 bluetooth +phone +purchase +player +recommend	. 9
	4+battery +life +'battery life' +'sound quality' sound +quality +easy +excellent +last +set +hour +iphone +charge +love +buy	. 5
	5+battery +charge +device +day +hour +last +life +iphone +'bluetooth headphone' +connect +'battery life' +pair +headphone +problem +button	20
	3+motorola +bud +s305 +ear +wear +head +find +comfortable +fit +recommend bluetooth +audio +listen +control +set	. 8
	7+product +great +price +love +sound +gym +recommend great +fit +good +purchase +thing +buy +excellent +range	. 13
	Btooth +blue +'mp3 player' +player +answer running able +exercise +wire +phone +problem +audio right +computer +pause	. 3
	9+watch +movie +tv +ipad +computer bluetooth +play +turn +motorola +audio +range +device +player +find +want	. 4

Figure 19. Clusters formed for positive reviews on Weekend

Price was good, but these hurt my ears, and the audio had very high treble, so when I turned up the volume it hurt my ear drums. I was getting these to replace my RocketFish RF-MAB2, but it turns out that the RocketFish is much better. They have a graphic equalizer built in and they fit much more comfortably. I'm sticking with those.	2	Okay so this thing works. It works fine as a bluetooth device and a headphone. Having said that, there are several issues. The quality is just not upto par. I'm not expecting audiophile quality but c'mon, it's noisy and it at times picks up other signals and noise. The volume is also anemic and thin. The whole thing feels a tad shabby in terms of construction. I returned these immediately.	2
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Figure 20. Examples of negative reviews about volume on Weekend

SOUND IS GREAT THE VOLUME DOESN'T GO THAT LOUD BUT GREAT SURROUND AND BASS ON MY MUSIC WHICH IS JAZZ AND OLD SCHOOL AND ONE AWESOME THING IS THE FLEXIBILITY OF THE HEADPHONE I HAVE DROP THEM ABOUT 10 TIMES SMASH THEM IN MY BOOKBAG AGAINST MY BOOKS BY MISTAKE OF COURSE EVERYDAY AND THEY STILL WORK AND LOOK	So far I love the quality of sound both for calls, and for listening to music, or audio books. The volume control is great because it emits a tone change so that you know if you are increasing or decreasing the sound. Over-all good quality at a great price!
--	---

Figure 21. Examples of positive reviews about volume on Weekend

LENGTH OF REVIEW EFFECT

Analysis Variable : no_words							
Rate	N Obs	Mean	Std Dev	Minimum	Maximum	Ν	
Bad	1100	94.2209091	100.2060993	2.0000000	884.0000000	1100	
Good	5339	74.0513205	98.0537850	1.0000000	1460.00	5339	
Neutral	620	101.7822581	124.6959313	1.0000000	1559.00	620	

Figure 22.	Mean	number	of words	for type	of review

Comparisons significant at the 0.05 level are indicated by ***.					
	Difference				
Rate	Between	Simultaneous 9	5% Confidence		
Comparison	Means	Lin	nits		
Neutral - Bad	7.561	-4.329	19.452		
Neutral - Good	27.731	17.685	37.777	***	
Bad - Neutral	-7.561	-19.452	4.329		
Bad - Good	20.170	12.330	28.010	***	
Good - Neutral	-27.731	-37.777	-17.685	***	
Good - Bad	-20.170	-28.010	-12.330	***	

Figure 23. ANOVA results for effect of number of words in a review

CONCLUSION

This paper illustrated the influence the day of the week has on the way a customer writes a review. Specifically, in the two examples, data is extracted and prepared for text analytics, performed text analytics, prepared data for exploration and reporting. Manufacturers can leverage this analysis or elements thereof, in their own analysis.

In this paper I explored the behavior changes observed using text analysis. Text cluster node and text topic node helped in finding out the significant review written by the customer. Text filter node helped in finding out the terms that are related to those significant topics. With this information a company can greatly synthesize its services and optimize the efforts spent on the betterment of the product.

FUTURE SCOPE AND LIMITATIONS

- The analysis performed in this paper does not take into consideration of the nature of the customer. So if the type of a customer could be known then evaluation of the results would make much more sense.
- The paper applies to a working customer who has work Monday through Friday and has off from work on Saturday and Sunday.
- May be the product from product categories specific to certain age groups could be analyzed so
 as to understand if there is an age factor. For example, most video games products are used by
 younger generations and home appliances are purchased by older generation so products from
 those products categories would help us understand if there is any age factor on how the
 behavior changes according to day of week.
- The length of the review written also has a significance in the intension of the review written. Surveys tell us that when customers write a short but harsh comment in most cases it is not a true one and biased by some factor. So we can add the length of a comment factor so as to tune the comments.
- Time of the day is one more important factor, which needs to be tracked. Comments would be better in the early part of the day than in later part by following the concept of change in behavior because of work pressures. So if we could bring in the time factor that would greatly aid the company

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CONTACT INFORMATION

Your comments and questions are valued and encouraged. Contact the author at:

Sujal Reddy Alugubelli Oklahoma State University (405)385-2039 sujalreddyalugubelli@gmail.com, sujal_reddy.alugubelli@okstate.edu https://www.linkedin.com/in/sujal-reddy-alugubelli-08651093