

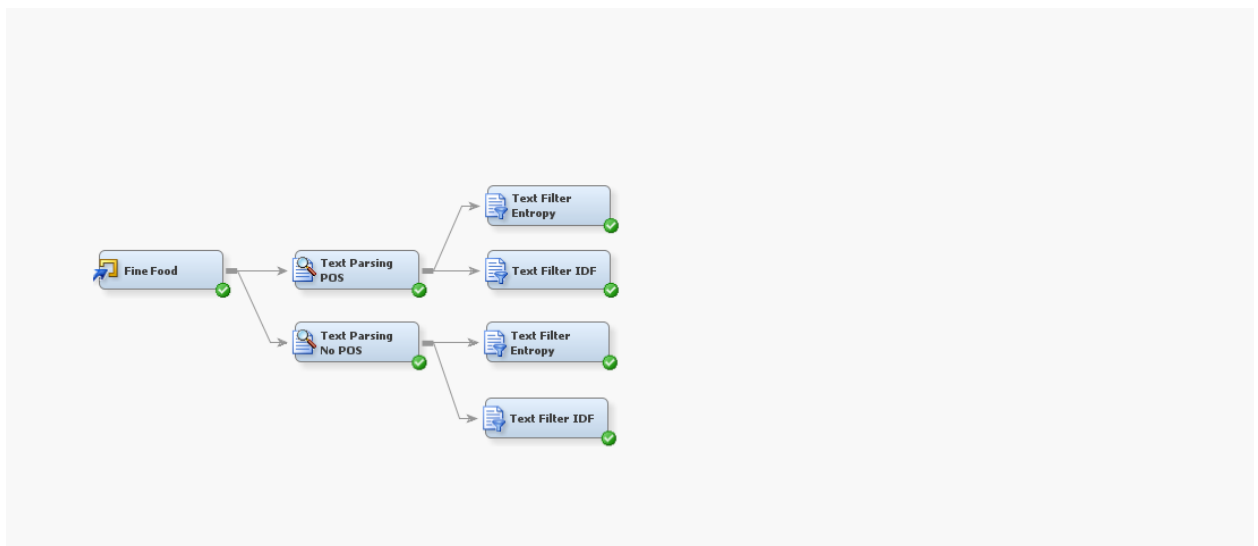
Assignment 2

The goal of this assignment is to apply text parsing and filtering in real context. To do so do the following:

- 1- Go to the dataset: <https://www.kaggle.com/snap/amazon-fine-food-reviews/downloads/amazon-fine-food-reviews.zip/2>

(You may need to register to Kaggle.com to download the dataset)

- 2- Import the data file to SAS enterprise miner
- 3- Apply text parsing and filtering on the dataset.



4- Take screen captures of the outcome of your graphs and findings

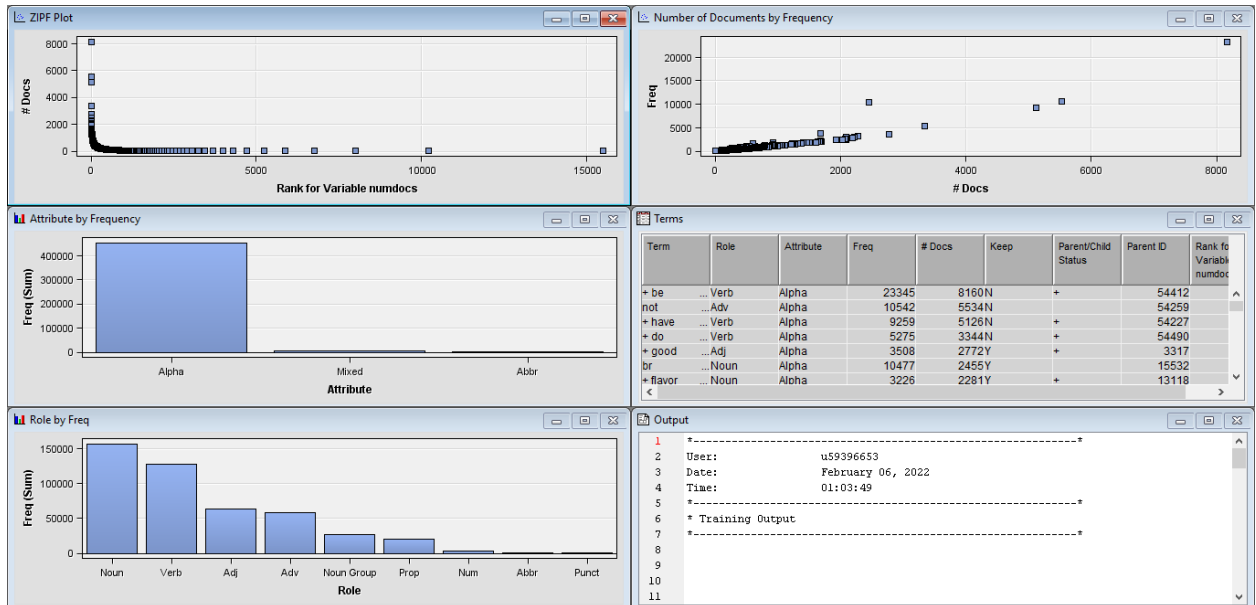


Figure 1 Text Parsing POS Results

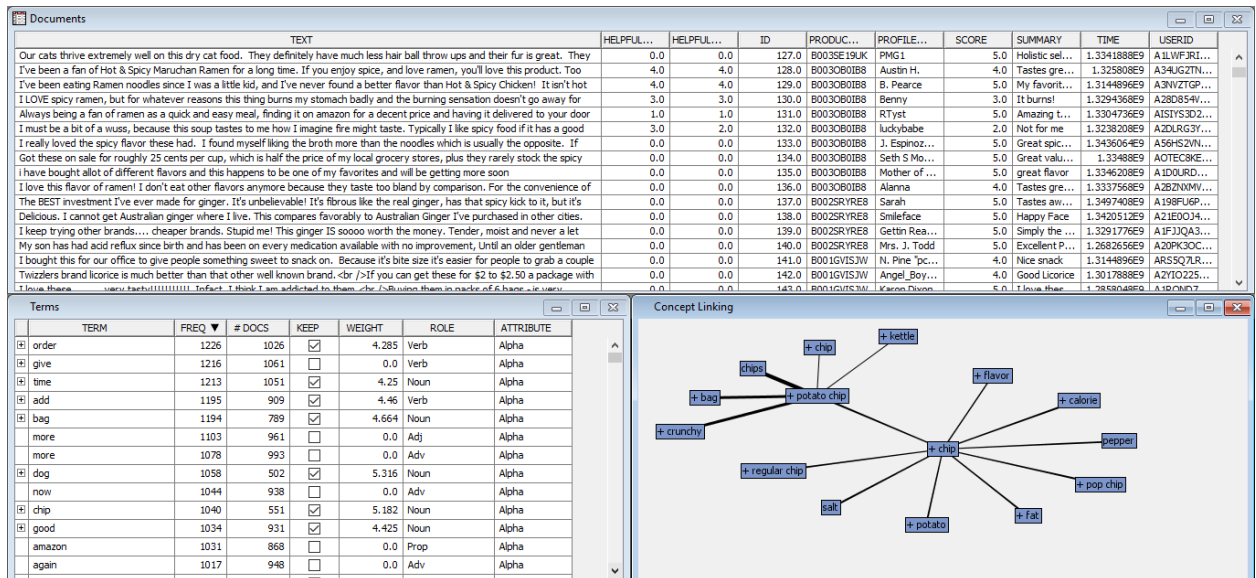


Figure 2 Text Filter Interactive View

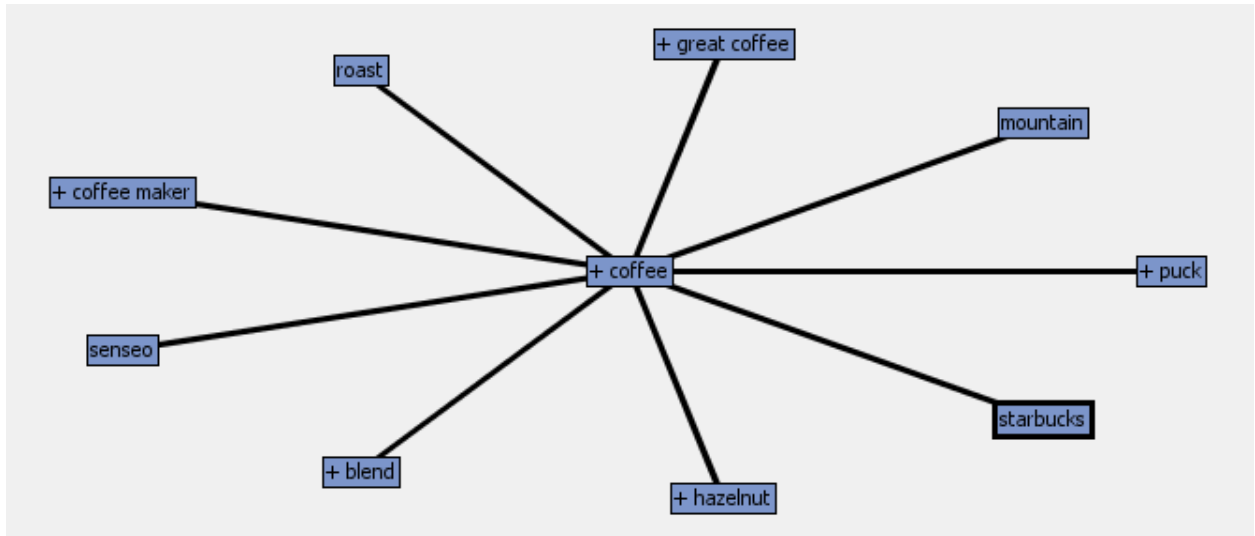


Figure 3 Identifying the word Coffee and words associated with it

5- Identify synonyms in your data and combine them together using interactive text explorer

6- What words/word-combinations have you combined?

- br occurred 10477 and it is code for line break, Amazon (Prop & noun) Occurred 497 times and it adds no meaning therefore they were dropped.
- Purchase & buy gives the same meaning in the data set.
- Great & Excellent has the same level in sentiment and depending on the analysis other words can be added as synonyms like good.
- Cat & Dog can be added if we are referring to pets.
- Coffee, Tea, Cocoa (Cocoa & Hot Cocoa are synonyms) can be combined as hot beverages.

7- Give some examples of the entities identified by the text parser

Starbuck, coffee, tea, bag, sugar, chocolate, juice, soda, cookie, potato chip

8- Repeat the previous steps while disabling POS. how are your results affected?

While using parts of speech words like great and excellent had to be matched with the right POS before treating them as synonyms, if POS was not used then it's not a problem.

By comparing the two methods some words changed in position in terms of frequency, but I don't think it make much of a difference while looking at the top 100 words for example.

Term	Role	Attribute	Freq ▼
+ be	...	Alpha	23526
not	...	Alpha	10554
br	...	Alpha	10477
+ have	...	Alpha	9261
+ good	...	Alpha	6184
+ do	...	Alpha	5430
+ taste	...	Alpha	4565
+ flavor	...	Alpha	4047
+ coffee	...	Alpha	3947
+ make	...	Alpha	3104
+ try	...	Alpha	3037
+ product	...	Alpha	3010
+ love	...	Alpha	2903
+ great	...	Alpha	2831
very	...	Alpha	2822
just	...	Alpha	2813
+ get	...	Alpha	2741
+ buy	...	Alpha	2673
+ use	...	Alpha	2626
+ find	...	Alpha	2228
more	...	Alpha	2183
+ other	...	Alpha	2138
+ food	...	Alpha	2074
+ like	...	Alpha	2058
+ little	...	Alpha	2005
+ no	...	Alpha	1818
so	...	Alpha	1799
+ one	...	Alpha	1792
+ tea	...	Alpha	1790
really	...	Alpha	1777
+ eat	...	Alpha	1708
+ drink	...	Alpha	1708
+ cup	...	Alpha	1610
too	...	Alpha	1598
+ order	...	Alpha	1597

Figure 4 No POS

Term	Role	Attribute	Freq ▼
+ be	... Verb	Alpha	23345
not	... Adv	Alpha	10542
br	... Noun	Alpha	10477
+ have	... Verb	Alpha	9259
+ do	... Verb	Alpha	5275
+ coffee	... Noun	Alpha	3755
+ good	... Adj	Alpha	3508
+ flavor	... Noun	Alpha	3226
+ make	... Verb	Alpha	3015
+ product	... Noun	Alpha	2999
very	... Adv	Alpha	2812
+ get	... Verb	Alpha	2710
+ buy	... Verb	Alpha	2513
+ love	... Verb	Alpha	2484
+ taste	... Noun	Alpha	2449
+ try	... Verb	Alpha	2446
+ use	... Verb	Alpha	2283
+ taste	... Verb	Alpha	2097
+ find	... Verb	Alpha	2039
+ like	... Verb	Alpha	1970
just	... Adv	Alpha	1903
so	... Adv	Alpha	1789
really	... Adv	Alpha	1775
no	... Adv	Alpha	1767
+ food	... Noun	Alpha	1763
+ eat	... Verb	Alpha	1693
+ tea	... Noun	Alpha	1686
too	... Adv	Alpha	1596
+ great	... Adj	Alpha	1561
+ little	... Adj	Alpha	1542
other	... Adj	Alpha	1480
one	... Num	Alpha	1475
also	... Adv	Alpha	1441
+ go	... Verb	Alpha	1440
+ cup	... Noun	Alpha	1440

Figure 5 POS