

UBUNTU - A TEXT ANALYSIS EXPLORATION

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In order for us to achieve such a task, first we have to:

OUR GOAL

Text Analysis that will help create and embed chatbot to assist users debug their queries.



Problem

Anytime someone has a question or query, they ask and have to wait for someone to respond



Solution

With past information and data, we can create a FAQ and a chatbot that replies instantly

Impact

More people will be willing to purchase product as debugging will be easier.

Overview of Dataset & Pre-processing

Our Dataset is made up of 6 Main columns

1.Folder:

The folder the query was retrieved from

2.DialogueID:

ID that a set of text corpus is part of

3. Date:

Timestamp when query was submitted

4. From:

The user who sent the query

5. To:

The user whom the answer is for

6. Text - Our heart

The line of text that is going to help our analysis. The question and response

Initial Dataset



1,038,325 Lines



11,035,331 Words



116,070,597 Characters

- Removed1630 Duplicate rows
- Filled in 87 empty cells in text column
- Removed Stop words , Punctuation, Spaces
- Removed Digits
- Added Parts of Speech
- · Lemmatized the dataset and
- Made all our letters into lower cases

Final Dataset



1,038,237 Lines



6,101,882 Words



37,451,381 Characters

1 .							
		folder	dialogueID	date	from	to	text
	0	3	126125.tsv	2008-04-23T14:55:00.000Z	bad_image	NaN	Hello folks, please help me a bit with the fol
	1	3	126125.tsv	2008-04-23T14:56:00.000Z	bad_image	NaN	Did I choose a bad channel? I ask because you
	2	3	126125.tsv	2008-04-23T14:57:00.000Z	lordleemo	bad_image	the second sentence is better english and we
	3	3	64545.tsv	2009-08-01T06:22:00.000Z	mechtech	NaN	Sock Puppe?t
	4	3	64545.tsv	2009-08-01T06:22:00.000Z	mechtech	NaN	WTF?

1 •	folder	dialogueID	date	from	to	text	Tokens	LongWords	word_count	char_count	part_of_speech
	0 3	126125.tsv	2008-04-23 14:55:00+00:00	bad_image	moderator	hello folk please help bit following sentence	[hello, folk, please, help, bit, following, se	[please, following, sentence, personal, allowe	19	126	[(hello, NN), (folk, NN), (please, NN), (help,
	1 3	126125.tsv	2008-04-23 14:56:00+00:00	bad_image	moderator	choose bad channel ask seem dumb like window user	[choose, bad, channel, ask, seem, dumb, like,	[choose, channel, window]	9	49	[(choose, RB), (bad, JJ), (channel, NNS), (ask
	2 3	126125.tsv	2008-04-23 14:57:00+00:00	lordleemo	bad_image	second sentence better english dumb	[second, sentence, better, english, dumb]	[second, sentence, better, english]	5	35	[(second, JJ), (sentence, NN), (better, RBR),
	3 3	64545.tsv	2009-08-01 06:22:00+00:00	mechtech	moderator	sock puppe	[sock, puppe]	0	2	10	[(sock, NN), (puppe, NN)]
	4 3	64545.tsv	2009-08-01 06:22:00+00:00	mechtech	moderator	wtf	[wtf]	0	1	3	[(wtf, NN)]

EDA Findings

Crucial Findings Important for Modeling

Most Used Words -

Help us determine most popular topics

Years -

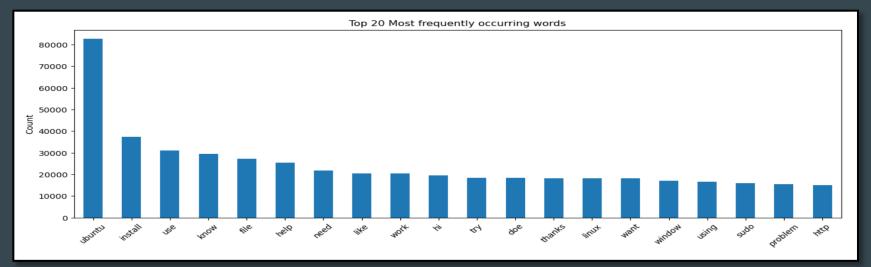
2008-2012. Work with Old Data to Better understand New Data

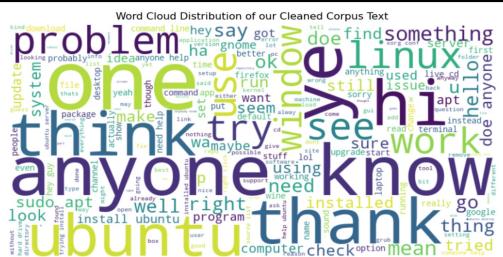
Active Users –

Knowing who is helping us answer user enquiries

DialogueId -

Total of 1,036,607 conversations had, out which 346,108 are unique.





Top 5 User with	Questions:
ubuntu	1578
jrib	1342
Pici	1289
bazhang	1276
ActionParsnip	1182
Name: to, dtype:	int64

Top 5 Users w	ith answers			
bazhang	5278			
ActionParsnip	5010			
jrib	4586			
Pici	4297			
ikonia	4069			
Name: from, d	type: int64			

Baseline Model & Evaluations

We have a unique dataset from other Machine Learning models, based on how we don't have a target.

Just working with the Text Column

Due to this we will do **Unsupervised Learning**.

Word2Vector

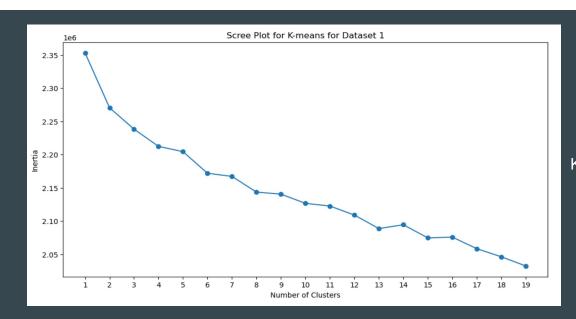
Generate Word
Embeddings. Vector
Representation of
words capturing
similarities between
words in context

Latent Dirichlet Allocation (LDA)

Topic Modelling via Statistics. Document is topics and topics is words

K-Means

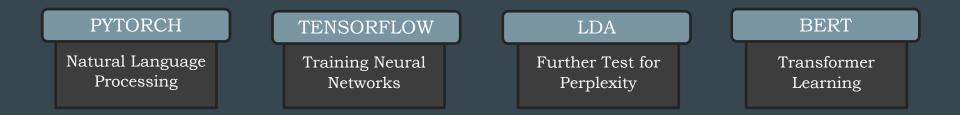
Principal Component Factor Analysis to retain Dimensionality reduction Word Embeddings for 'Ubuntu' Vector for 'word': [-0.3782954 -1.116179 1.2351896 0.87763566 -0.70066255 -0.27339867 0.47078404 -0.86039793 1.4903036 0.8904295 0.7389216 1.4677155 0.614419 1.730918 -1.7647492-0.9134033-0.56066620.36198437 1.1075292 2.649643 1.0301251 2.3485653 -0.81350666 1.6191515 1.3545189 -1.3117663 0.8473911 -1.2159309 0.06076604 2.2887218 -0.39250907 - 2.02817750.04664294 -0.5923905 0.6952565 -1.04123680.9254655 0.3610684 1.7880002 1.193306 -1.93962650.6639015 0.7561573 0.38758996 -1.1962883-0.64587230.18257742 1.4984949 -0.522098961.8927822 0.41670617 0.55838144 -2.1684086 0.08351981 2.3480287 -2.3033633 1.6280334 -0.08160885 -0.14495052 -1.8332355-1.1863732 -1.5611942-1.5474703-0.42522330.8345557 -1.18592720.36316046 -0.20094004 2.0028698 0.5570511 1.2199756 1.5931635 -1.70950330.8225964 1.1340362 -0.16668206 -1.5949805 -0.12305047 -2.4854198 -1.93178640.03640938 0.8557327 0.9996959 -0.03400733-0.10375319 -0.8452185-0.0380686 0.96657485 0.40598986 -0.03777267 -0.20533822 -1.44204842.2234952 0.50122136 - 0.5545503-2.291512 -0.215510580.01879653 -0.45425466 1.5494255]



K-Means Inertia Plot

Next Steps for Advanced Modeling

We have gotten to know our dataset even more now and have understood what it is made up of and our limitations.



- Adding new features in our code and Constantly training the model, we should have enough analysis to create our bot and push it through production
- The models we wish to create, there exists multiple other version in different classes. We can even use what exists there to better our model and learn from them.
- In addition to all of this, Deep Learning and Language Learning Models will help us make our model stronger.