TedTalk Data Analysis

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Introduction & Project Scope



A research university trying to break into the online learning community



- **Increase** student enrollment rate
- **Enhance** professors recognition
- **Highlight** university's public image
- Gain status in higher academic circles



Ted Talk Dataset

- Variables that affect online video views and ratings
- Transcript data using NLP analysis
- Insights in online educational video industry



- Understand how different qualities of videos affect views and rating score
- Predict video rating and views with exist and new variables from feature engineering and text mining

Database Schema

- Dataset Description: all audio-video recordings of TED Talks uploaded to TED.com until Sep 21st, 2017
- Variables:

```
[1] "comments"
                          "description"
                                                "duration"
[4] "event"
                          "film date"
                                                "languages"
[7] "main_speaker"
                          "name"
                                                "num_speaker"
[10] "published_date"
                          "ratings"
                                                "related talks"
[13] "speaker_occupation" "tags"
                                                "title"
[16] "url"
                           "views"
```

- Additional Details:
 - Two datasets:
 - Main dataset: 2550 tuples, 17 features; information about all talks
 - Transcript dataset: 2464 tuples, 2 attributes; the transcripts for all talks

ETL Process

- Merged data sets
- Split up columns with multiple values
 - Lists
 - Nested jsons
- Created unique identifiers for each table
- Transformed and created new variables
- Ensured all records in each table were unique
- Enforced referential integrities

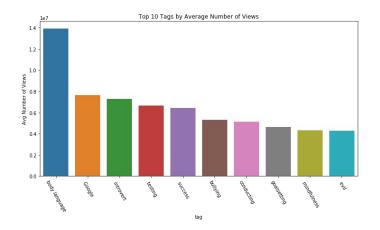
main df.head()

	comments	description	duration	event	film_date	languages	main_speaker	name	num_speaker	published_date	 transcript	vic
0	4553	Sir Ken Robinson makes an entertaining and pro	1164	TED2006	24-02- 2006	60	Ken Robinson	Ken Robinson: Do schools kill creativity?	1	26-06-2006	 Good morning. How are you? (Laughter)It's been	0
0	4553	Sir Ken Robinson makes an entertaining and pro	1164	TED2006	24-02- 2006	60	Ken Robinson	Ken Robinson: Do schools kill creativity?	1	26-06-2006	 Good morning. How are you? (Laughter)It's been	0
0	4553	Sir Ken Robinson makes an entertaining and pro	1164	TED2006	24-02- 2006	60	Ken Robinson	Ken Robinson: Do schools kill creativity?	1	26-06-2006	 Good morning. How are you? (Laughter)It's been	0
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Analytical Procedures: Data Exploration

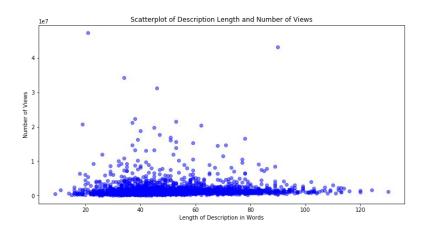
Data Exploration:

- Focused on outcome variable (views)
- Analyzed relationships with predictor variables through Matplotlib & Seaborn



Feature Engineering:

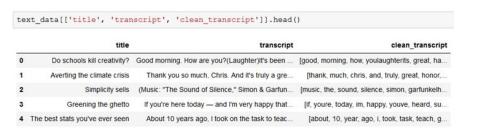
- Title Length and Description Length (both measured in characters and words)
- Data transformation (next slide)

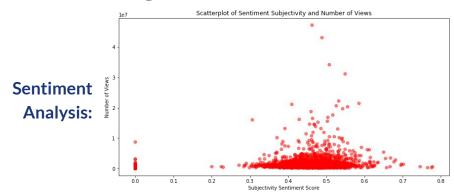


Analytical Procedures: Text Mining

Transcripts for each TED Talk were cleaned using a defined process:

- Remove punctuation
- Remove stop words
- Make characters lower-cased
- Lemmatize remaining words
- Store in list format





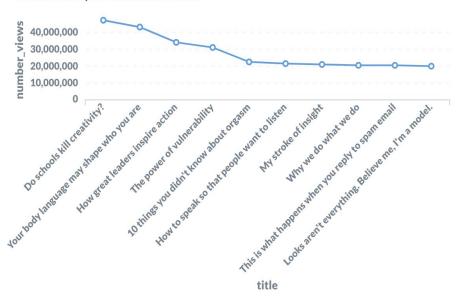
Count Vectorization with Scikit Learn:

		0	01	05	1	10	100	1000	10000	100000	 zerosum	zimbabwe	zip	zombie	zone	zoning	Z00	zoom	zooming	zurich
0	0	0	0	0	0	0	0	0	0	0	 0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0	0	0	0	 0	0	0	0	0	0	0	0	0	0
2	0	0	0	0	0	1	0	0	0	0	 0	0	0	0	0	0	0	0	0	0
3	0	0	0	0	0	0	0	0	2	0	 0	0	1	0	0	2	0	0	0	0
4	0	0	0	0	0	4	3	1	0	0	 0	0	0	0	0	0	0	0	0	0

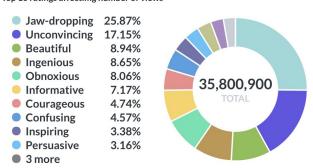
5 rows × 11763 columns

Metabase Dashboard: C Level

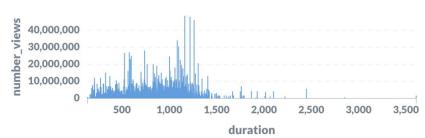




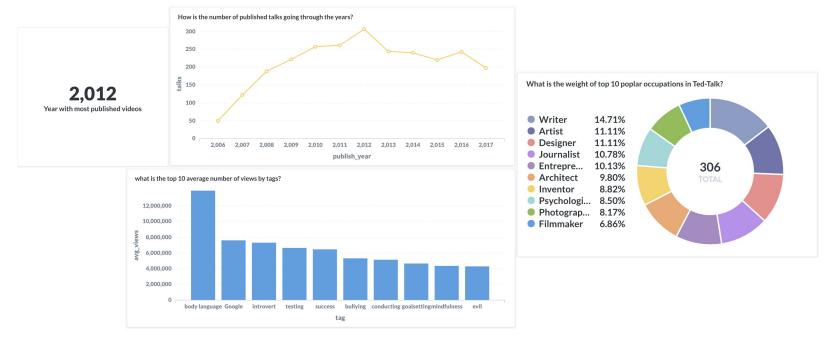
Top 10 ratings affecting number of views



Does duration of a video affect the avg number of views?



Metabase Dashboard: Analysts



Conclusion & Key Takeaways

- PostgreSQL effective way to store this database
 - ETL process important for selected dataset
 - o If not in 3NF, data would likely fit best in a NoSQL system such as MongoDB
- Python makes it easy to pull this data from our database and manipulate for new tasks
- Metabase was an effective tool for dealing with our database at all stakeholder levels
 - Analysts can use SQL directly in the system
 - Executives can view dashboards with only the most critical information
- Additional feature engineering and data exploration must be done to better understand the factors that affect video view counts