

# **Capstone Project - 2**

## **TED Talk Views Prediction**

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#### **Problem Statement**



- TED is devoted to spreading powerful ideas on just about any topic. These datasets contain over 4,000 TED talks including transcripts in many languages is a nonprofit organization that aimed at bringing experts from the fields of Technology, Entertainment, and Design together.
- TED Conferences have gone on to become the Mecca of ideas from virtually all walks of life.
- As of 2015, they published more than 2000 talks for free consumption by the masses and its speaker list boasts of the likes of Al Gore, Jimmy Wales, Shahrukh Khan, and Bill Gates.
- The main objective is to build a predictive model, which could help in predicting the views of the videos uploaded on the TEDx website.



# **Data Summary:**

Data set name: data\_ted\_talks

#### **Shape:**

- Rows -- 4005
- Columns--19

#### **Features:**

```
'talk_id', 'title', 'speaker_1', 'all_speakers', 'occupations', 'about_speakers', 'recorded_date', 'published_date', 'event', 'native_lang', 'available_lang', 'comments', 'duration', 'topics', 'related_talks', 'url', 'description', 'transcript'
```

**Target Variable: 'views'** 

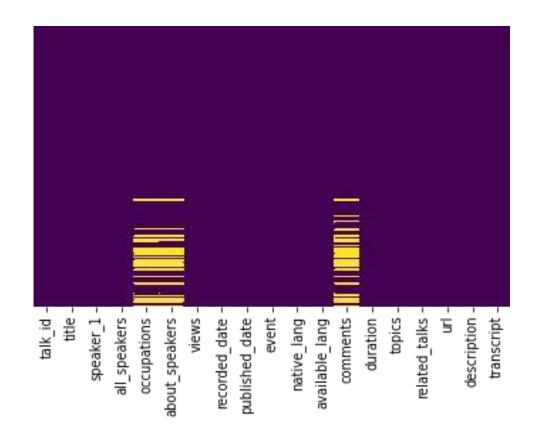


# **Exploratory Data Analysis on Features**



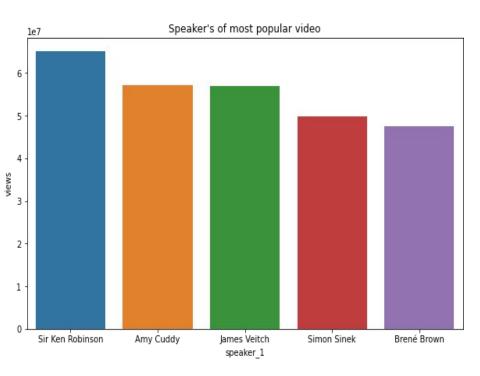
# **Missing Data Check**

- KNN imputation for Numerical Features
- Replaced Categorical
  Features Nan values with
  'Unknown' category





# **Speakers with Views:**



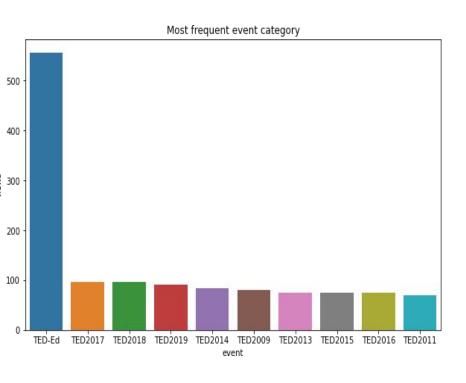
Most popular Speaker according to total views on their talks 1.2 1.0 0.8 0.6 0.4 0.2 0.0 James Veitch Sir Ken Robinson Simon Sinek Brené Brown Alex Gendler speaker 1

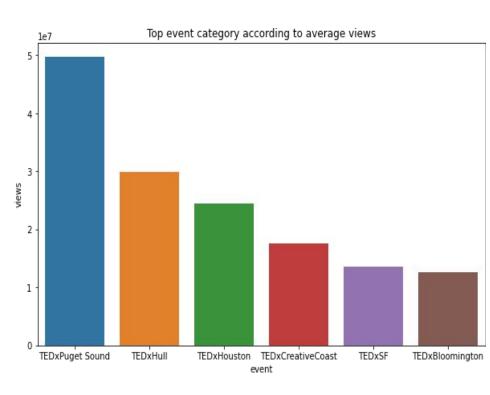
Speakers of most popular video

**Top Speakers by total Views** 



### **Events with Views:**



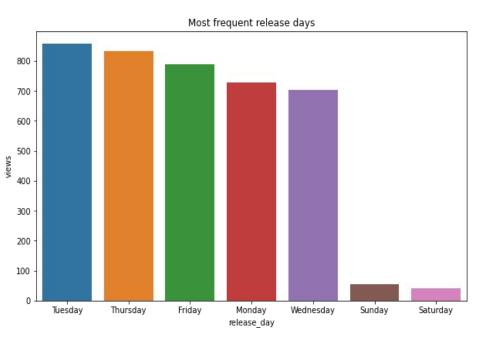


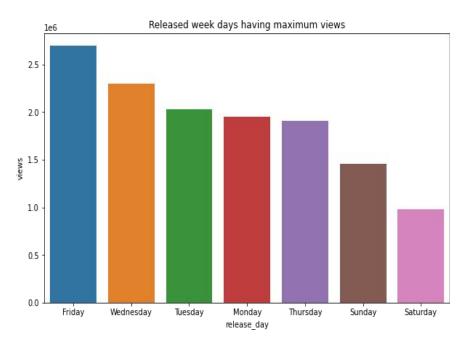
**Most Frequent event category** 

**Top Events by Average Views** 

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# **Published Days with Views:**





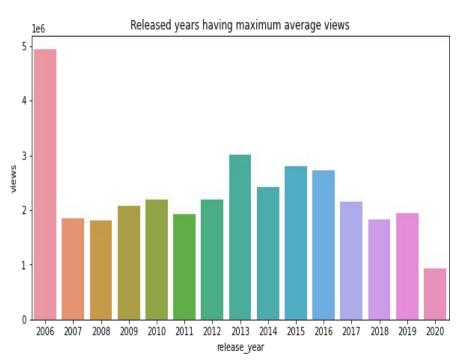
**Frequent Released Days** 

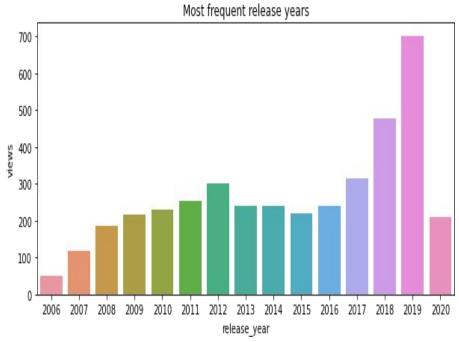
**Released Days by avg Views** 

Friday release is impacting the views of the video



### **Published Year with Views:**





Released Year with Max average views

**Most Frequent Released Year** 

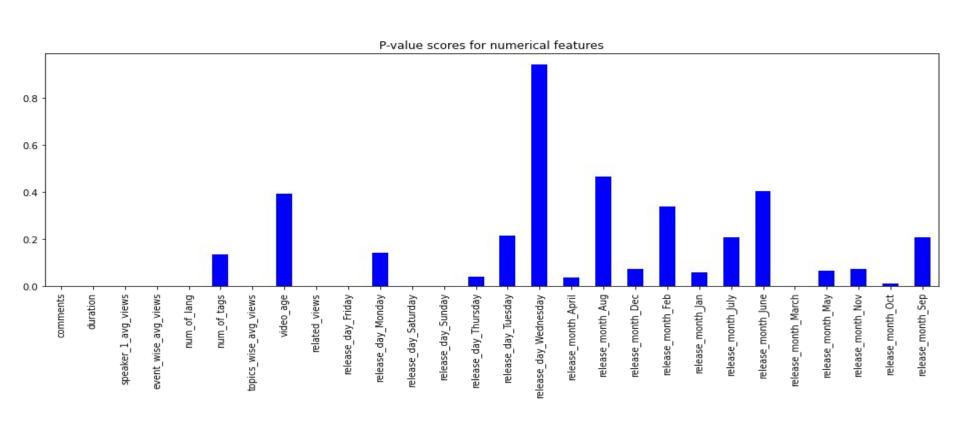


# **Feature Engineering**

- Speaker\_avg\_views
- Event\_wise\_avg\_views
- Related\_views
- Topic\_wise\_avg\_views
- Num\_of\_languages
- Num\_of\_tags
- Release\_day
- Release\_month
- Video\_age



# Features selection(f regression):





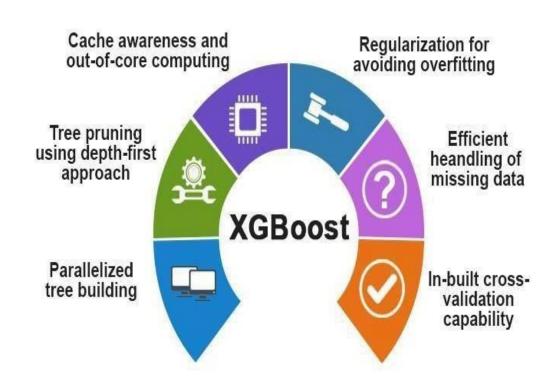
# Models used:

- XGBoost Regressor
- Extra Trees Regressor
- Random Forest Regressor



# **XGBoost Regressor:**

- Criterion = MAE
- R\_Square for train= 0.9
- R\_Square for test= 0.83
- MAE train = 164091.33
- MAE test= 226944.86
- RMSE train= 315411.38
- RMSE test= 454270.75





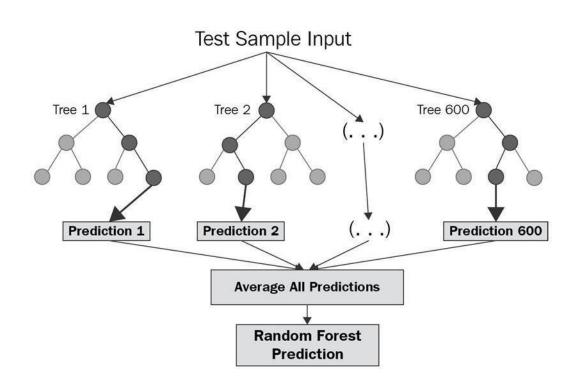
# **Extra Trees Regressor:**

- Criterion = MAE
- R\_Square for train= 0.79
- R\_Square for test= 0.83
- MAE train = 207304.04
- MAE test= 204793.75
- RMSE train= 497317.34
- RMSE test= 484832.84



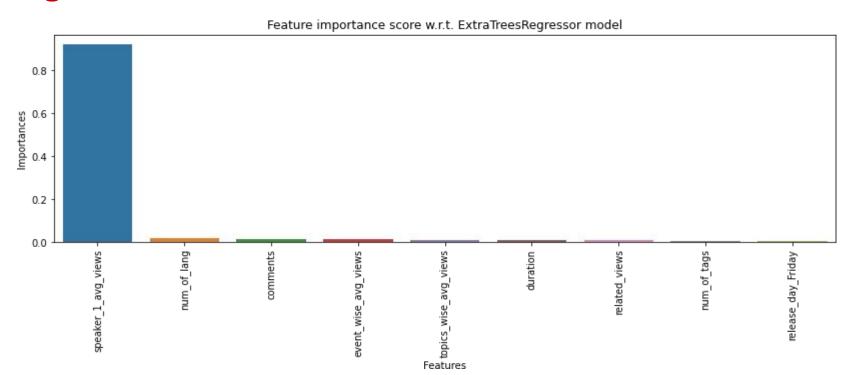
# **Random Forest Regressor:**

- Criterion = MAE
- R\_Square for train= 0.80
- R\_Square for test= 0.80
- MAE train = 186583.31
- MAE test= 191844.53
- RMSE train= 485371.33
- RMSE test= 488927.13



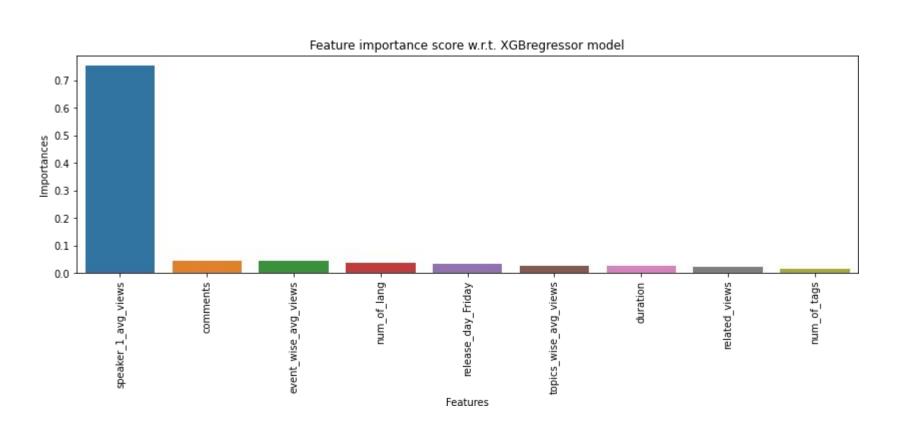


# Feature importance wrt Extra Trees Regressor:



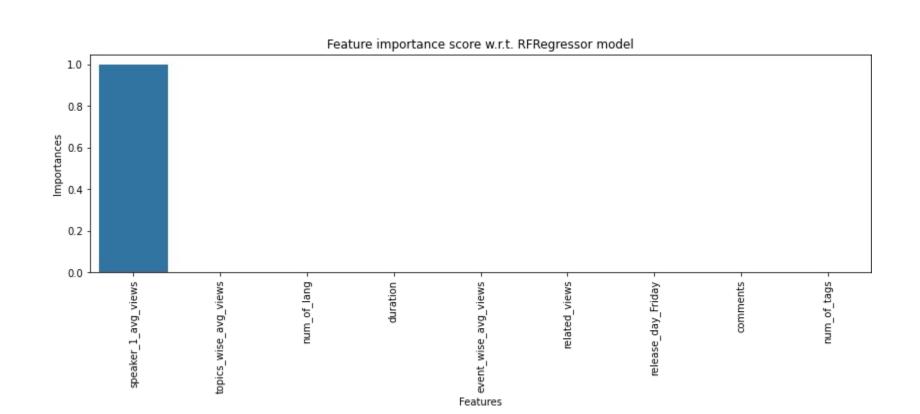


# **Feature importance wrt XGBoost Regressor:**



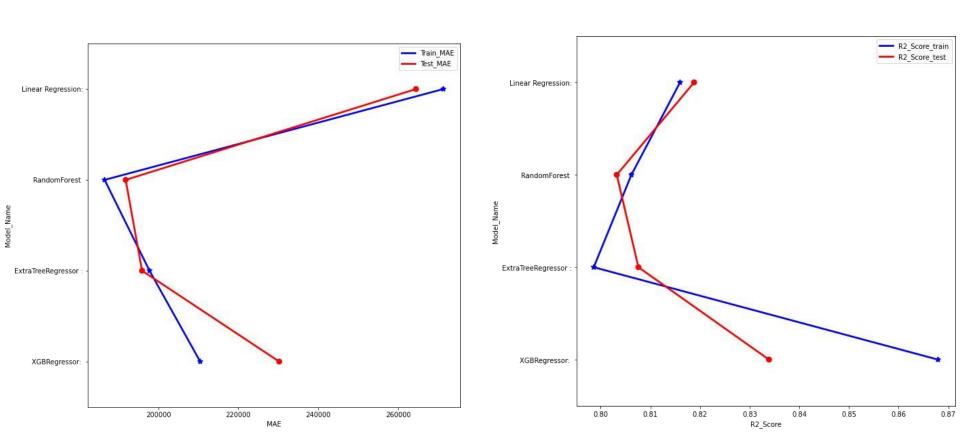


# Feature importance wrt Random Forest Regressor:





# **Model Comparison:**





# Which model did we choose and why?

- Out of all these models RandomForestRegressor is the best performer in terms of MAE.
- MAE is the best deciding factor because it isn't affected by outliers.
- MAE is linear and RMSE is quadratically increasing.



# Challenges

- Dataset have lots of textual and categorical data having high ordinal number. So the conversion to meaningful numerical data was a challenge.
- Treating the outliers in numerical features.
- Generation of new features which needs to be added in the model.
- Choosing the right features for modelling.
- Choosing the right models to get the best scores.



#### **Conclusion**

- We build a predictive model, which could help TED in predicting the views of the talks uploaded on the TEDx website.
- TED can increase their views and popularity by increasing videos on sections like
  Technology and Science.
- TED can tackle the sectors like Music by inviting more popular speakers in this sectors like 'OK GO' in this category.