

Capstone Project-2

Ted Talk Views Prediction

ML Supervised Regression

Individual Project:

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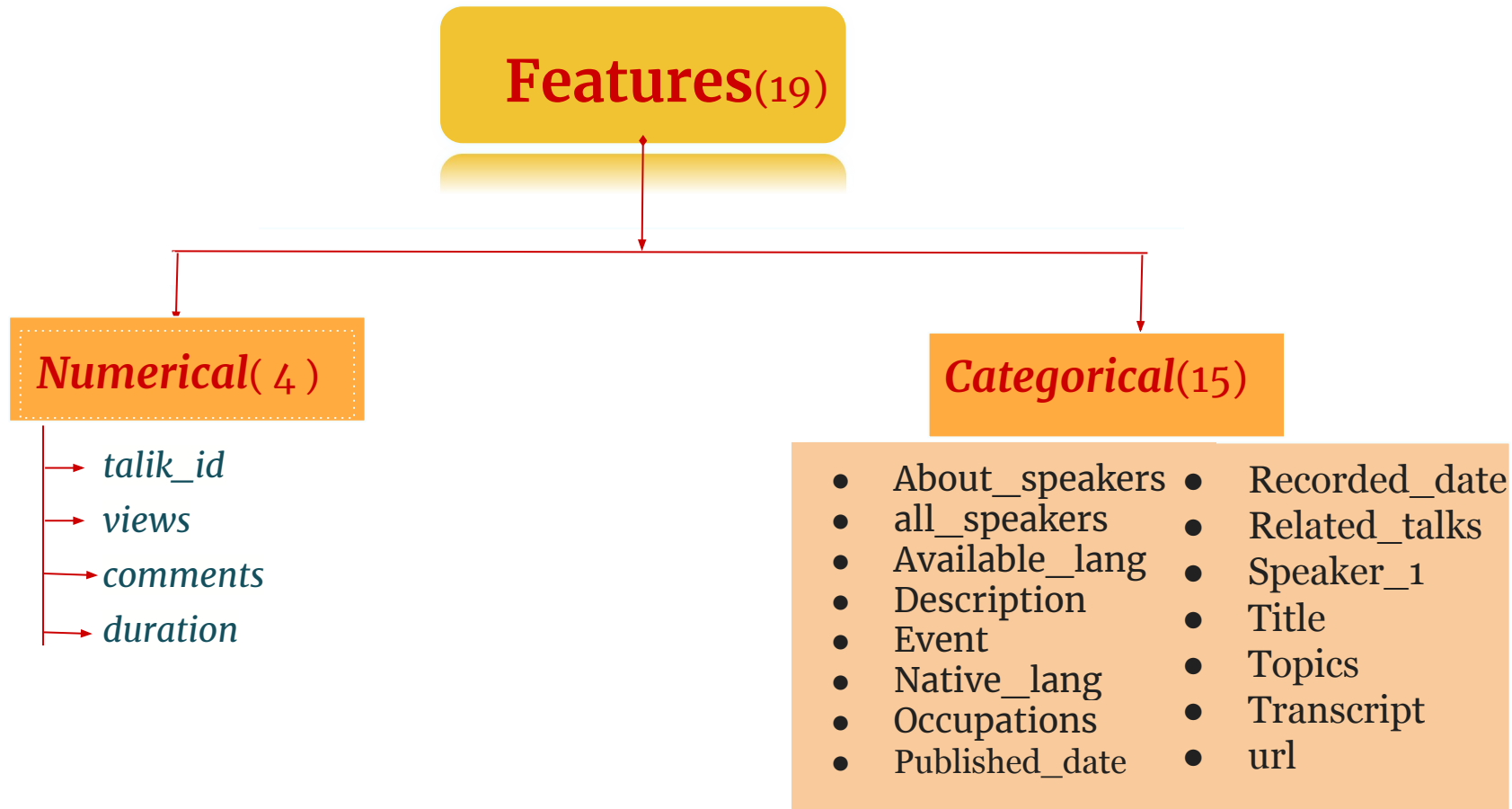
Problem Statement

AI

Prediction of the views of the videos uploaded on the TEDx website.



Let's see the features'



Basic Data Exploration

- This dataset is having 4005 observations & 19 features.
- Most of the features are categorical .
- No duplicate values.

Dataset Shape: (4005, 19)

```
>>> <class 'pandas.core.frame.DataFrame'>
Int64Index: 4005 entries, 1 to 62794
Data columns (total 18 columns):
 #   Column                Non-Null Count  Dtype
---  -
 0   title                 4005 non-null   object
 1   speaker_1            4005 non-null   object
 2   all_speakers          4001 non-null   object
 3   occupations           3483 non-null   object
 4   about_speakers        3502 non-null   object
 5   views                 4005 non-null   int64
 6   recorded_date         4004 non-null   object
 7   published_date        4005 non-null   object
 8   event                 4005 non-null   object
 9   native_lang           4005 non-null   object
10   available_lang        4005 non-null   object
11   comments              3350 non-null   float64
12   duration              4005 non-null   int64
13   topics                4005 non-null   object
14   related_talks         4005 non-null   object
15   url                   4005 non-null   object
16   description           4005 non-null   object
17   transcript            4005 non-null   object
dtypes: float64(1), int64(2), object(15)
memory usage: 116.5 MB
```

Data Exploration(NaN values)

	Feature_Name	Missing	Uniques	%age of missing values
11	comments	655	601	16.35
3	occupations	522	2049	13.03
4	about_speakers	503	2977	12.56
2	all_speakers	4	3306	0.10
6	recorded_date	1	1334	0.02
0	title	0	4005	0.00
16	description	0	4005	0.00
15	url	0	4005	0.00
14	related_talks	0	4005	0.00
13	topics	0	3977	0.00
12	duration	0	1188	0.00
9	native_lang	0	12	0.00
10	available_lang	0	3902	0.00
1	speaker_1	0	3274	0.00
8	event	0	459	0.00
7	published_date	0	2962	0.00
5	views	0	3996	0.00
17	transcript	0	4005	0.00

NaN

- 16% NaN values are present in *comments*
- 13% NaN values are present in *occupations*
- 12.5% NaN values are present in *about_speakers*

Unique value

Most of the columns except **native_lang** are containing unique values.

Data Processing

- Initially the datatype of ***published_date*** , ***recorded_date*** was in string format, i have used pandas ***to_datetime*** function to convert the datatype

```
published_date    object
recorded_date     object
dtype: object
```



```
published_date    datetime64[ns]
recorded_date     datetime64[ns]
dtype: object
```

- Created ***month*** , ***day*** , ***year*** columns based on ***published_date*** column

	published_date	month	year	day
talk_id				
92	2006-06-27	Jun	2006	27
110	2007-04-14	Apr	2007	14

Data Processing

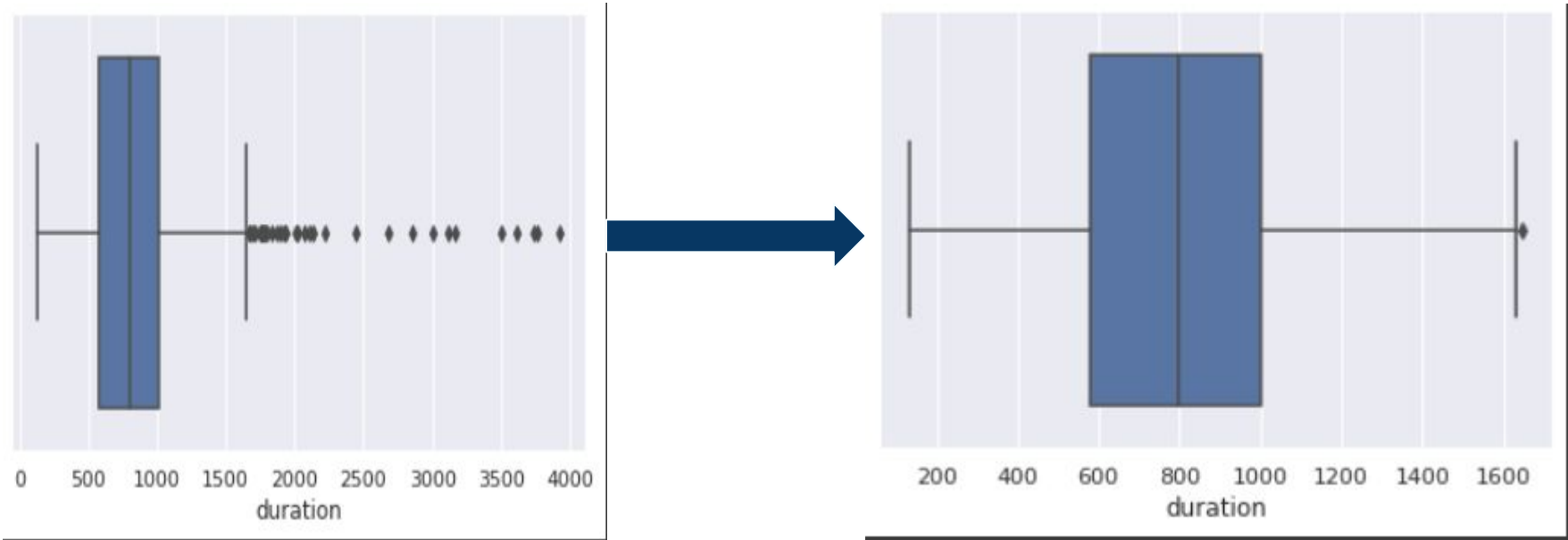
- Created `time_since_published` column based on `published_date` & `current_date`

	<code>published_date</code>	<code>time_since_published</code>
<code>talk_id</code>		
64	2006-09-06	4983 days
45	2006-08-08	5012 days

- Created `daily_views` column based on `views` & `time_since_published_date`

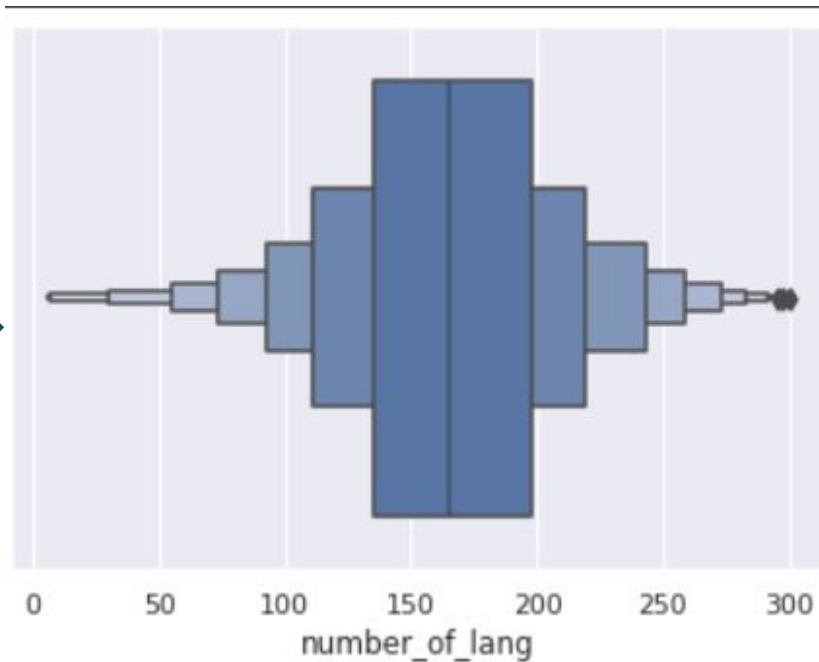
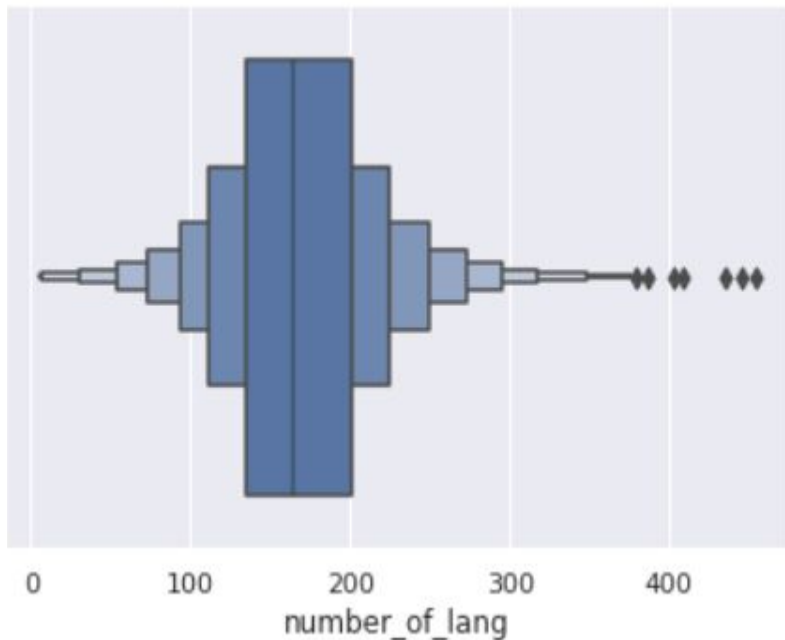
	<code>published_date</code>	<code>time_since_published</code>	<code>views</code>	<code>daily_views</code>
<code>talk_id</code>				
820	2010-04-07	3674 days	2248059	611
60	2007-02-09	4827 days	1214012	251
2588	2016-09-26	1310 days	2712894	2069

Removing Outliers



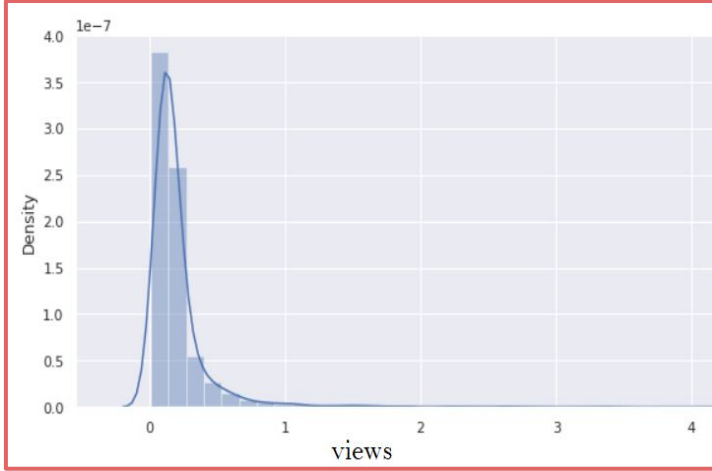
*Replaced outliers with mean value of **duration***

Removing Outliers

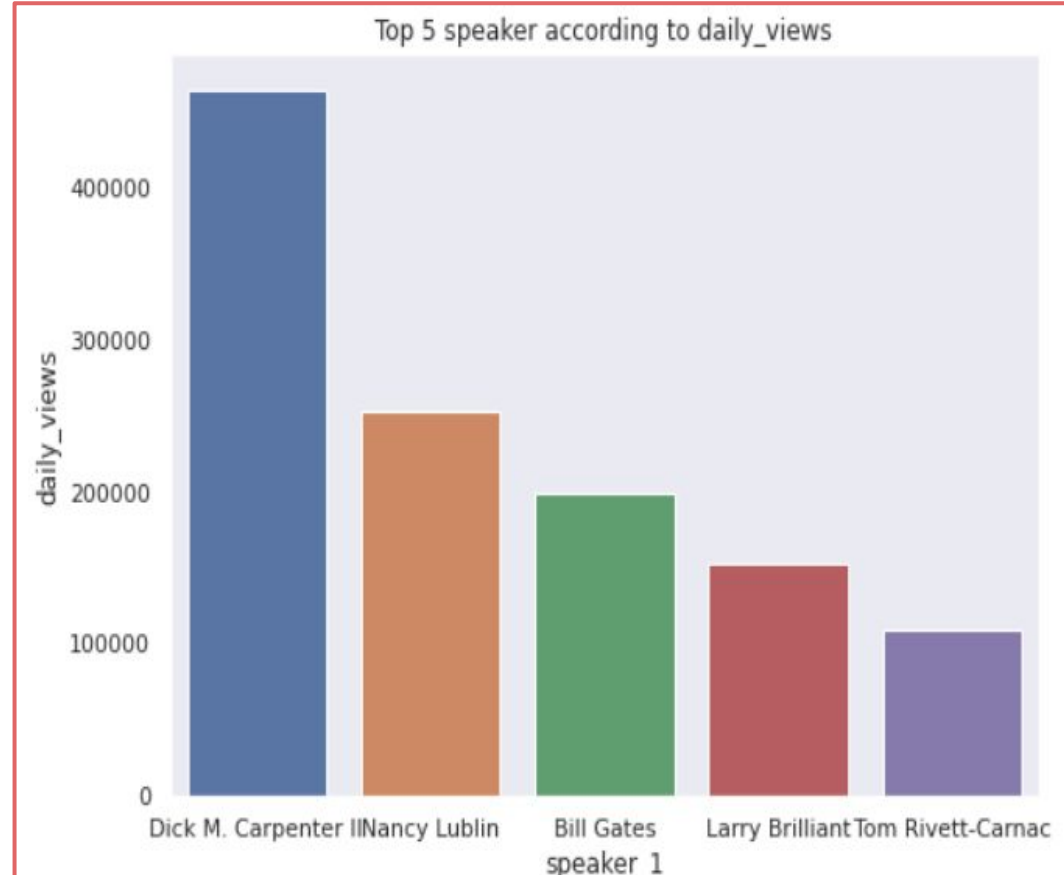


*Replaced outliers with mean value of **number_of_languages***

Visualization

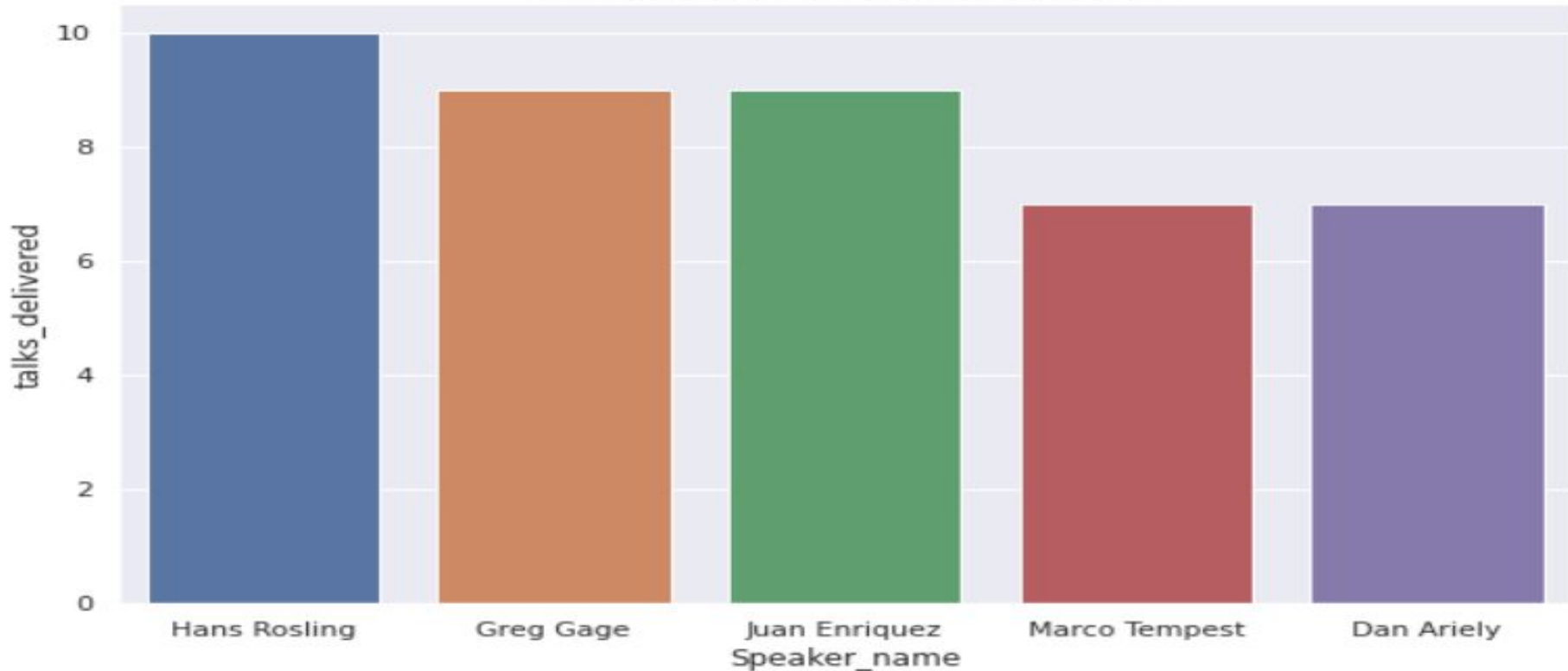


Views is positively skewed which basically means there are very few talks available which got a lot of views



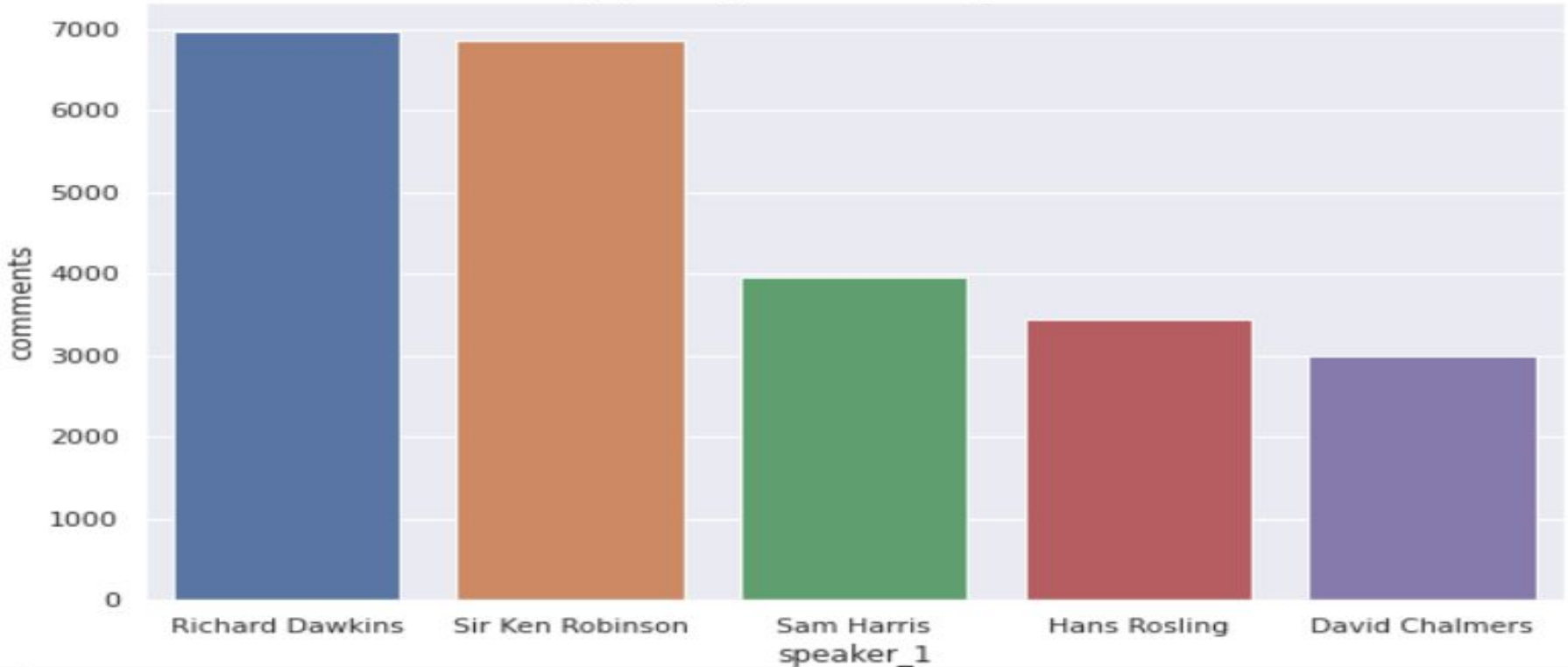
Visualization

(Top 5 speakers who delivered most talks)



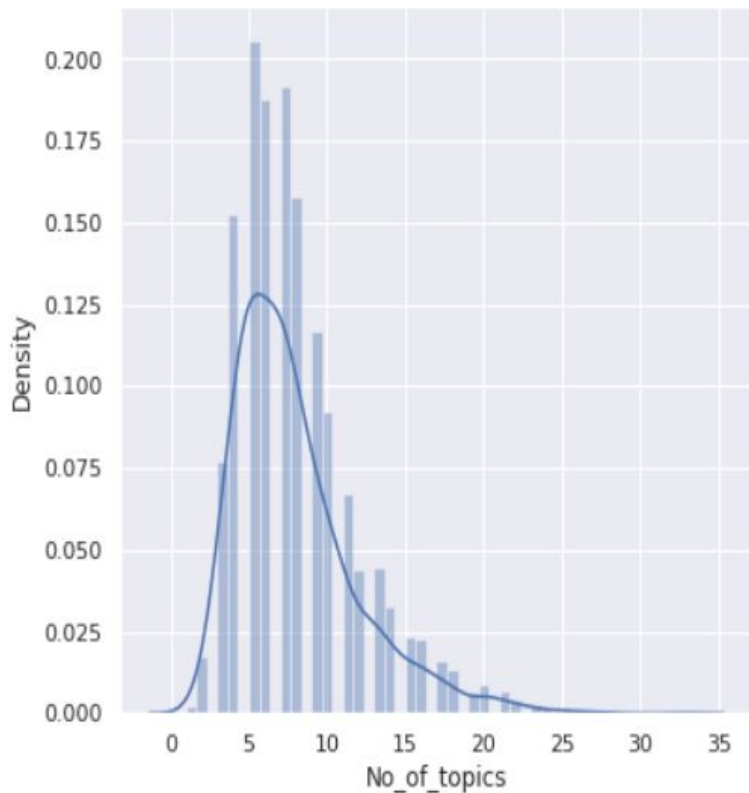
Visualization

(Most popular speakers according to Comments)

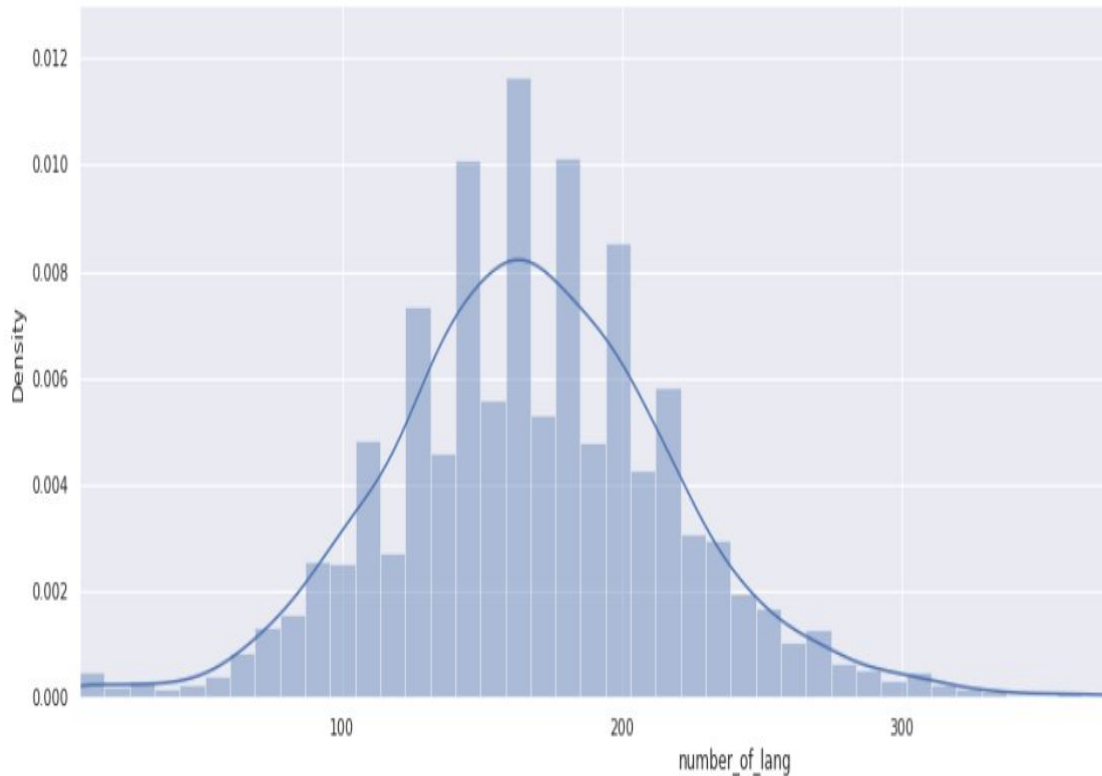


Visualization

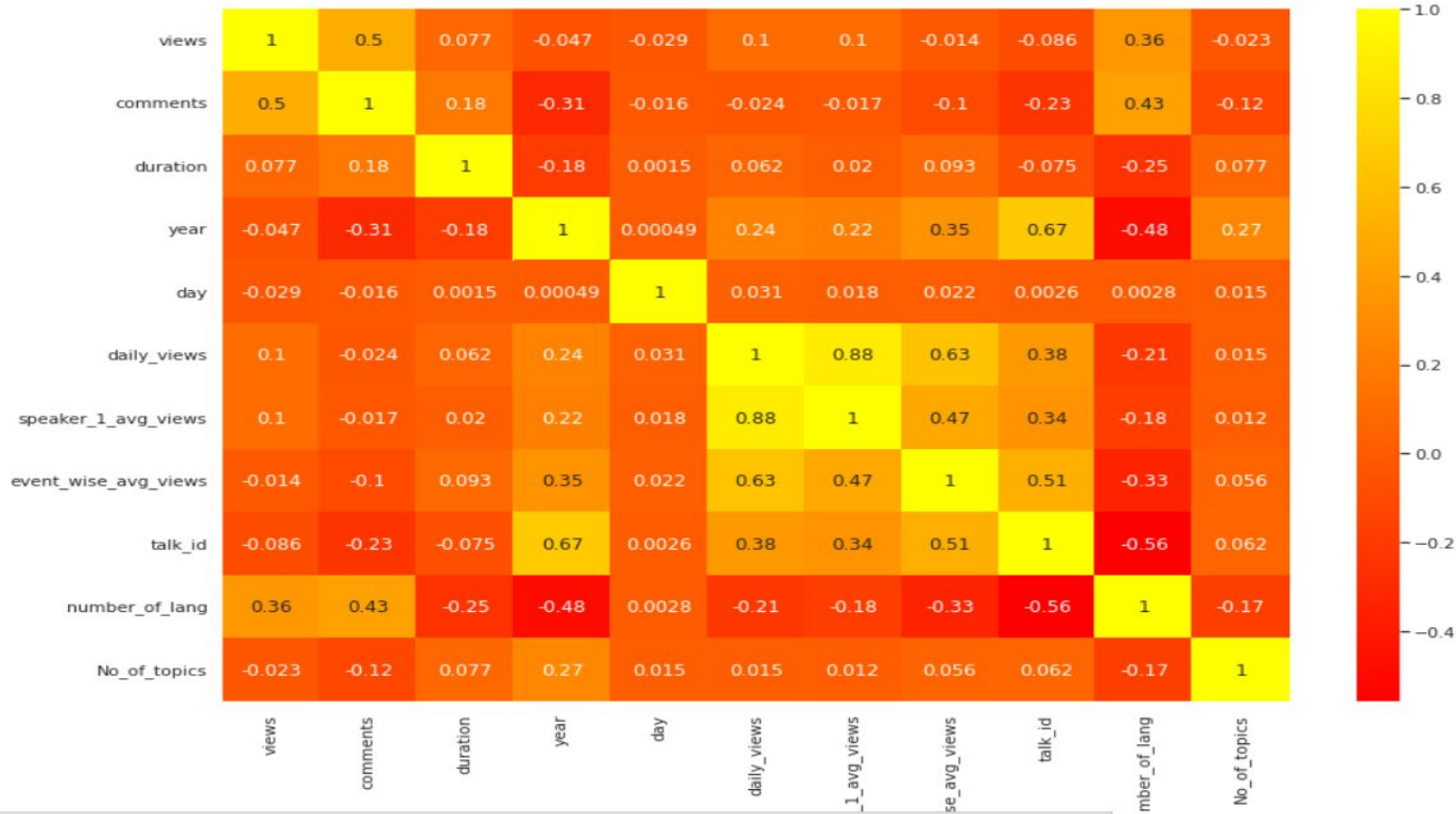
Density plot for no of topics per talk



Density plot for no of languages per talk



Correlation



We can conclude that daily_views column is highly correlated with Speaker_1_avg_views , event_wise_avg_views,

Feature removing

- Most of the speakers delivered their talk in **english**

	en	es	fr	hi	pt	it	ko	ja	de	ar	pt-br	zh-cn
native_lang	3306	15	7	2	1	1	1	1	1	1	1	1

- Removed unnecessary features like

'talk_id'	'title'	'speaker_1'	'all_speakers'
'occupations'	'about_speakers'	'views'	'recorded_date'
'published_date'	'event'	'native_lang'	'available_lang'
'topics'	'related_talks'	'url'	'description'
'transcript'			

Data Preparation

- ***Independent features :-***

comments, duration, time_since_published, month, year ,day, Speaker_1_avg_views ,
event_wise_avg_views, Number_of_lang , No_of_topics , topics_wise_avg_views

- ***Dependent feature :- daily_views***
- Splitted data into 80:20 ratio
- Used **StandScaler**

Let's compare those models

	Name	MAE_train	MAE_test	R2_Score_train	R2_Score_test	RMSE_Score_train	RMSE_Score_test
6	GradientBoostingRegressor:	380.283699	759.061577	0.994977	0.399657	857.067785	6248.226254
7	XGBRegressor:	429.726238	680.309046	0.993294	0.766738	990.303692	3894.743516
4	RandomForest	921.695436	839.076255	0.168246	0.335713	11029.234762	6572.562452
3	KNeighborsRegressor:	1031.112739	909.538141	0.541709	0.921037	8186.886733	2266.042923
1	Lasso:	1271.992955	1205.618639	0.859364	0.703730	4535.204569	4389.356547
2	Ridge:	1272.276531	1205.799311	0.859363	0.703867	4535.205410	4388.337808
0	Linear Reg.:	1272.640632	1206.337301	0.859364	0.703543	4535.203672	4390.738157
5	ExtraTreeRegressor :	1528.927152	1371.692837	0.147758	0.305693	11164.243764	6719.433676

We have chosen MAE and not RMSE as the ***deciding factor*** of our model selection

Hyperparameter Tuning

	Name	MAE_train	MAE_test	R2_Score_train	R2_Score_test	RMSE_Score_train	RMSE_Score_test
0	XGBRegressor_without_hyper	429.888181	680.397815	0.993294	0.766717	990.312172	3894.911052
1	XGBRegressor_with_hyper	102.932939	645.449187	0.999718	0.766717	203.092063	3929.319520

- Used **GridSearchCV** to do hyperparameter tuning
- Hyperparameters I have used :-
 - gamma
 - Learning_rate
 - max_depth
 - n_estimators

Conclusion

- Most of the columns are categorical
- After hyper parameter tuning, we have prevented overfitting
- Out of all these models *XGBRegressor* is the best performer in terms of MAE & r2 score.
- In all the features *speaker_1_avg_views* is most important this implies that speakers are directly impacting the views.
- R2_score for the final model is 0.99 (train data) & 0.76 (test data)
- Most of the talks are available in English.
- Hans Rosling delivered most of the talks .
- According to the number of comments Richard Dawkins is the most popular speaker.
- Top 3 most popular speakers are Dick M. Carpenter, Nancy Lublin(CEO of Crisis Text Line) , Bill Gates.

