

**EDA on Netflix Data** 

NumPy (or Numpy) is a Linear Algebra Library for Python, the reason it is so important for Data Science with Python is that almost all of the

NumPy



filtered\_cast=filtered\_cast.to\_frame() filtered cast.columns=['Actor'] actors=filtered\_cast.groupby(['Actor']).size().reset\_index(name='Total Content') actors=actors[actors.Actor !='No Cast Specified'] actors=actors.sort\_values(by=['Total Content'], ascending=False) actorsTop5=actors.head() actorsTop5=actorsTop5.sort\_values(by=['Total Content']) actorsTop5 **Actor Total Content** 23624 27 Om Puri 15541 Julie Tejwani 28 30303 Takahiro Sakurai 30 26941 Rupa Bhimani 31 2612 39 Anupam Kher From the above plot, it is derived that the top 5 actors on Netflix are: Anupam Kher Om Puri Shah Rukh Khan Takahira Sakurai Boman Irani **Analyzing Content on Netflix:** The next thing to analyze from this data is the trend of production over the years on Netflix: df1=dff[['type','release\_year']] df1=df1.rename(columns={"release\_year": "Release Year"}) df2=df1.groupby(['Release Year','type']).size().reset\_index(name='Total Content') df2=df2[df2['Release Year']>=2010] df2 type Total Content **Release Year** 95 2010 Movie 154 40 96 2010 TV Show 97 2011 Movie 145 98 2011 TV Show 40 99 2012 Movie 173 100 2012 TV Show 64

101 2013 225 Movie 102 2013 TV Show 63 103 2014 Movie 264 104 2014 TV Show 88 105 398 2015 Movie 106 2015 TV Show 162 107 2016 Movie 658 2016 TV Show 108 244 109 2017 767 Movie 110 2017 TV Show 265 111 2018 767 Movie 2018 TV Show 380 112 113 2019 Movie 633 114 2019 TV Show 397 115 2020 517 Movie TV Show 2020 436 116 117 2021 Movie 277 118 2021 TV Show 315 df1=dff[['type','release\_year']] dfl=df1.rename(columns={"release\_year": "Release Year"}) df2=df1.groupby(['Release Year','type']).size().reset\_index(name='Total Content') df2=df2[df2['Release Year']>=2010] fig3 = px.line(df2, x="Release Year", y="Total Content", color='type',title='Trend of content produced over the fig3.show() Trend of content produced over the years on Netflix 800

type Movie 700 TV Show 600 Total Content 500 400 300 200 100

2010 2012 2014 2016 2018 2020 Release Year The above line graph shows that there has been a decline in the production of the content for both movies and other shows since 2018. At last, to conclude our analysis, I will analyze the sentiment of content on Netflix: **TextBlob** Natural Language Processing for Beginners: Using TextBlob

1. About TextBlob? 2. Setting up the System 3. Having a go at NLP tasks using TextBlob 1. Tokenization Noun phrase extraction 3. POS-Tagging 4. Words inflection and lemmatization 5. N-grams 6. Sentiment Analysis 4. Other cool things to do with TextBlob 1. Spelling correction 2. Creating a short summary of a text

3. Translation and language detection 5. Text classification using TextBlob 6. Pros and Cons **About TextBlob?** TextBlob is a python library and offers a simple API to access its methods and perform basic NLP tasks. A good thing about TextBlob is that they are just like python strings. So, you can transform and play with it same like we did in python.

Below, I have shown you below some basic tasks. Don't worry about the syntax, it is just to give you an intuition about how much-related TextBlob is to Python strings. 6 - Sentiment Analysis Sentiment analysis is basically the process of determining the attitude or the emotion of the writer, i.e., whether it is positive or negative or neutral. The sentiment function of textblob returns two properties, polarity, and subjectivity. Polarity is float which lies in the range of [-1,1] where 1 means positive statement and -1 means a negative statement. Subjective sentences

generally refer to personal opinion, emotion or judgment whereas objective refers to factual information. Subjectivity is also a float which

We can see that polarity is 0.8, which means that the statement is positive and 0.75 subjectivity refers that mostly it is a public opinion and

fig4 = px.bar(dfx, x="Release Year", y="Total Content", color="Sentiment", title="Sentiment of content on Netf]

Sentiment

Negative Neutral

Positive

>> Analytics Vidhya is a great platform to learn data science.

dfx=dfx.groupby(['Release Year','Sentiment']).size().reset\_index(name='Total Content')

2014

2012

2016

Release Year

So the above graph shows that the overall positive content is always greater than the neutral and negative content combined.

2018

2020

Sentiment(polarity=0.8, subjectivity=0.75)

dfx=dfx.rename(columns={'release year':'Release Year'})

lies in the range of [0,1].

print (blob)

blob.sentiment

not a factual information.

**if** p==0:

elif p>0:

else:

fig4.show()

1200

1000

800

600

400

200

Total Content

In [24]:

**Sentiment Analysis** 

dfx=dff[['release year','description']]

p=testimonial.sentiment.polarity

dfx.loc[[index,2],'Sentiment']=sent

Sentiment of content on Netflix

for index,row in dfx.iterrows(): z=row['description'] testimonial=TextBlob(z)

sent='Neutral'

sent='Positive

sent='Negative'

dfx=dfx[dfx['Release Year']>=2010]

Let's check the sentiment of our blob.