# Movies and Video Series recommendation using emotional information.

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# 1. Objective:

Creating dataset using given Emotional dataset and IMDB dataset to building recommended system.

## 2. About datasets

#### 2.1 Emotional Dataset:

Emotional dataset contains 3659 records of data. It contains emotional scores for the movies and series episodes. It contains scores for 20 different emotions (Aggressiveness, Anger, Anticipation, Awe, Contempt, Disgust, Disapproval, Fear, Joy, Love, Negative, Optimism, Positive, Remorse, Sadness, Submission, Surprise, Trust, AFINN (-4&-5), Ero).

#### 2.2 IMDB Dataset:

IMDB dataset contains data about every movie, Tv-series, episodes, Tv-shows etc, it contains all information about crew, production, writer, director, actress, rating, start and end year etc.

There were 7 files in IMDB dataset. They were name.basics.tsv, title.akas.tsv, title.basics.tsv, title.crew.tsv, title.episode.tsv, title.principals.tsv, title.ratings.tsv. Each file contains huge data. For sample 'names.basics.tsv' file is opened in virtual code studio and shown below.

```
nconst primaryName birthYear deathYear
nm0000001 Fred Astaire 1899
nm0000002 Lauren Bacall 1924
                                          1987
                                                    sound track, actor, \verb|miscellaneous| t t 0050419, t t 0031983, t t 0072308, t t 0053137
                                           2014 actress, soundtrack tt0037382, tt0117057, tt0071877, tt0038355
nm00000002
                                           \verb|N| actress, soundtrack, \verb|music_department| tt0049189, tt0057345, tt0059956, tt0054452|
              John Belushi 1949
                                           1982
                                                  actor,soundtrack,writer tt0080455,tt0072562,tt0078723,tt0077975
              Ingmar Bergman 1918
                                           2007
                                                    writer.director.actor tt0060827.tt0083922.tt0050976.tt0050986
              Ingrid Bergman
                                                    actress,soundtrack,producer tt0036855,tt0038109,tt0034583,tt0038787
                                                    \verb|actor|, soundtrack|, producer & tt0040897, tt0033870, tt0034583, tt0043265|| \\
              Marlon Brando 1924
                                           2004
                                                    actor, soundtrack, director
                                                                                     tt0070849,tt0068646,tt0078788,tt0047296
                                                    actor, soundtrack, producer
                                                                                     tt0061184,tt0057877,tt0087803,tt0059749
              Richard Burton 1925
                                                    actor,soundtrack,director tt0029870,tt0042041,tt0035575,tt0031867
                ames Cagney
              Gary Cooper 1901
                                               actor,soundtrack,producer tt0027996,tt0035896,tt0044706,tt0034167
                                              actress,soundtrack,make_up_department tt0056687,tt0042192,tt0031210,tt0035140
soundtrack,actress,producer tt00
bandu tt0054084,tt0048317,tt0060463
actress,soundtrack tt0031381,tt0029843,tt0040806,tt0041452
              Bette Davis 1908
nm0000013 Doris Day 1922 2019
nm0000014 Olivia de Havilland 1916
```

Figure 1: Sample IMDB Data

For our project we were using 3 files from IMDB

- 1. name.basics.tsv for IMDB id(tconst), titles and titleType
- 2. title.crew.tsv for directors Id(nconst)
- 3. title.crew.tsv for directors' names.

# 3. Cleaning Dataset

## 3.1 Removing unwanted data.

The emotional dataset contains many unwanted columns which were not required for the project, so they were removed manually. The removed data columns were as below LoveA, OptimismA, AgressivenessA, ContemptA, SubmissionA, AweA, DissaprovalA, RemorseA, erowords and nrc words.

## 3.2 Adding IMDB ids to Emotional data.

Titles of the emotional data movie and tv-series are not same as titles in IMBD. Emotional data titles contain -, .srt for every movie and some titles are separated by ".", "white space" as shown in the below image.



Figure 2 About Emotional data titles

So, titles were not used to merge two datasets and there is no other way to merge. So IMDB ID's (tconst) are assigned to emotional data titles manually.

	IMBD-tconst	0	Score	Normalise	Year
0	tt12389600	12 Angry Men 1957 en	8.9	87.23404	1957
1	tt0047478	1954 Seven Samurai	8.6	80.85106	1954
2	tt12409982	1984 Once Upon a Time in America II etait une fois en Ameriq	8.3	74.46809	1984
3	tt0062622	2001 A Space Odyssey 1968 REMASTERED 720p BluRay X264	8.3	74.46809	1968
4	tt1187043	3 Idiots 2009 Hindi 1080p Blu Ray x264 DD 5 1 MSubs HDSector	8.3	74.46809	2009
5	tt0066921	A Clockwork Orange (1971) 1080p H 264 Multi (moviesbyrizzo	8.3	74.46809	1971

Figure 3 Emotional dataset with IMDB id's

# 3.3 Cleaning TV-series data of Emotional dataset

Emotional data contains data of Tv-series episodes but there is only one IMDB id for complete series and for recommended system we cannot give recommendations by episode. To overcome this problem, the emotional scores of each Tv-series episode were calculated manually

Boardwalk Em	8.6	80.8511	2010	2148.241	840.017241	50.8930586	63.306953	13.21022296	22.9929	34.914888	13.589	22.92337
Boston Med St	8.6	80.8511	2010	3164	1083	37.6941607	60.903251	4.059095106	15.8834	52.965668	23.818	7.9841509
Boston Med St	8.6	80.8511	2010	2980	1000	35.8562612	71.5881	3.297	15.5658	49.44	8.5985	13.363291
Boston Med St	8.6	80.8511	2010	3170	1110	39.8471547	62.561953	3.96036036	17.7613	44.438984	10.328	25.494355
Boston Med St	8.6	80.8511	2010	3114	1108	41.3942191	68.149225	4.959386282	12.8067	52.924188	12.934	16.317461
Boston Med St	8.6	80.8511	2010	3162	1131	41.9065255	64.043421	3.886825818	18.037	46.666667	12.671	19.460766
Boston Med St	8.6	80.8511	2010	2689	971	42.8410383	64.078342	9.054582904	15.7257	55.315045	13.775	20.238825
Boston Med St	8.6	80.8511	2010	2673	990	45.3770831	60.704819	2.22020202	16.1207	44.565657	5.7902	19.850403
Boston Med St	8.6	80.8511	2010	2603	1033	52.6215241	68.11875	6.383349468	8.92136	51.42128	7.3989	8.3706055
Boston Med St	8.6	80.8511	2010	2944.375	1053.25	42.1922458	65.018483	4.727600245	15.1028	49.717186	11.914	16.384982

Figure 4 Average of emotional scores for Tv-series

The average scores of tv-series, movies data were taken into another file to build a clean and reduced dataset.

# 4. Merging Emotional dataset and IMDB dataset

# 4.1 Merge with *name.basics.tsv*

'Titles' and 'title-type' were added and 'titles' were removed from emotional dataset in name.basics.tsv using the below code.

```
data_with_titles = pd.merge(ed,title, on ='tconst')
```

Figure 5. Data Merge.

The results were as below.

,	Trust	AFINN(-4&-5)	Ero	titleType	primaryTitle
	34.706143	0.040480	10.181415	tvEpisode	12 Angry Men (1957)
,	30.772156	19.278291	20.043469	movie	Seven Samurai
ļ	41.001855	14.087598	18.451172	tvEpisode	Once Upon A Time In America (1984)
j	52.073334	2.011218	4.221965	movie	2001: A Space Odyssey
;	62.270554	3.627473	14.311616	movie	3 Idiots

 $\textit{Figure 6}. \ \textit{emotional data with IMDB titles}$ 

## 4.2 Merge with title.crew.tsv

Using tconst data was merged in emotional data with title.crew.tsv to get director id's (ncosnt) for respective movies and tv series.



Figure 7. emotional data with nconst

## 4.3 Merge with title.crew.tsv

Merging emotional data with title.crew.tsv on nconst to get director names for respective movies and tv-series.

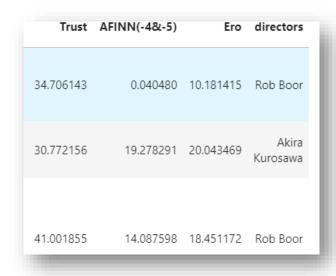


Figure 8. emotional data with director names

# 5. Recommended System:

Considering the dataset, 4 different types of recommended systems were built as below.

- 1. Recommendations Based on Emotion type
- 2. Recommendations based on Emotion and Time period
- 3. Recommendations based on Director Name.
- 4. Recommendations Based on Movie name.

## 5.1 Recommendations Based on Emotion type

The aim of this recommendation type is movies, tv-series suggestions based on emotion and count.

```
sort_by_emotion("Agressiveness", 5)
   Score
                 primaryTitle titleType Agressiveness
                                                                   director
                  City Lights
                                   movie
23
     8.5
                                             47.370690 ['Charles Chaplin']
10
     8.3
                      Amadeus
                                   movie
                                              40.008172
                                                           ['Milos Forman']
0
     8.9 12 Angry Men (1957) tvEpisode
                                              32.022145
                                                               ['Rob Boor']
6
     8.2
                 A Separation
                                   short
                                              29.772674
                                                               ['Yalan Hu']
     8.2
                     Scarface
                                              26.732432
                                                          ['Brian De Palma']
                                   movie
```

Figure 9. Recommendations Based on Emotion type

All the emotions were captured in a graphical representation as below, which displays 5 movies, tv-series and it shows 'Agressiveness' as the main emotion for the taken data.

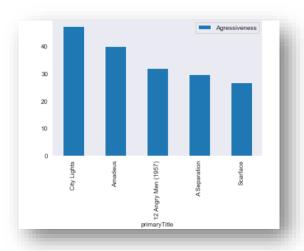


Figure 10. Bar graph for Recommendations Based on Emotion type

## 5.2 Recommendations based on Emotion and Time period

The aim of this recommendation type is movies, tv-series suggestions based on emotion, time period (like 2010 - 2015) and count.

```
sort_by_year_emotion("Anger", 1985,1990,5)
   Score Year
                         primaryTitle titleType
                                                    Anger
36
     8.3 1987
                   Full Metal Jacket
                                         movie 33.475033
     8.6 1990
                           Goodfellas
39
                                         movie 31.188208
40
     8.4 1988 Grave of the Fireflies
                                        movie 31.141707
                      Cinema Paradiso
21
     8.4 1988
                                        movie 22.406215
9
     8.3 1986
                               Aliens
                                        movie 20.884473
                 director
36
      ['Stanley Kubrick']
39
      ['Martin Scorsese']
        ['Isao Takahata']
40
21 ['Giuseppe Tornatore']
9
        ['James Cameron']
```

Figure 11. Recommendations based on Emotion and Time period

All the emotions were captured in a graphical representation as below, which displays 5 movies, tv-series, time period is between 1985 – 1990 and it shows 'Anger' as the main emotion for the taken data.

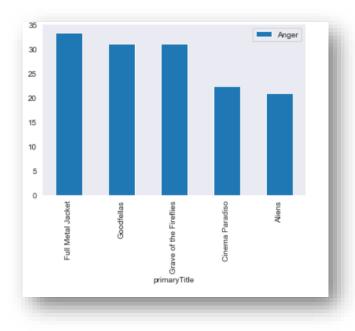


Figure 12. Bar graph for Recommendations based on Emotion and Time period

## 5.3 Recommendations based on Director Name:

This recommendation type is based on director and the sample is as below.

```
director('[\'Rob Boor\']')

Score Year primaryTitle titleType director
0 8.9 1957 12 Angry Men (1957) tvEpisode ['Rob Boor']
2 8.3 1984 Once Upon A Time In America (1984) tvEpisode ['Rob Boor']
```

Figure 13. Recommendations based on Director name.

Here we passed "Rob Boor" as director name and we got result of movies director by Rob boor.

## 5.4 Recommendations Based on Movie name

This recommendation type is based on the specific emotion which has highest percentage in the given movie.

```
by_title('Scarface')
                     primaryTitle titleType
                                               Trust
   Score Year
     8.4 1957
                   Paths of Glory
                                     movie 95.731117
60
     8.2 1962 Lawrence of Arabia
                                     movie 91.630259
48
5
     8.3 1971 A Clockwork Orange
                                     movie 86.796997
                        Gladiator
37
     8.5 2000
                                     movie 78.046007
23
     8.5 1931
                      City Lights
                                     movie 74.755836
```

Figure 14. Recommendations Based on Movie name

As we seen in the above picture, the movie name was given, the code helps to finds the highest emotion score among all emotion scores and give results based on the emotion.