
YouTube comments on Oscar-nominated movie trailers

— Data Mining and Business Intelligence —

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Introduction

Database pertains to Youtube comments on Oscar nominated movie trailers for 4 select few movies:

- **32070** rows x **8** columns
- Comments on movie trailers with date timestamp
- Movies selected are as below:
 - **Dunkirk** - History/War
 - **The Shape of Water** - Romance/Fantasy
 - **Phantom Thread** - Romance/Drama
 - **Get Out** - Psychological Thriller/Horror



Data pre-processing:

| timestamp | comment_text | likes | has_replies | number_of_replies | Movies | Time | Date |
|-----------|--|-------|-------------|-------------------|---------|----------|-----------------------|
| 1.52E+12 | OSCAR | 0 | FALSE | 0 | Dunkirk | 11:40:00 | Monday, March 5, 2018 |
| 1.52E+12 | Christopher Nolan + Hans Zimmer = perfection | 0 | FALSE | 0 | Dunkirk | 11:40:00 | Monday, March 5, 2018 |
| 1.52E+12 | This deserves an Oscar and not shape of you. | 0 | FALSE | 0 | Dunkirk | 11:40:00 | Monday, March 5, 2018 |
| 1.52E+12 | This movie was like watching a loud plane fly over you for 3 hours...there was r | 0 | FALSE | 0 | Dunkirk | 11:40:00 | Monday, March 5, 2018 |
| 1.52E+12 | Get fucked Nolan, no oscar for you. | 0 | FALSE | 0 | Dunkirk | 11:40:00 | Monday, March 5, 2018 |
| 1.52E+12 | HARRYYYYYYÄ°Ä°ËœÄ°Ä°ËœÄ°Ä°ËœÄ°Ä° like if you want to see it for harry | 0 | FALSE | 0 | Dunkirk | 11:40:00 | Monday, March 5, 2018 |

In the process of cleaning the dataset:

- Combine all 4 selected movies in one dataset
- And keep all 4 selected movies separated
- Change timestamp to Time and Date
- Added one more column to name the movies

| Name / | Role |
|-------------------|----------|
| comment_text | Text |
| Date | Time ID |
| has_replies | Rejected |
| likes | Rejected |
| Movies | Label |
| number_of_replies | Rejected |
| Time | Rejected |
| timestamp | Rejected |

SAS Enterprise Miner Workstation process:

- Importing the file:
 - All 4 selected movies in one dataset
 - Separated 4 selected movie name in 4 dataset
- In Text Parsing, we added the table: *type of emotions* in Filter tab for separated movie dataset

| | |
|----|-----------|
| 1 | TERMS |
| 2 | accused |
| 3 | acquitted |
| 4 | adorable |
| 5 | adored |
| 6 | affected |
| 7 | afflicted |
| 8 | aghast |
| 9 | agog |
| 10 | agonized |

Exploration of Data : 4 Movies together

| Topics | | | | | |
|------------------------------------|----------|-------------|-----------------|-----------------|--------|
| Topic | Category | Term Cutoff | Document Cutoff | Number of Terms | # Docs |
| Racism | User | 0.011 | 0.094 | 154 | 1820 |
| junk | User | 0.01 | 0.147 | 11 | 0 |
| War + History | User | 0.017 | 0.07 | 245 | 2542 |
| Raw excitement emotions | User | 0.014 | 0.061 | 52 | 461 |
| Harry Styles Fanbase | User | 0.014 | 0.093 | 33 | 1154 |
| Christopher Nolan Movie Reviews | User | 0.015 | 0.084 | 44 | 1233 |
| People's Choice of Oscar Accolades | User | 0.015 | 0.079 | 52 | 1342 |
| Marine + Related Movies to SOW | User | 0.014 | 0.072 | 48 | 542 |
| Mixed Emotions | User | 0.015 | 0.067 | 62 | 753 |
| Disgust + Fear | User | 0.015 | 0.069 | 95 | 1441 |
| Movie Soundtrack | User | 0.014 | 0.061 | 28 | 449 |
| Dunkirk Actors Fanbase | User | 0.015 | 0.066 | 58 | 520 |

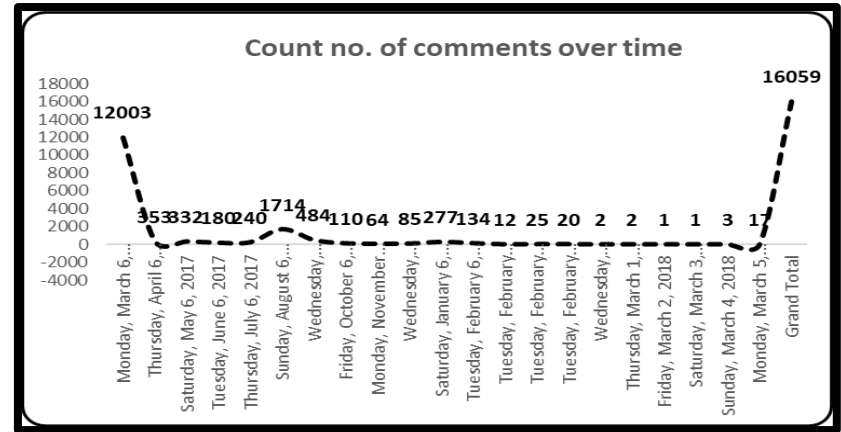
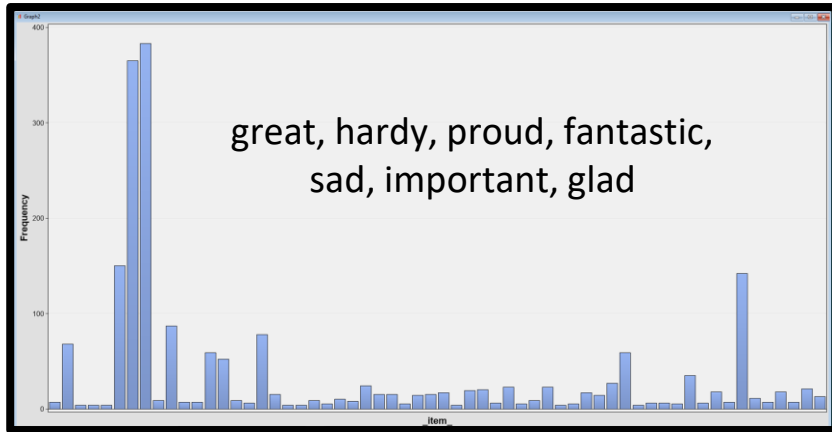
- For all 4 movies together, mixed set of reactions
- Racism is a widely talked about topic
- Nolan movies carry a huge fan base

Diving deeper into data

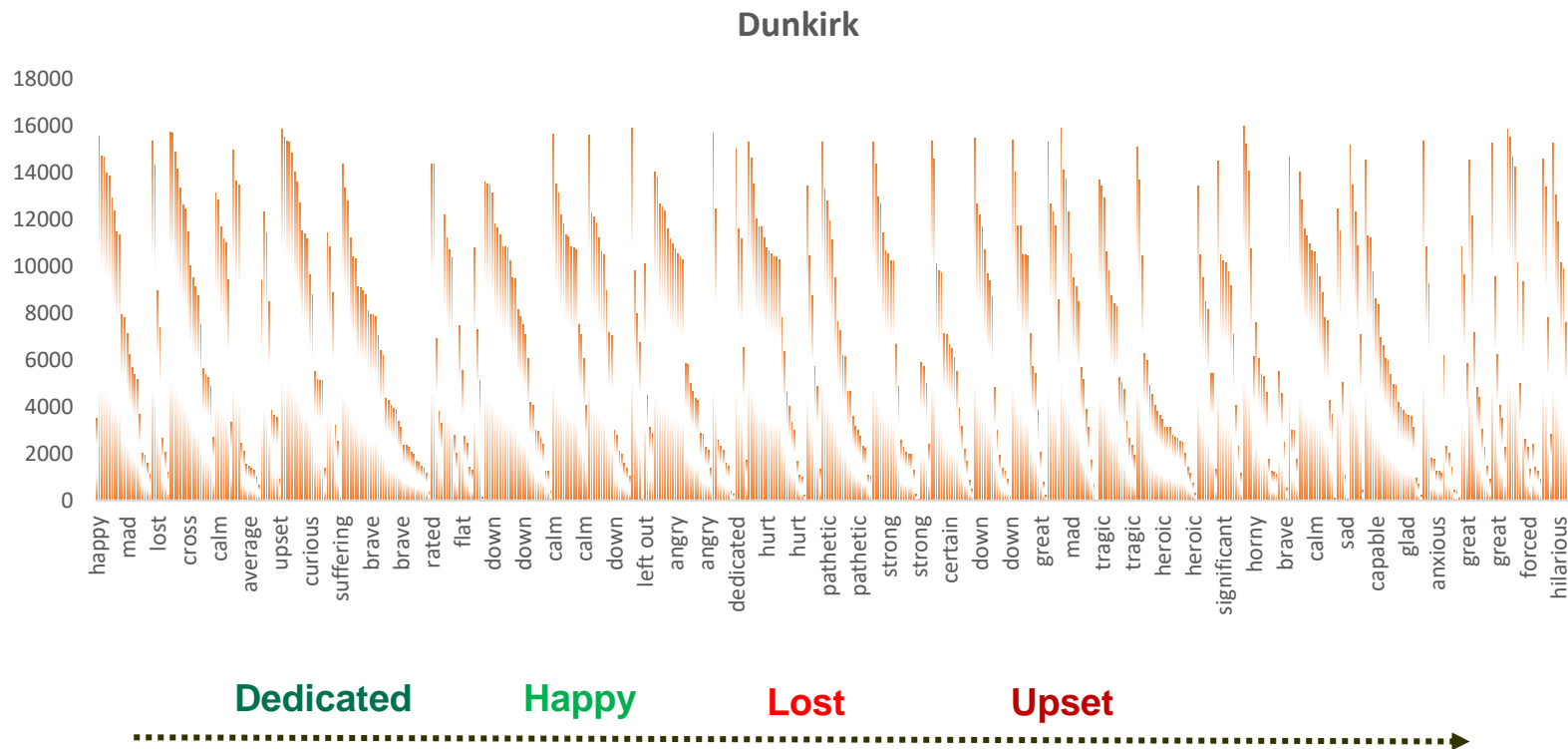
Exploration of Data - Dunkirk

Key topics

| Topic | Category |
|---|----------|
| hardy,+sad,+happy,+dark,+great | Multiple |
| +great,+happy,+dark,important,significant | Multiple |
| +great,+great,hardy,great,important | Multiple |
| hardy,hardy,+great,+dark,+great | Multiple |
| +proud,+great,+great,important,+lucky | Multiple |
| great,+great,hardy,important,+proud | Multiple |
| +sad,+proud,+dark,important,+great | Multiple |
| excited,great,+proud,+happy,hardy | Multiple |
| +happy,+dark,+sad,+proud,glad | Multiple |
| hardy,mad,+great,important,+great | Multiple |



Sentiment Analysis - Dunkirk



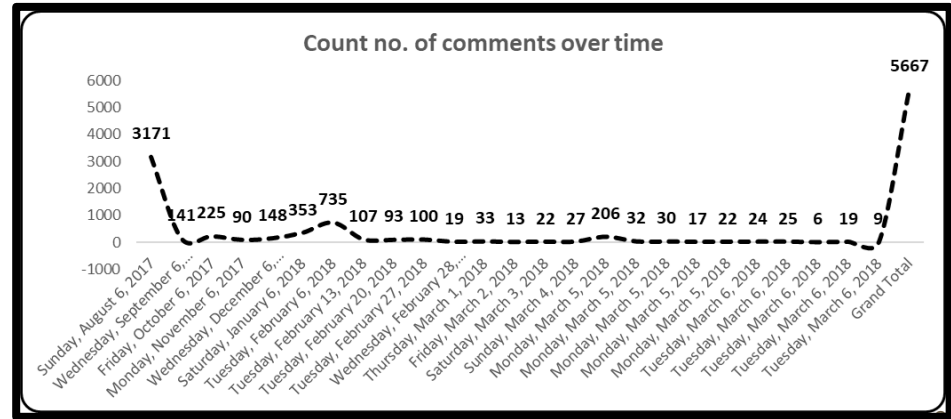
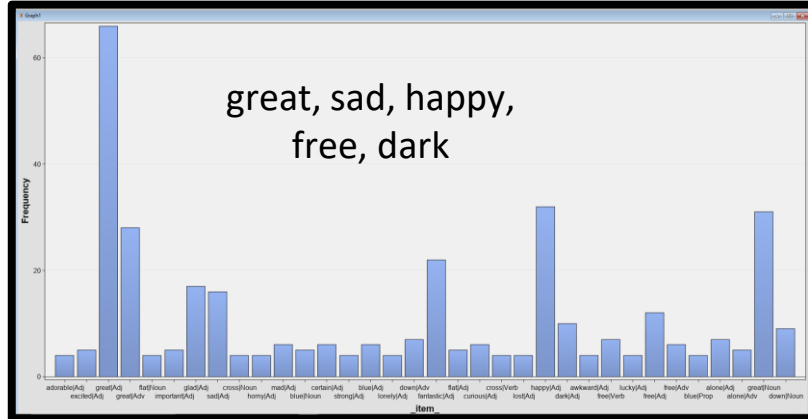
Exploration of Data - Dunkirk



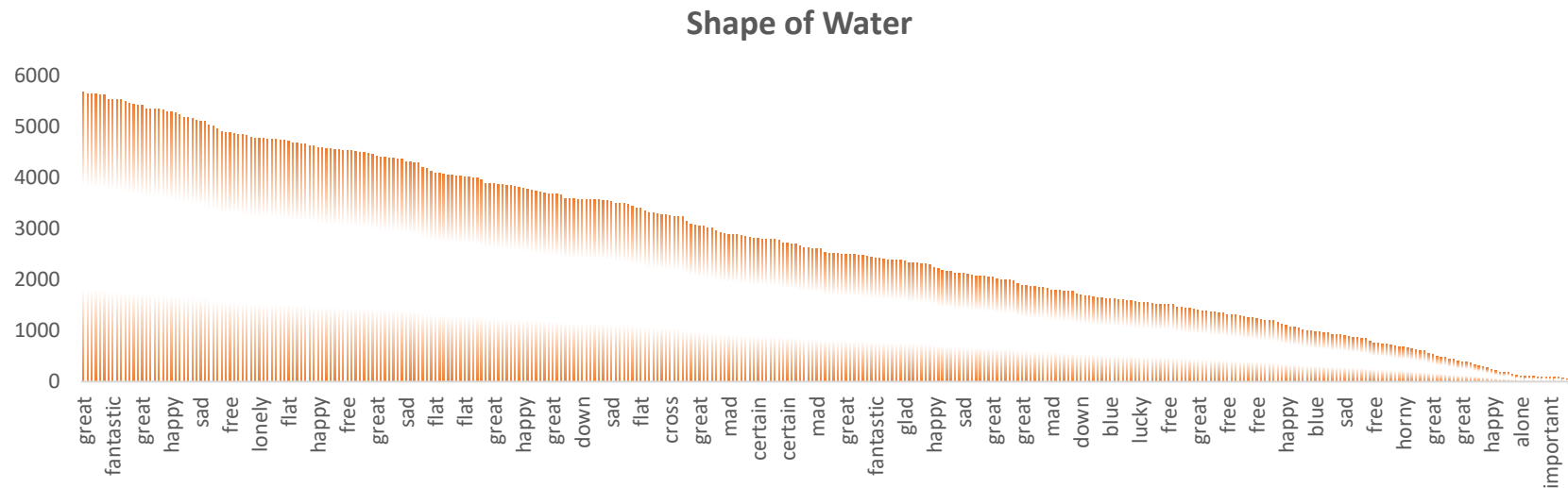
Exploration of Data - The Shape of Water

Key topics

| Topic | Category |
|---------------------------------------|----------|
| +great,cross,happy,+free,important | Multiple |
| +great,mad,great,+great,happy | Multiple |
| happy,sad,lucky,adorable,excited | Multiple |
| great,+great,happy,awkward,adorable | Multiple |
| fantastic,horny,+great,great,adorable | Multiple |
| glad,happy,great,adorable,+great | Multiple |



Sentiment Analysis - The Shape of Water



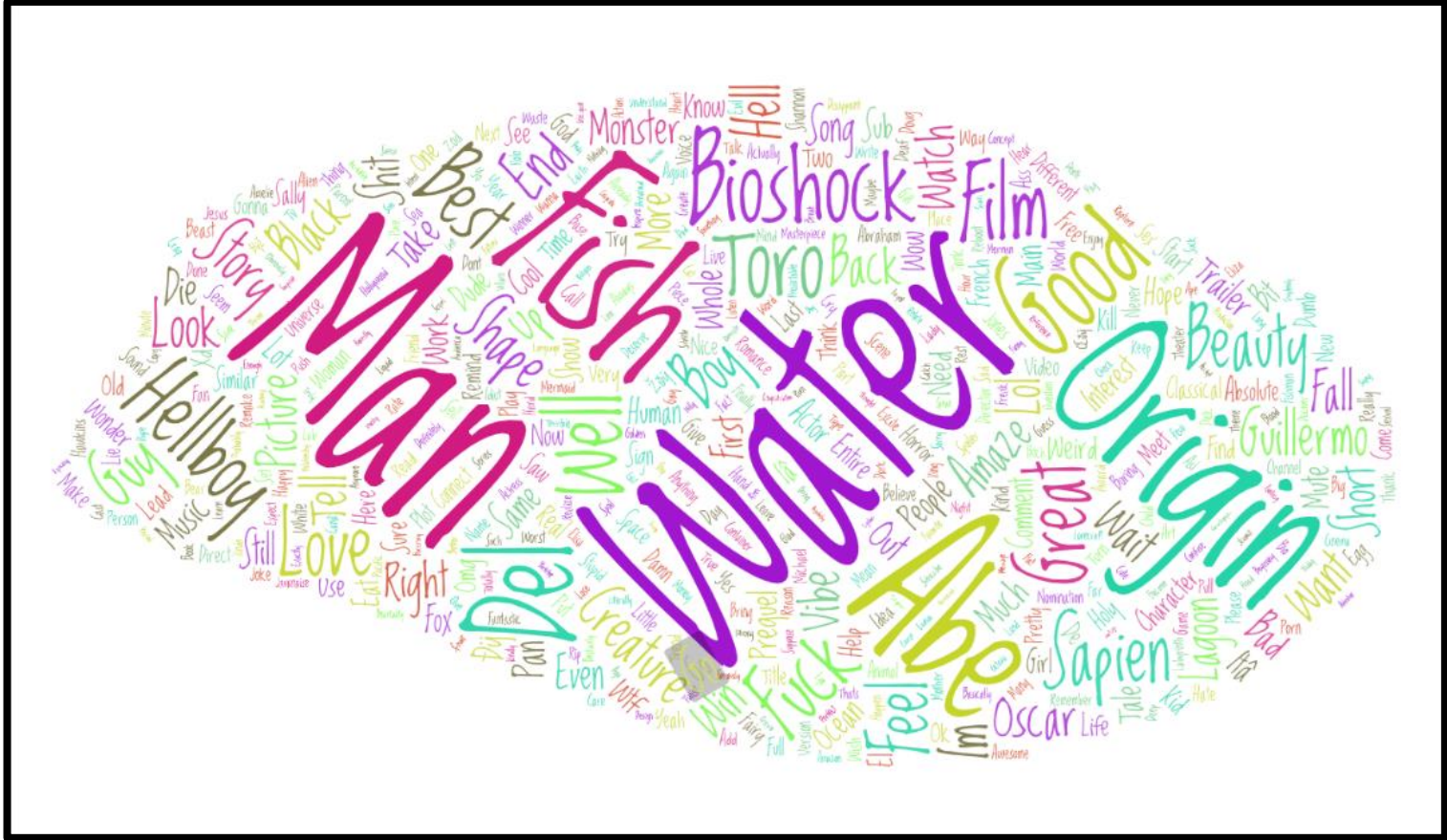
Great

Fantastic

Happy



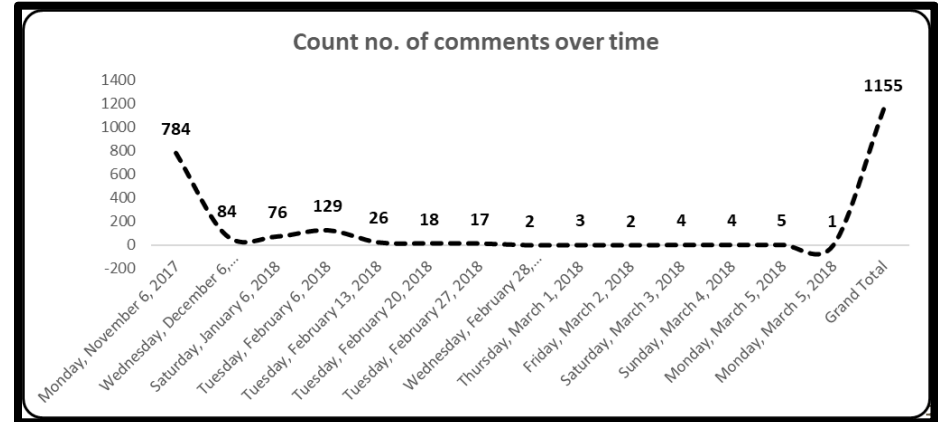
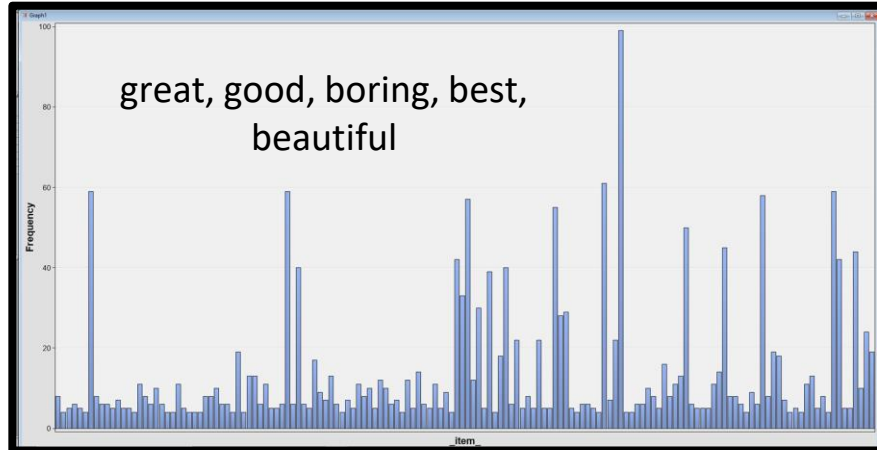
Exploration of Data - The Shape of Water



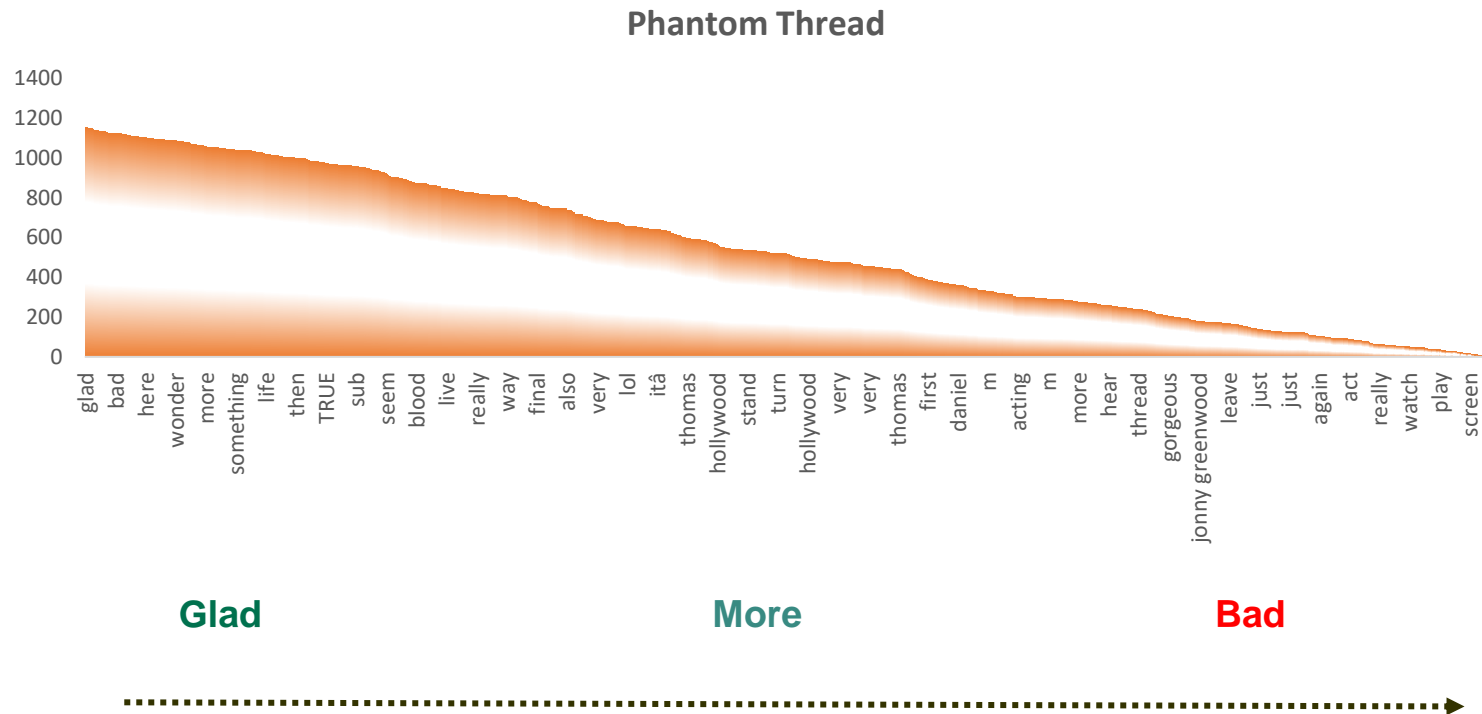
Exploration of Data - Phantom Thread

Key topics

| Topic | Category |
|--------------------------------------|----------|
| not,+do,+wait,+film,+have | Multiple |
| daniel,+day,lewis,daniel day,+retire | Multiple |
| anderson,paul,thomas,+be,+make | Multiple |
| +oscar,+go,+get,4th,bait | Multiple |
| ðy,â,thread,+movie,phantom | Multiple |
| +be,+movie,+think,blood,i | Multiple |
| +look,boring,+good,+movie,+be | Multiple |
| +movie,+see,i,+make,+want | Multiple |
| pta,ddl,+film,+see,+year | Multiple |
| +good,+actor,day-lewis,+film,+year | Multiple |



Sentiment Analysis - Phantom Thread



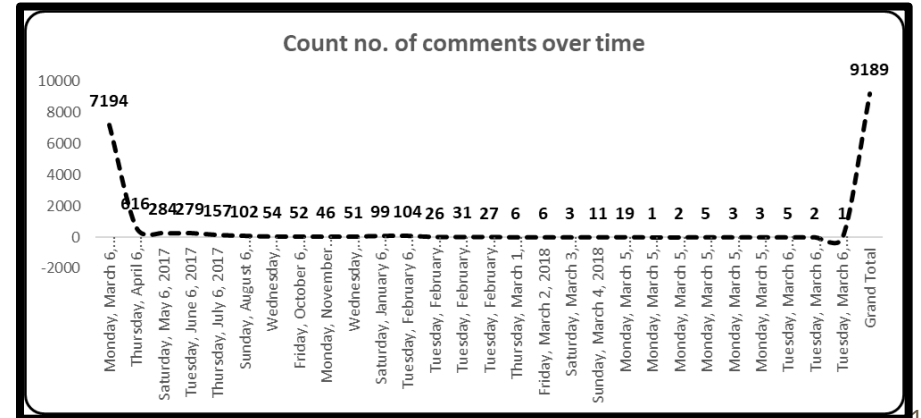
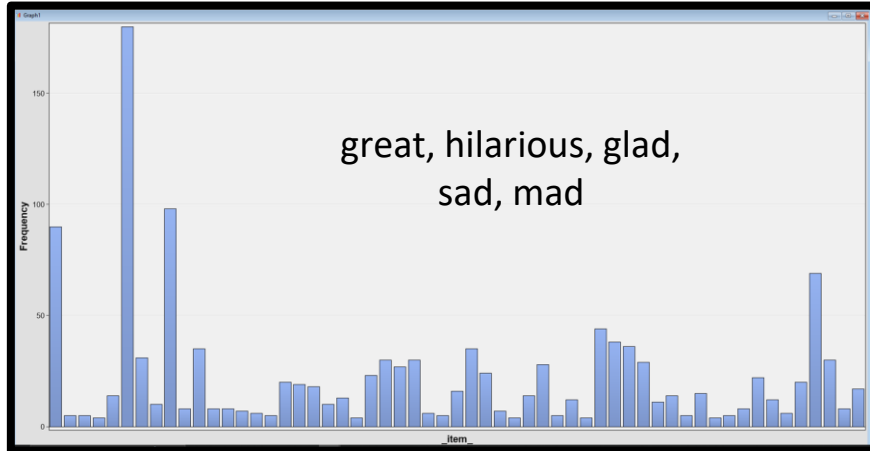
Exploration of Data - Phantom Thread



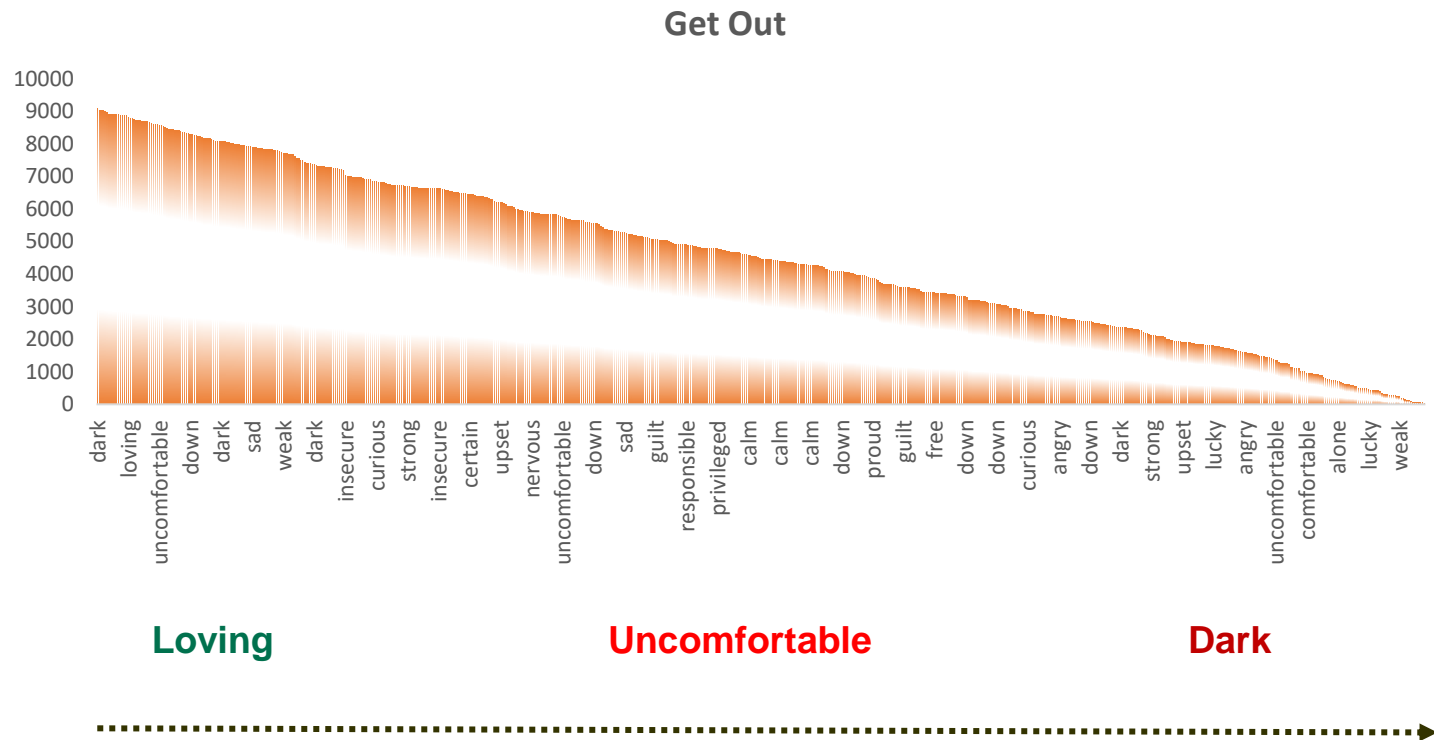
Exploration of Data - Get Out

Key topics

| Topic | Category |
|---|----------|
| +great,glad,important,alone,uncomfortable | Multiple |
| glad,great,mad,nervous,fantastic | Multiple |
| hilarious,glad,uncomfortable,curious,afraid | Multiple |
| great,+great,common,+sad,angry | Multiple |
| mad,hilarious,angry,+down,+upset | Multiple |
| nervous,uncomfortable,angry,afraid,+upset | Multiple |
| +happy,+sad,nervous,+hurt,excited | Multiple |
| great,alone,free,+great,+sad | Multiple |
| +sad,guilt,+down,+dark,+upset | Multiple |
| down,calm,+down,calm,+dark | Multiple |



Sentiment Analysis – Get Out



Exploration of Data - Get Out



Key trends and insights



Key trends and insights



People gravitate towards war and action movies



People tend to reflect negatively on topics which are sensitive and attracts a lot of flak amongst viewers if not portrayed well



Actors and Actresses can overpower the fame a movie can get and influence viewership driven by crazy fan-base to a very large extend

Business Recommendations



War and **history** are the topics of interest to the reviewers. Youtube and other media platforms can look at creating content along those lines.



A huge chunk of revenue generation depends on the actor / actress casted. **Highly acclaimed actors** earn higher viewership and hence increase revenue generation.



The degree of viewership tends to increase as there are **events** like nomination announcements, trailer launches, etc. Doing events around those time frames can **help drive revenue generation** due to higher CTR and viewerships.