

#### Research seminar HW2

Accomplished by Petr Yakovlev BASB 212, 24<sup>th</sup> March 2022



### Key task

To analyze the textual dataset (more than 200 reviews from Amazon Movies), present top-10 most frequent words and visualize some interesting statistics

### Dataset description



#### Positive reviews

101 text files each with the positive review on some movie

### Negative reviews

101 text files each with the negative review on some movie

### PySpark Initialization

### To perform word count, we do the following:

- Use findspark() and build SparkSession()
- Initialize SparkContext()
- Import multiple files through textFile()

```
Ввод [1]: # we download the findspark library and
          import findspark
          findspark.init()
          # we create Spark context in order to wo
          from pyspark.sql import SparkSession
          from pyspark.conf import SparkConf
Ввод [3]: # Spark set-up
          spark=SparkSession.builder\
              .master("local[*]")\
              .appName("WordCount")\
              .getOrCreate()
Ввод [4]: # initialize Spark context
          sc=spark.sparkContext
Ввод [5]: # import files (101 text files with posi
          file positives = sc.textFile(r"C:\Users\
          file negatives = sc.textFile(r"C:\Users\
          rdd positives = file positives.collect()
          rdd negatives = file negatives.collect()
          rdd positives
          rdd negatives
```

# Next, we clean the dataset

```
Ввод [6]: # perform function to remove punctuation and transform all words
           def lower clean str(x):
               punc='!"#$%&\'()*+,./:;<=>?@[\\]^ `{|}~-'
               lowercased str = x.lower()
               lowercased str = lowercased str.replace('<br />', '')
               for ch in punc:
                   lowercased str = lowercased str.replace(ch, '')
               return lowercased str
Ввод [7]: # we transform our instances
           file_positives = file_positives.map(lower clean str)
           file_negatives = file_negatives.map(lower_clean_str)
Ввод [8]: # we use split function to separate the words
           file positives = file positives.flatMap(lambda y: y.split(" "))
           file negatives = file negatives.flatMap(lambda y: y.split(" "))
Ввод [9]: # exclude whitespaces from the instances
           file positives = file positives.filter(lambda x: x!='')
           file negatives = file negatives.filter(lambda x: x!='')
BBog [10]: # to count the frequency of words we first need to apply transfro
           positives_count = file_positives.map(lambda word:(word,1))
           negatives count = file negatives.map(lambda word:(word.1))
```

- remove punctuation
- transform all words to lowercase
- separate the words
- exclude whitespaces

#### Word count

### We use reduceByKey() to count the most frequent words

As you can see, the most frequent words are approximately the same for negative and positive dataset except for words 'bad' for negative dataset and 'great' for positive dataset

```
Ввод [14]: # we exclude stopwords values
           positives_count = positives_count.filter(lambda x: x[1] not in stopwords).sortByKey(False)
           negatives count = negatives count.filter(lambda x: x[1] not in stopwords).sortByKey(False)
Ввод [15]: #display the result for positive datasets
           positives count.sortByKey(False).take(10)
  Out[15]: [(139, 'movie'),
             (130, 'film'),
             (113, 'one'),
             (63, 'see'),
             (58, 'good'),
             (54, 'even'),
             (49, 'great'),
             (46, 'time'),
             (44, 'like'),
             (44, 'show')]
Ввод [16]: #display the result for negative datasets
           negatives count.sortByKey(False).take(10)
 Out[16]: [(177, 'film'),
             (136, 'movie'),
             (117, 'one'),
             (77, 'like'),
             (65, 'even'),
             (64, 'horror'),
             (55, 'good'),
             (51, 'bad'),
             (51, 'really'),
             (50, 'get')]
```

```
ВВОД [17]: from pyspark.sql.types import *
           schema = StructType([StructField("frequency", IntegerType(), True),
                                 StructField("word", StringType(), True)])
BBOQ [18]: counts df positives = spark.createDataFrame(positives count, schema)
           counts df negatives = spark.createDataFrame(negatives count, schema)
ВВОД [19]: counts_df_positives.printSchema()
           counts df negatives.printSchema()
           root
             |-- frequency: integer (nullable = true)
             |-- word: string (nullable = true)
           root
             |-- frequency: integer (nullable = true)
             |-- word: string (nullable = true)
Ввод [20]: df positives = counts df positives.toPandas()
           df negatives = counts df negatives.toPandas()
Ввод [21]: df positives
  Out[21]:
                 frequency
                      139
                          movie
                      130
                      113
            3873
                        1 vells
            3874
                        1 vovo
```

3875

1 yuppie

## We present some visualisations for the dataset next

We convert pyspark object to pandas dataframe and work with it

### Word clouds



```
BBOД [125]: from textblob import TextBlob

def preprocess(ReviewText):
    ReviewText = ReviewText.str.replace("(<br/>", "")
    ReviewText = ReviewText.str.replace('(<a).*(>).*(</a>)', '')
    ReviewText = ReviewText.str.replace('(&amp)', '')
    ReviewText = ReviewText.str.replace('(&gt)', '')
    ReviewText = ReviewText.str.replace('(&gt)', '')
    ReviewText = ReviewText.str.replace('(\&gt)', '')
    ReviewText = ReviewText.str.replace('(\xa0)', '')
    return ReviewText

files_df['Review'] = preprocess(files_df['Review'])

files_df['polarity'] = files_df['Review'].map(lambda text: TextBlob(text).sentiment.polarity)
    files_df['review_len'] = files_df['Review'].astype(str).apply(len)
    files_df['word_count'] = files_df['Review'].apply(lambda x: len(str(x).split()))
```

### We also can calculate some interesting statistics with aggregated datasets with more than 200 reviews

We can show the sentiment polarity of the review with one library in python and show the reviews with negative and positive sentiment

### 5 random reviews with negative sentiment polarity



5 random reviews with the highest negative sentiment polarity:

Seriously, I can't imagine how anyone could find a single flattering thing to say about this movie, much less find it in themse lves to write the glowing compliments contained in this comment section. How many methamphetamines was Bogdonovitch on during the filming of this movie? Was he giving a bonus to the actor that spath his lines out with the most speed and least inflection or thought? The dialogue is bad, the plot atrocious, even for a "screwball" comedy, and claims that the movie is an homage to classic film comedy is about the most inane thing I've ever heard. The cinematography is below the quality and innovation of that exhibited by the worst made-for-TV movies, the acting is awful (although I get the feeling that the fault for that lies squarely in the lap of the director), and speaking of which, did I mention the direction is so haphazard and inscrutable that it defies the definition of the word? The whole thing is a terribly unfunny (even in the much-beleaguered world of so-bad-it's-funny clunkers), soul-sucking, waste of two hours of your life that you'll never get back. Be afraid, be very afraid...

An obscure horror show filmed in the Everglades. Two couples stay overnight in a cabin after being made a little uneasy by the unfriendliness of the locals. Who, or what, are the Blood Stalkers? After awhile they find out. Watch for the character of the village idiot who clucks like a chicken, he certainly is weird.

This film, which I rented under the title "Black Voodoo" should be avoided. I was expecting a blaxploitation/horror flick; but what I got was a very dull, standard "ghost extracts vegence". In this case the ghost was that of a religious cult leader who t ried to refuse treatment, but who's plea was ignored and he died in an operation. The result: his spirit posesses Nurse Sherry and forces her to commit acts of murder. The only voodoo connection was to one of the three black characters, in this case a blinded ex-football player, who's mom practiced voodoo. The film is very slow and very dull. There is a very standard ending that provides on excitement, followed by a horrificly stupid ending (warning: SPOILER)<br/>
or />In which a woman actually manages to defend herself against murder charges by saying she was possessed. This movie is slow, and bad in a non-funny, just stupefying way. Avoid it at all costs.

We brought this film as a joke for a friend, and could of been our worst joke to play. The film is barely watchable, and the acting is dire. The worst child actor ever used and Hasslehoff giving a substandard performance. The plot is disgraceful and at p oints we was so bored we was wondering what the hell was going on. It tries to be gruesome in places but is just laughable.<br/>
/>cbr />Just terrible

I found this movie really hard to sit through, my attention kept wandering off the tv. As far as romantic movies go..this one is the worst I've seen. Don't bother with it.

### 5 random reviews with positive sentiment polarity

5 random reviews with the highest positive sentiment polarity:

The great Vincent Price has done many fantastic Horror films, some of which range among the greatest genre gems of all-time. Pr ice's greatest achievements were doubtlessly his films in the 60s, with films such as Roger Corman's brilliant Poe-cycle (still the greatest Horror cycle of all-time), Michael Reeves' "Witchfinder General" (1968) or Ubaldo Ragona's "The Last Man on Earth" (1964) marking the ultimate highlights of this brilliant man's career. The films that made the man famous and thereby made him the immortal Horror icon he is, however date back to the 50s, with "House of Wax" (1953) marking his rise to stardom. "The Mad Magician" of 1954 follows a plot that is very similar to that of its successful predecessor. This is not to say, however, that this film isn't an original, delightfully macabre and absolutely wonderful gem itself. As the lines above may suggest, Vincent Price is my favorite actor, and, while I personally would not allow myself to miss anything the man has been in, none of my fel low fans of the man may miss this little gem. <br/> />r />Price stars as Don Galico (aka, Galico the Great), an underrated maste r magician and inventor of magic devices, whose boss, a sleazy businessman, stole his wife (Eva Gabor) from him. When the boss takes away one of Galico's ingenious inventions and gives it to his rival, The Great Rinaldi (John Emery), Galico snaps, and a murderous spree of revenge begins...<br/>br />cbr />cbr />Don't we love Vincent Price when he's out for revenge? Some of his most famous and greatest films such as "The Abominable Dr. Phibes" (1971) or "Theater of Blood" (1973) were about absurd and delightfully m acabre revenge murders, and this earlier film in his Horror career is another proof that no one takes revenge as Vincent Price does. This film provides a wonderfully eccentric leading role for Price, who, as always, delivers a brilliant performance, and guarantees 70 minutes of outrageously entertaining and macabre fun for every Horror fan. Another must-see for my fellow Price f

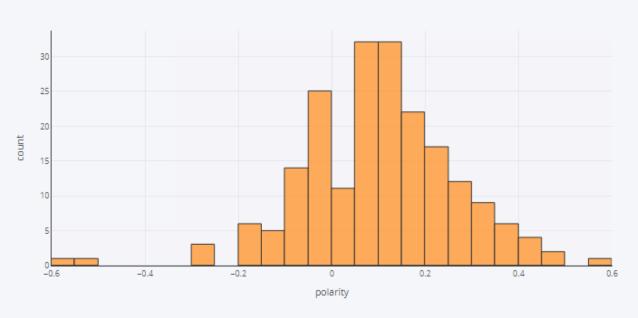
Liked Stanley & Iris very much. Acting was very good. Story had a unique and interesting arrangement. The absence of violence a nd porno sex was refreshing. Characters were very convincing and felt like you could understand their feelings. Very enjoyable movie.

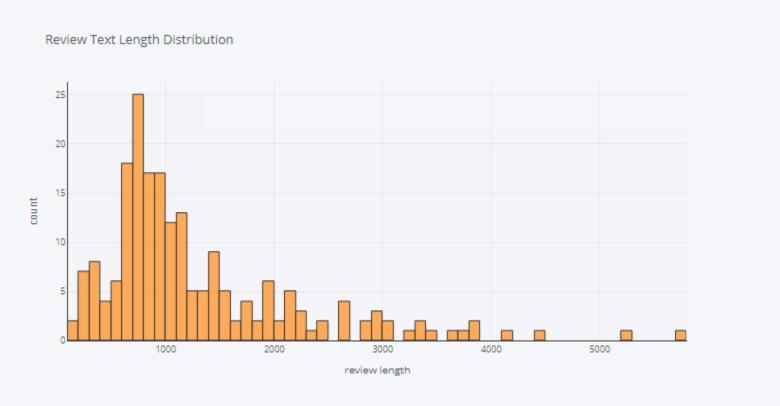
My children just happened to stop at this movie the other night and as things started to play out it really piqued my interest. I had to head out for bowling league so I had them record it for me on the dvr so I could watch the rest later. Well I just got done watching it and the front of my shirt must be soaked after crying buckets. It was an excellent movie even though I could a lmost feel the pain and anguish these girls were experiencing. And I never in a million years would have guessed the reason why Alissia had gone from this beautiful girl to an anti-social goth. This was probably WHY my shirt was soaked because I've experience that same pain that Alissia was feeling. I too would not have sought out this movie, but I'm sure glad I saw it. Very moving, very touching. Great for those who love a good drama or tear-ferker.

My first child was born the year this program came out, and I played the record album for the boys every Christmas thereafter. When the CD came out, I bought about ten copies and still give them to friends and relatives as they start families...it invariably becomes their favorite Christmas album. I recently found several DVD's (made on DVD-R from video tapes, probably) for sale on eBay. The one I bought was an excellent copy, and it was so great to see the show again after more than 25 years. There are some songs on the show that were not on the album, and some of the songs on the album were studio versions of the same songs on

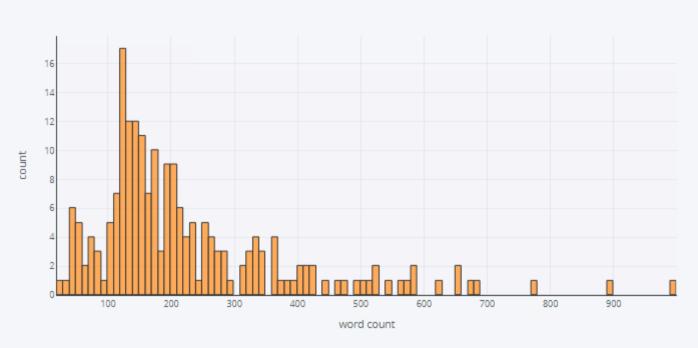


#### Sentiment Polarity Distribution

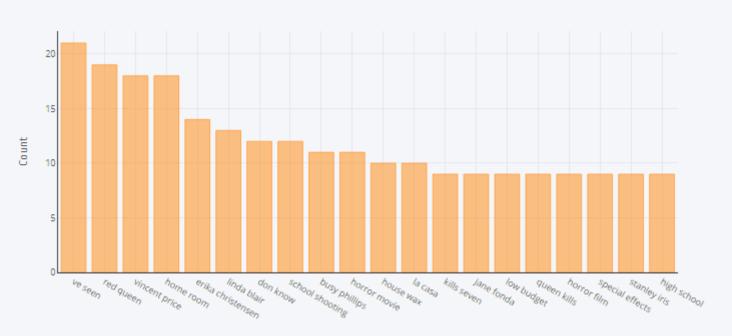




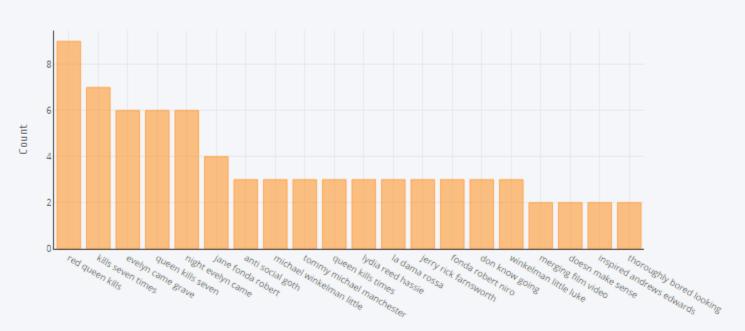
#### Review Text Word Count Distribution



Top 20 bigrams in review after removing stop words



Top 20 trigrams in review after removing stop words



### Thanks!