# AML 2304 – Natural Language Processing Movie Sentiment Analysis (Mid Submission)

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#### **Submitted to:**

Prof. Bhavik Gandhi

## **Submitted by Group 3:**

**Abhishek Natani** 

**Auradee Castro** 

Bhumika Rajendra Babu

Miraj Sinya

Olivia Deguit

Rochan Mehta

Roger Mais

Varun Sharma



# Case Background

- Movie Reviews have always been a reference point for the audience to decide weather or not to watch movies but from a production standpoint it has not be utilized to its full extent.
- This application will allow the producers in the mass media and entertainment industry to not only understand the sentiment of the audience but also the reasons behind those sentiments.
  - Behind the scenes, the application uses Natural Language Processing models along with clustering techniques to make sense of the sentiment and get the intent behind those sentiments giving deeper insights into audience opinions, preferences, and pain points.
  - Help potential clients to potentially get insights on trends in the audience's sentiment towards specific genre over the years, actors, directors which will aid them in making informed decisions.
- Dataset containing movie reviews, details of the movie (genre, budget, cast and crew, ratings) is taken from Kaggle along with the data accessible to the public on IMDb, which is used for cross-validating the data collected from Kaggle.
  - Dataset containing 500K reviews for movies

# Case Study (1/4): Marvel Studios

Thor: Love and Thunder released in 2022 rated significantly lower than its prequal Thor: Ragnarok released in 2017



#### THOR: RAGNAROK

PG-13 , 2h 10m Action,Adventure,Sci-Fi,Fantasy,Comedy

Directed By: Taika Waititi In Theaters: Nov 3, 2017 Streaming: Feb 17, 2018

Marvel Studios



#### THOR: LOVE AND THUNDER

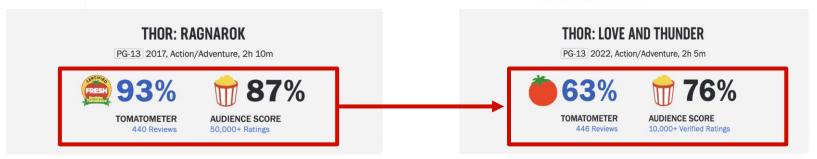
PG-13, 2h 5m

Action, Adventure, Fantasy, Comedy

Directed By: Taika Waititi In Theaters: Jul 8, 2022 Streaming: Sep 8, 2022

Marvel Studios, Walt Disney Pictures,

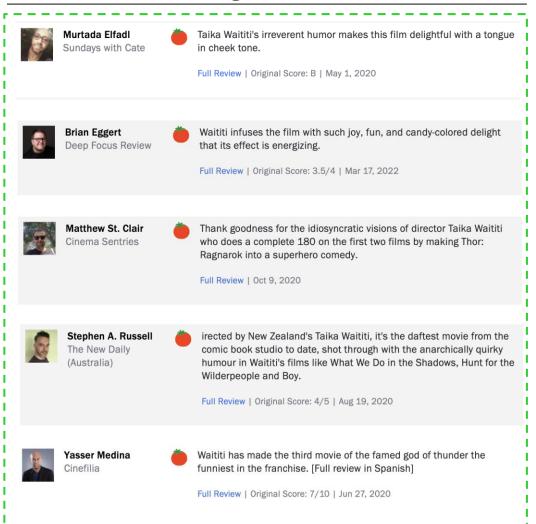
Fox Studios Australia



# Case Study (2/4)

## Factors contributing to the movie being well received by the audience

#### **Thor: Ragnarok Reviews**



Significant mentions of director Taika and directing style infusing comedy into the franchise in the positive reviews of the first movie

# Case Study (3/4)

### Factors contributing to the movie being criticized by the audience

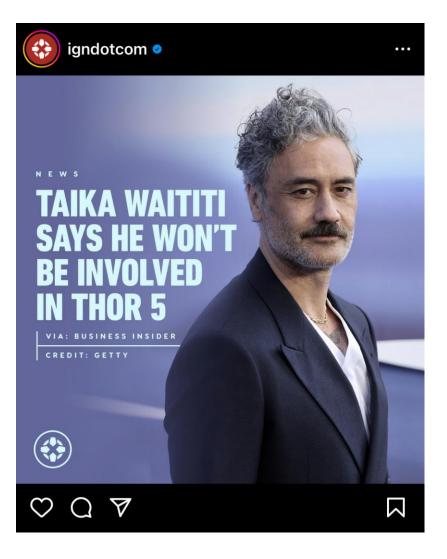
#### **Thor: Love and Thunder Reviews**



Significant mentions of director Taika, movie humor and story in the negative reviews of the first movie

# Case Study (4/4)

Failure of sequel along with negative reviews of movies potentially led to Director Taika Waititi's subsequent departure from the Thor franchise



# Mass Media & Entertainment Domain (1/2)

Producers, a potential customer for the movie review sentiment analyzer

#### **Distributor** Cinema **Producer** Key Resource: Key Resource: **Distribution Rights** Cinema Key Resource: Theatre Content Cost: Distribution Costs Cost: Contract Contract Cost: Distributor's \*OTT **Theatrical** Production share of Ticket Distributor Distributor Costs Fees Distributor sales share **On Demand Content** Cinema Experience Subscription **Audience** -▶: Revenue : Content

**Business Model** 

#### **Details**

- Movie Production Studios establish contractual relation with Distributors to distribute their original content
- Distributors own the distribution rights to movie content which is shown across cinema or OTT platforms
  - Theatrical distributors distribute film content to movie theatre for a portion of revenue generated from ticket sales
  - OTT distributors make film content available to audience on demand on their platform in exchange for subscription fees
- Movie Production Studio:
  - The Walt Disney Studios (Global Box Office Sales: \$79B)
  - Warner Bros. Entertainment Inc. (Global box Office Sales: \$48B)
  - Universal City Studios LLC (Global Box Office Sales: \$47.9B)

Top Players

Description

\* OTT : Over-The-Top (Film content over the internet) Global Box Office Sales figures based on 2022 data

# Mass Media & Entertainment Domain (2/2)

Movie Review Sentiment Analyzer a useful tool for OTT Distributors Original Content creation

#### **Business Model OTT Distributor In-House Producer** Key Resource: Content Cost: Production Costs **Producer OTT Streaming Platform** Key Resource: Content Contract Original Third-Party Cost: Content Content Production Fees Costs Subscription On Demand Content ▶ : Revenue : Content **Audience**

#### **Details**

- Some OTT distributors like Netflix, Amazon prime Video, Disney+ have also entered the content production space
  - Inhouse production unit to develop and stream original content
- Subscription models adopted by popular OTT distributors:
  - Subscription Video On Demand (SVOD): Monthly plan for unlimited access to content anytime
  - Transactional Video On Demand (TVOD): Only pay for content audience want to watch
  - Hybrid Model: Combination of SVOD
     & TVOD
- OTT Distributor:
  - Netflix (Global Subscribers: 231M)
  - Amazon Prime Video (Global Subscribers: 200M)
  - Disney+ (Global Subscribers: 138M)

Top Players

Description

# Stage 1: Data Acquisition

- Sourcing relevant data to address the identified business problem
- Cross-validating dataset
- Data Sources used:

kaggle



# Stage 2: Model Development

#### Sentiment Analysis Model Building

\*Intent Clustering

\*Model Integration

Stage 3: Model Deployment

- Labelled Data Generation for model training
- Steps involved:

Movie Reviews Sentiment label generation using Roberta Model



Data Preprocessing



Model Training & Evaluation

- Naïve Bayes
- \*RNN (LSTM)
- \*SVM

- Intent Clustering Technique Evaluation
- Steps Involved:

Apply Clustering technique



Movie Review Dataset



Intent Clustering
Technique
Evaluation

- Integrating intent clustering & sentiment model on business queries.
- Steps Involved:

Movie Review

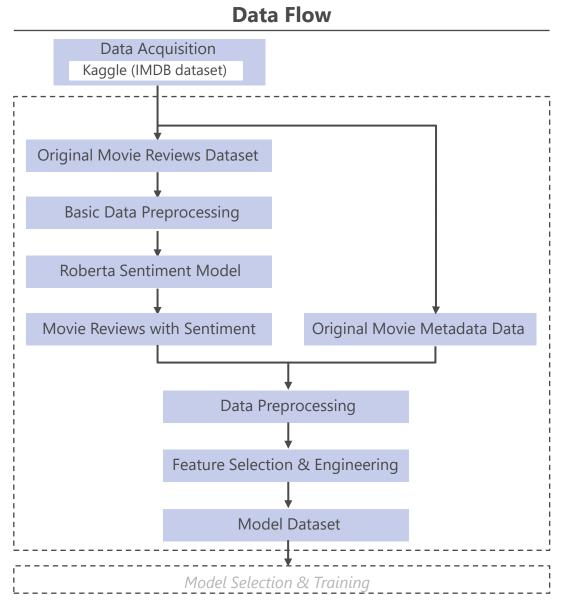


Best Performing Intent Clustering Technique



Best Performing Sentiment Analysis Model Deployment of the model together with User Interface allowing the user to get the sentiment and the intent behind the sentiment that helps them in the movie production process

# Data Pipeline (2/3): Data Preprocessing & Creation



#### **Details**

Description

 RoBERTa Sentiment Model for generating sentiment labels on movie reviews since it has higher accuracy than Vader

Data Pre-Processing

- Null values and duplicate records
- \*\* Non-grammatical text (emails and URLs)
- Non-ascii and diacritics characters
- Emojis, Slangs and \* Abbreviations
- Word contractions
- Name Entity Recognition
- \*Spellchecker and \*POS Tagging
- Lowercasing and \*\* whitespaces
- Non-alphanumeric characters
- \*\*Stopwords and lemmatization (Spacy)

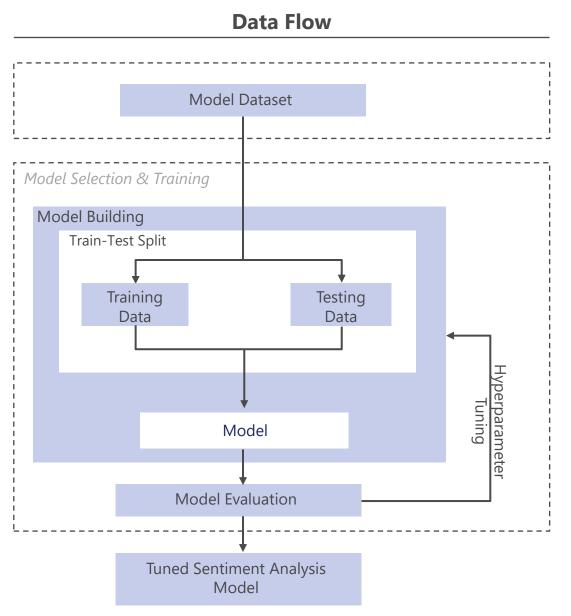
Feature Selection & Engineering

- Count Vectorizer (better than TF-IDF)
- Label encoding:

Sentiment Labels → Numbers

- o 0: Negative
- o 1: Neutral
- 2: Positive

# Data Pipeline (3/3): Model Building and Evaluation



#### **Details**

# Model Creation

- Employed **Multinomial Naïve Bayes** for the initial model as it is robust for overfitting
- Percentage of training data: 80%

# Model Evaluation

- Classification Report
  - o Accuracy: 0.71
  - Precision: 0.64 (Negative), 0.63 (Neutral), 0.78 (Positive)
  - o Recall
  - o F1-score
- \*ROC curve and AUC:
  - 0.88 (Negative, Positive), 0.76 (Neutral)

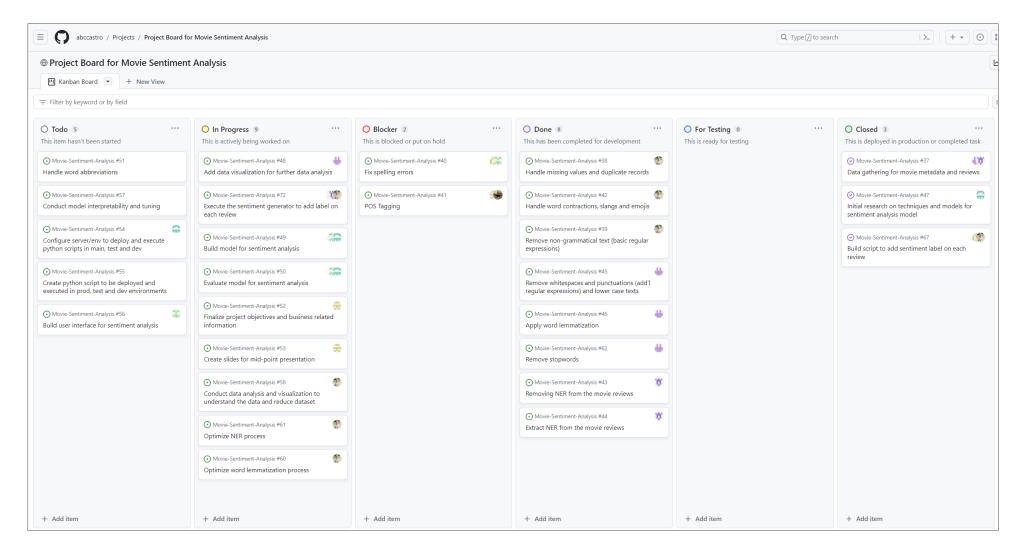
# Hyperparameter Tuning

- GridSearchCV for hyperparameter tuning
- Parameter: Alpha

Note: Minor change when adjusting alpha; higher alpha values improve accuracy, while lower alpha values improve precision. Alpha 1 has an optimal balance across accuracy, precision, recall, and F1-score

# Project Board Walkthrough

# Using Github's Project to create Kanban Board for model development



<sup>\*</sup> Project Board for Movie Sentiment Analysis: https://github.com/users/abccastro/projects/1

#### Team's Best Practices

- **Feature Breakdown:** Dividing features into smaller tasks for better prioritization and more manageable components
- Distributed Responsibilities: Assigning tasks to individual team members to ensure clear responsibility
   and accountability
- Project Progress Checkpoint: Conducting regular team meetings, via MS Teams or in person, for updates,
   blockers and planning next task
- Model Development Workflow: Leveraging \* GitHub as a model development repository, adhering to
  industry standards with distinct branches for production, testing, development, and tasks, and conducting
  code reviews before merging





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