# Reddit Sentiment Analysis

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## **Executive Summary**

Our project aims to analyze and compare different communities. By leveraging machine learning models, NLP models, and Snowflake/AWS we focused on extracting insights from textual data. The primary goal was to use a pre-trained transformer from hugging face to analyze different communities on reddit through sentiment analysis on comments.

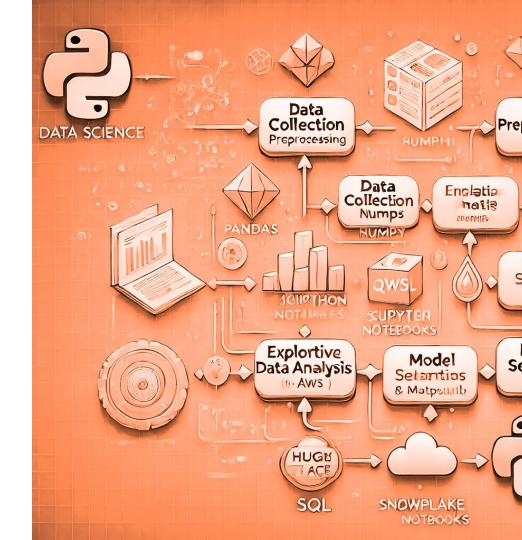


# Approach to Achieve Goals

#### Workflow:

- Data collection and preprocessing.
- 2. Exploratory data analysis for insights.
- 3. Model selection through Hugging Face.
- 4. Testing, evaluation, and comparison using real-world examples.

**Tools Used:** Python (Pandas, NumPy, Scikit-learn, Matplotlib), Jupyter Notebooks, Snowflake (SQL, AWS), Hugging Face (DistilBERT, utrobinmv), Reddit API.





### **Data Collection**

### → Source

Data was gathered from communities using Reddit API

### → Volume

Current dataset includes [1200 entries, 5 features]

### → Rationale

These platforms were selected due to their high volume of user-generated content and relevance to our analysis goals.

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# Data Collection & Data Cleanup

### **Data Cleanup:**

- Handled missing values (deleted/removed comments)
- Removed special characters from comments
- Standardized text formats (e.g., lowercasing, punctuation removal).

#### **Data Collection:**

 Created a function that uses the Reddit API to pull comments from a given number of posts from a given community

**Challenges:** Removed or deleted comments, and removal of special characters

```
def preprocess_comment_data(comments):
    # removing [deleted] , [removed] and special character strings from comments
    # applying regex for this
    comments = comments.replace("[deleted]", "").replace("[removed]", "")
    comments = re.sub(r'[^a-zA-Z0-9\s]', '', comments)
    return comments
```

```
def fetch reddit data(subreddit name, post limit=10):
    """Fetch titles and comments from a subreddit."""
   reddit = praw.Reddit(
     client id=REDDIT API ID.
     client_secret=REDDIT_API_SECRET,
     user agent=REDDIT USER
   subreddit = reddit.subreddit(subreddit name)
   posts_data = []
   for post in subreddit.hot(limit=post limit):
       post info = {
            "title": post.title,
            "comments": []
       post.comments.replace more(limit=0)
       for comment in post.comments.list()[:10]:
           post info["comments"].append(comment.body)
       posts_data.append(post_info)
   for post in posts data:
       post["comments"] = [preprocess_comment_data(comment) for comment in post["comments"]]
   return posts data
```

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## Sentiment Analysis

### **Sentiment Analysis:**

- Created a function that returned a dataframe with the predicted sentiments attached to each comment and their features
- Used distilBERT for sentiment analysis, and utrobin

**Challenges:** Could only take in 512 words, so if a comment had too many words, we would use the utrobinmy summarization model from Hugging Face.

```
# Sentiment analysis model
sentiment_model = pipeline("sentiment-analysis", model="distilbert-base-uncased-finetuned-sst-2-english", to
summarizer = pipeline("summarization", model="utrobinmv/t5_summary_en_ru_zh_base_2048")
analysis results = []
for post in posts data:
    for comment in post["comments"]:
        tokenized comment = sentiment model.tokenizer(comment, truncation=False, return tensors="pt")
        print(len(tokenized comment['input ids'][0]))
        while len(tokenized_comment["input_ids"][0]) > 512:
           print(len(tokenized_comment['input_ids'][0]))
            # Summarize the comment to fit within the token limit
            summary = summarizer(comment, max length=512, min length=100, do sample=False)
            comment = summary[0]["summary_text"] # Replace comment with its summary
            tokenized comment = sentiment model.tokenizer(comment, truncation=False, return tensors="pt")
            # print("After Summary")
            # print(len(tokenized comment['input ids'][0]))
```

```
# Perform sentiment analysis
    comment_sentiment = sentiment_model(comment)
    analysis_results.append({
        "text": comment,
        "type": "Comment",
        "label": comment_sentiment[0]['label'],
        "score": comment_sentiment[0]['score'],
        "subreddit": subreddit_name
    })

return pd.DataFrame(analysis_results)
```

# Final Program & Snowflake

### **Everything Brought Together:**

 Created a main function that asks user to input a subreddit topic and specify the number of top posts to analyze, then the number of top posts to be analyzed from that community

### **Snowflake/Data Storage:**

- Used Snowflake to store data in one area every time the program is ran.
- The cloud we used was AWS.

```
## Creating Table for storing Sentiment Data from df
sentimentss = pd.read csv('sentiments data.csv')
conn = snowflake.connector.connect(
    user=SNOWFLAKE USER,
    password=SNOWFLAKE PASSWORD.
    account=SNOWFLAKE ACCOUNT,
    warehouse=SNOWFLAKE WAREHOUSE.
    database=SNOWFLAKE DATABASE.
    schema=SNOWFLAKE SCHEMA
cursor = conn.cursor()
create_table_query = """
CREATE OR REPLACE TABLE comments analysis (
     "type" STRING.
    "score" FLOAT
cursor.execute(create table query)
cursor.close()
```

conn.close()

```
text type label score subredding to the subreddi
```

```
def main():
    subreddit_name = input("Enter the subreddit topic: ").strip()
    post_limit = int(input("Enter the number of posts to analyze: ").strip())
    print(f"Fetching data from r/(subreddit_name)...")
    post_data = fetch_reddit_data(subreddit_name, post_limit)

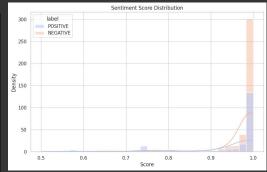
# print(posts_data)

post_data_df = pd.DataFrame(posts_data)
    post_data_df.to_csv("posts_data.csv", index=False)

print("Performing sentiment analysis...")
    sentiments_df = perform_sentiment_analysis(posts_data , subreddit_name)
    load_dataFrame_to_snowflake(sentiments_df)

sentiment_results = fetch_data()

# converting tuples in to dataFrame
    sentiments_results_df = pd.DataFrame(sentiment_results, columns=['text', 'ty
    print("Visualizing sentiments...")
    visualize_sentiments(sentiments_results_df)
    print("Analysis complete!")
```

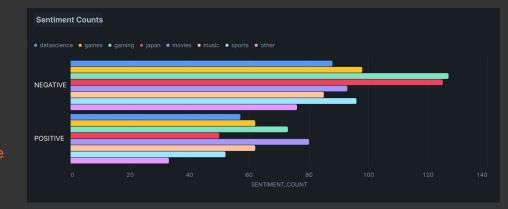


## Findings

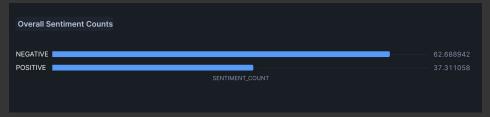
- There always seemed to be proportionally more negative comments than positive in each community.
  - a. People may be more inclined to interact or comment if they share opposite opinions.

### **Challenges:**

- Wanted to attach dates to comments and analyze the spike in comments based on events but would have had to go back and change lists to dictionaries, which we didn't have enough time to do.
- Model didn't take context into account









### **Future Development**

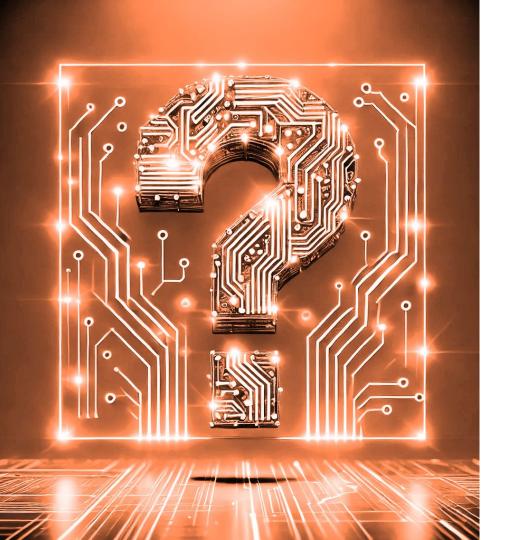
#### **Unanswered Questions:**

- How can we improve sentiment analysis accuracy further? (95.8%, but most likely lower)
- What additional data sources could enhance results?

### **Next Steps:**

- Extend analysis to more features such as dates.
- Extend analysis to other platforms or languages.
- Use techniques to take in and incorporate context into sentiment analysis.
- Learn more of what Snowflake and SQL have to offer





### **Questions & Answers**

We are happy to answer any questions about our project!