Yahoo News Comments Classifier

NLP-Based Classification of Yahoo
Comments for Enhanced Content Moderation

Presented by

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Introduction

- The surge of social media and online forums has transformed communication dynamics,
 with comments serving as crucial elements for user engagement.
- However, moderating user-generated content poses challenges due to the decentralized nature of these platforms.
- To address this, comment categorization using natural language processing offers a scalable solution to classify comments and ensure the safety of online communities.

Problem Statement

- Proliferation of harmful or inappropriate comments due to platform anonymity.
- Negative impact of inconsistent and biased categorization on user experience.
- Manual categorization is limited by time constraints, scalability issues, and biases.
- Need for an automated solution to overcome these challenges.

Proposed Solution & Methodology

- We are introducing a comment categorization tool as a promising approach to automate content moderation.
- We aim to presents a methodology tailored for Yahoo comments, aiming to deploy a
 robust system capable of categorizing comments into six distinct categories (Humour,
 Spam, Neutral, Consolidating, Ideological, Abusive), thereby contributing to the
 advancement of automated content moderation.

Dataset

- Utilization of web scraping techniques for dataset collection.
- Classification of comments into six categories:
 Ideological, Humorous, Consolidating, Abusive,
 Spam, and Neutral.
- Training dataset consisting of 20x6 comments categorized across multiple topics.

```
// Get the child count
const childCount = document.querySelector("#spotim-specific > div > div").
 shadowRoot.querySelector("div > div 
div.ToastWrapper providerContainer--11-4-15" +
      "> div.spcv conversation > div.ToastWrapper providerContainer--11-4-15 >
div:nth-child(2) > ul").children.length;
// Create an array to hold the CSV data
const csvData = []:
 // Loop starts from index 1 (second child) to account for the initial loop check
 for (i = 1; i <= childCount; i++) {
          const cellData = document.querySelector("#spotim-specific > div > div").
           shadowRoot.guerySelector(div > div >
                   div.ToastWrapper providerContainer--11 - 4 - 15 >
                        div.spcv conversation > div.ToastWrapper providerContainer--11 - 4 - 15 >
                       div: nth - child(2) > ul > li: nth - child(${ i }) > article > div > div >
     div: nth - child(1) > div > div > div.components - MessageLayout - index_message
      - view > div >div.components - MessageContent - index messageEntitiesWrapper >
       div > span > div > p).textContent;
          console.log(cellData);
          csvData.push(cellData);
```

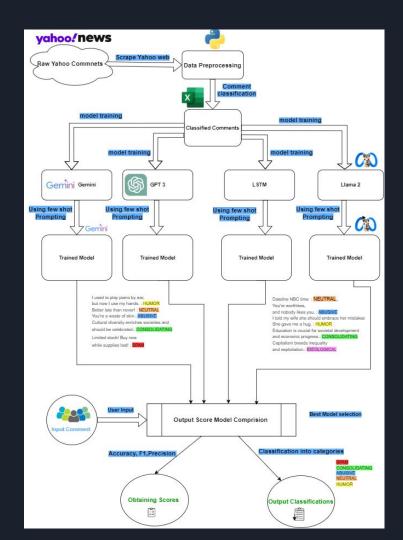
Dataset Cntd.

Headline	Comments
Chris Pratt and Katherine Schwarzenegger slamme	They couldn't just demolish a house without appro
	If it was that big of a deal, they why didn't the city
	Sorry, from the couple pictures I see here the hous
	A 1950's built home would probably suffer from fo
	It is their property. They paid \$12 million for it. It
	If the house was so rare and amazing, why was it s
	It is impossible these days to do something withou
	It's THEIR HOUSE they can do whatever they war
	I could have sworn the term "McMansion" specific
	If it wasn't preserved they have a right to do as the
	The thousandaires on the "internet" love to be ap
	They purchased the home and property and were
	I don't always, or even often, approve what people
	Hey! It's NOT even a Craig Ellwood designed prope
	Nobody cared until there was someone they could
	I don't blame them.
	The house wasn't even designed by Craig Ellwood,
	It's their house. WHO CARES what they did with it
	Properties in the LA area get torn down all the tim
	The Zimmerman House was located in Brentwood
	Like it or not, they paid for that property. It's their
	I would have torn it down, too. They paid for it, th
	-

Abusive Consolida Ideologica Neutral Spam Humor Get rich qui told my vYou're wo Climate chSocialism i Milkmen are a relic of the past fondly r Lose 20 pc Why don'll hope you Education Capitalism I'm happy his family finally got resoluti Unlock excl'm readin You're so Access to Feminism ... but we know there's another person Congratul I used to p Nobody calncome in Conservat He survives WW2 only to be murdered Increase y Why did tl You're a w Racial disc Libertariar I understand this case is decades old.b Make mor I'm on a w Go jump c Freedom (Anarchism Neutral Looking follm trying I wish you Technolog Environme Just make sure when you say "no crim Meet hot (Why don't You're ugl Democrac Nationalis Was Williams investigated between the Boost you I told my cl hope you Globalizat Communi: I have fond memories of when I was a Limited st(I'm readin You're not Mental he Islamophe dumpo will just make fun of the Veter Get a free Why did tl I hope you Cultural di Secularism Better late than never! Earn \$100 I used to LYou're sucThe role o Liberalism Price of a first class stamp in 1968 was Invest in c Parallel lin You don't Art and cu Populism ...and? It wasn't solved now. Snitching Need a loal'm writing wish you Corporate Multicultu Dateline NBC time. Upgrade t Why don't You're a w Technolog Globalism No crime goes unpunished? Didn't the Looking fo I told my vI hate evei Investmer Humanish Conveniently leave out any photos of c

Model Description

- Comparison of state-of-the-art models: GPT-3.5,
 Google Gemini, LLAMA2 and LSTM.
- Consideration of model performance metrics including precision, recall, F1 score, and accuracy.
- Evaluation of Model considering the best accuracy from all 4 techniques.



GPT-3.5

- Bidirectional Context Awareness
- Versatility
- Large-Scale Language Understanding

Llama 2

- Transformer Architecture
- Semantic Coherence
- Long-Range Dependencies

```
output = ""
        for event in replicate.stream("meta/llama-2-70b-chat", input=
        {"system prompt": "Predict which category does the given comment falls into
        Humor or Consolidating or Abusive, Just output the category",
        "prompt": "Why don't scientists trust stairs? Because they're always up to something."}):
            output += str(event)
        print(output)
output = replicate.run(
  "stability-ai/sdxl:39ed52f2a78e934b3ba6e2a89f5b1c712de7dfea535525255b1aa35c5565e08b",
 input={
        "system prompt": "Learn from the given comments. Just learn from them",
        "prompt": """Humor: I told my wife she should embrace her mistakes... She gave me a hug.
                    Humor: Why don't skeletons fight each other? They don't have the guts.
                   Humor: I'm reading a book on anti-gravity. It's impossible to put down!
       Abusive: You're worthless, and nobody likes you.
       Abusive: I hope you fail at everything you do.
       Abusive: You're so stupid, it's embarrassing.
       Consolidating: Climate change is a pressing issue that requires global cooperation to address.
       Consolidating: Education is crucial for societal development and economic progress.
       Consolidating: Access to healthcare should be a fundamental right for all individuals."""
print(output)
```

Google Gemini

- Graph Neural Networks
- Comprehensive Understanding
- Semantic Association Extraction

```
def to_markdown(text):
    text = text.replace('•', ' *')
    return Markdown(textwrap.indent(text, '> ', predicate=lambda _: True))

genai.configure(api_key="Enter your API Key")
model = genai.GenerativeModel('gemini-pro')
```

LSTM

- Sequential Data Processing
- Context Understanding
- Long-Term Dependencies

```
def build_and_train_lstm_model(X_train, y_train, vocab_size, max_seq_length, num_classes, embedding_dim=100, lstm_units=128):
    model = Sequential()
    model.add(Embedding(input_dim=vocab_size, output_dim=embedding_dim, input_length=max_seq_length))
    model.add(LSTM(units=lstm_units))
    model.add(Dense(units=num_classes, activation='softmax'))

model.compile(optimizer='adam', loss='sparse_categorical_crossentropy', metrics=['accuracy'])
    model.fit(X_train, y_train, epochs=10, batch_size=64) # Removed validation_split parameter
    return model
```

```
sequences = tokenizer.texts_to_sequences(comments)
max_seq_length = max([len(seq) for seq in sequences])
padded_sequences = pad_sequences(sequences, maxlen=max_seq_length, padding='post')

# Step 3: Split the data into training and test sets
X_train, X_test, y_train, y_test = train_test_split(padded_sequences, labels, test_size=0.2, random_state=42)

# Convert X_train and y_train to numpy arrays
X_train = np.array(X_train)
y_train = np.array(y_train)
X_test = np.array(y_train)
X_test = np.array(y_test) # Convert X_test to numpy array
y_test = np.array(y_test) # Convert y_test to numpy array

# Step 4: Build and train the LSTM model
num_classes = len(categories)
print("Building and training the LSTM model...")
model = build_and_train_lstm_model(X_train, y_train, vocab_size, max_seq_length, num_classes)
```

Evaluation Metrics & Results

GPT 3.5:

Gemini:

Accuracy: 0.33333333333333333

Precision: 0.25

Recall: 0.333333333333333333

F1 Score: 0.277777777777773

Llama 2:

Accuracy: 0.6916666666666667

Precision: 0.6920803782505911

Recall: 0.6916666666666667

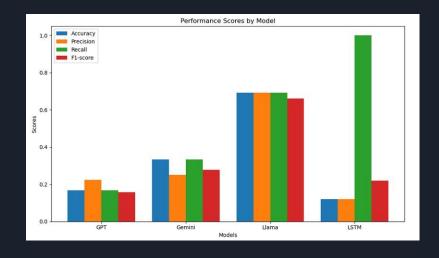
F1-score: 0.6618145318066764

LSTM:

	precision	recall	f1-score	support
Ideological	0.00	0.00	0.00	5
Humor	0.00	0.00	0.00	3
Consolidating	0.00	0.00	0.00	5
Abusive	0.00	0.00	0.00	5
Spam	0.12	1.00	0.22	3
Neutral	0.00	0.00	0.00	3
accuracy			0.12	24

Findings & Conclusions

- Llama excels with high accuracy, precision, and
 F1-score.
- LSTM achieves perfect recall but suffers from low precision.
- Gemini demonstrates decent performance but falls short in precision.
- Overall, Llama emerges as the best-performing model, balancing precision and recall effectively.



Application & Future Work

- Potential applications of the developed comment classification model in automated content moderation.
- Exploration of future research directions, such as enhancing model interpretability and generalizability.
- Consideration of scalability and adaptability of the model to other online platforms beyond Yahoo News.
- Importance of ongoing research and development to address evolving challenges in online content moderation.
- we envision building a Chrome extension based on our implementation, allowing users to seamlessly integrate our comment categorization tool into their browsing experience

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Thank You