

HUMOR DETECTION USING NLP TECHNIQUES

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BACKGROUND

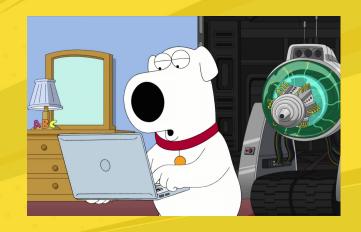
- Family Guy is an American adult animated sitcom created by Seth MacFarlane, which first aired in 1999
- Family Guy has become one of the most recognizable shows in the world due to its satirical humor which makes bold references to today's popular culture
- Recognizable by its distinctive style of cutaway jokes, flashbacks, and random comedy
- 424 episode across 22 seasons have been released





OUR DATA

 Downloaded a dataset from Kaggle which includes 19 seasons of Family Guy, and contains all of the dialogue from each episode along with timestamps



PROJECT MOTIVATION

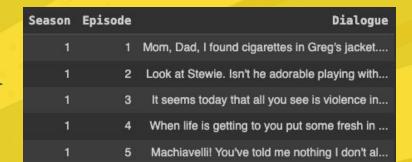
- Leverage NLP techniques learned in class to analyze humor within Family Guy
 - Since humor is very complex and subjective, our idea behind this is to gain a better understanding of what is generally considered "funny"
- Given our dataset has all of the episode dialogues, this gives us the opportunity to use the techniques learned in class and apply them to a real-world application



DATA CLEANING

Episodes average 20 - 27 minutes

Season	Episode	Time_Stamp	Dialogue
1	1	00:00:07	Mom, Dad, I found cigarettes in Greg's jacket.
1	1	00:00:10	Greg, were you smoking cigarettes?
1	1	00:00:10	No, Dad.
1	1	00:00:12	He's lying. There's no doubt about that.
1	1	00:00:14	Greg, I'm afraid your punishment will be four





OUR GOALS



- 1. Can we determine if an episode is funny?
- 2. Can we analyze jokes by length?
 - a. Can we determine if a sentence is considered a "joke" based just on length?
- 3. Can we determine if a joke is inappropriate?



objective

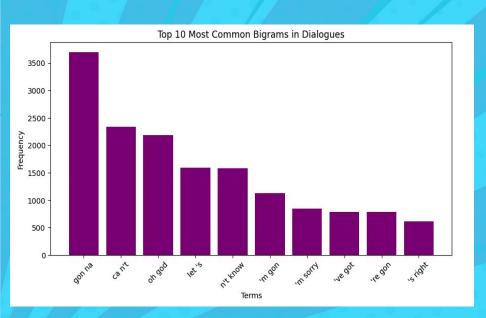
Analyze and rank episodes by humor using sentiment analysis

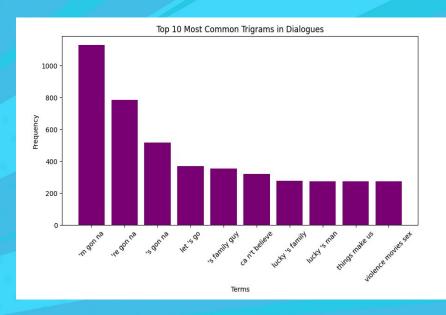
Methods

Keyword-based analysis and VADER sentiment analysis



INITIAL VISUALIZATIONS





Keyword Sentiment Scoring

Count based on predefined positive and negative keywords

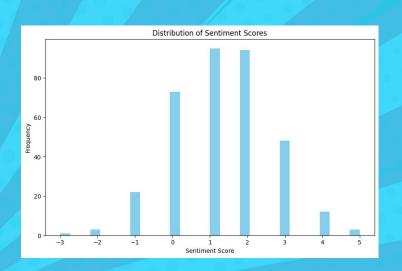
VADER SENTIMENT ANALYSIS

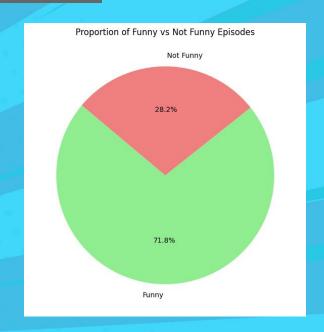
Utilizes pre-trained model for nuanced sentiment scoring.

		Season	Episode	Sentiment_Score	Is_Funny
	0	1	1	2	Funny
	1	1	2	0	Not Funny
	2	1	3	0	Not Funny
	3	1.	4	5	Funny
	4	1	5	2	Funny
	95	6	1	1	Funny
	96	6	2	2	Funny
	97	6	3	2	Funny
	98	6	4	0	Not Funny
	99	6	5	0	Not Funny

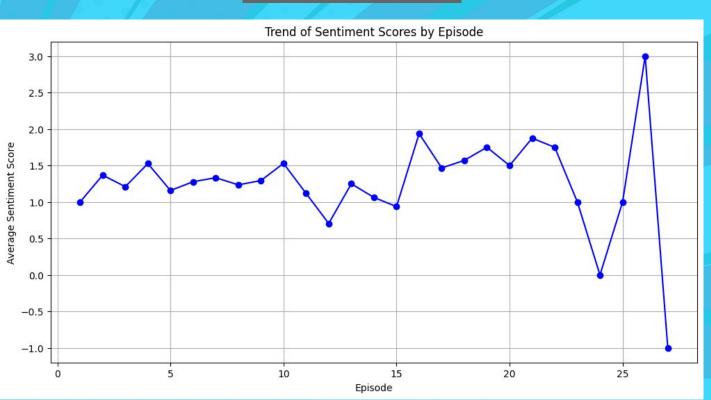
```
import pandas as pd
from nltk.sentiment import SentimentIntensityAnalyzer
# NLTK's VADER SentimentIntensityAnalyzer
sia = SentimentIntensityAnalyzer()
# get sentiment score using VADER
def evaluate sentiment vader(dialogue):
   score = sia.polarity_scores(dialogue)
   return score['compound'] # Return the compound score
fm combined dialogue only['VADER Score'] = fm combined dialogue only['Dialogue'].apply(evaluate sentiment vader)
# group by season and episode
# calculate the mean sentiment score
episode_sentiment = fm_combined_dialogue_only.groupby(['Season', 'Episode']).agg({
     'VADER Score': 'mean'
}).reset_index()
episode_sentiment.rename(columns={'VADER_Score': 'Average_Sentiment_Score'}, inplace=True)
```

HOW MANY FUNNY EPISODES ARE THERE?





season 1



Results

	Season	Episode	Average_Sentiment_Score	Humor_Rank
132	8	10	1.0	1.0
140	8	18	1.0	1.0
46	3	19	1.0	1.0
40	3	13	1.0	1.0
250	14	5	1.0	1.0
35	3	8	1.0	1.0
120	7	14	1.0	1.0
32	3	5	1.0	1.0
103	6	9	1.0	1.0
29	3	2	1.0	1.0

2. CAN A JOKE BE DETERMINED BY ITS LENGTH?



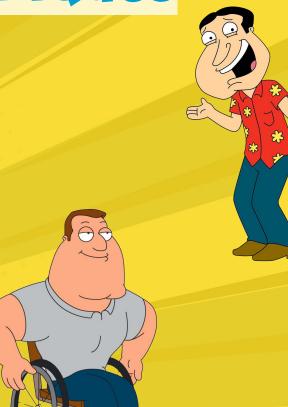
ANALYTICAL TECHNIQUES

Feature extraction

- Sentence length and Word Count: essential metrics to gauge dialogue brevity and content density
- Vocabulary complexity: Evaluated to understand linguistic complexity

Sentiment Analysis

- Tools Used: AFINN, BING, and TextBlob for sentiment and polarity scores
- Findings: Explored correlations between positive sentiments and increased humour



LINGUISTIC PATTERNS

Linguistic Analysis

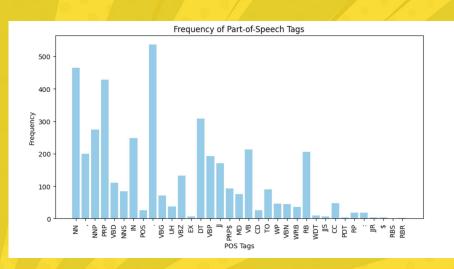
- POS Tagging: Identified common grammatical structures in humorous dialogues.
- NER: Highlighted frequent named entities involved in jokes
 - Indicating character or situational humour

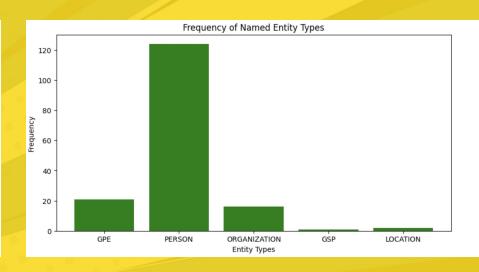
```
[44] nltk.download('maxent_ne_chunker')
    nltk.download('words')

def get_named_entities(text):
    tokens = nltk.word_tokenize(text)
    tags = nltk.pos_tag(tokens)
    entities = nltk.chunk.ne_chunk(tags)
    return [entity for entity in entities if isinstance(entity, nltk.tree.Tree)]

fm['Named_Entities'] = fm['Dialogue'].apply(get_named_entities)
```

VISUALIZING POS AND NER





CLUSTERING

Метнор

Used KMeans to group dialogues based on linguistic features

```
Examining dialogues in cluster 4
83
                       That's just crazy enough to work.
116
       Peter, I like you. But I need you to be more t...
135
       Just not as much. So it might get a little com...
146
       When she worries, she says, I told you so and:...
171
       Screw this! I just came over to buy some firew...
180
                               Lois, we just had dinner.
183
                                               I just...
188
                          It's just... It's not healthy.
212
                      I hate lying to Lois. It's just...
244
       No. I just rented it. But they're gonna be tic...
283
                                 Just like the Kennedys.
285
       The man I married would never think he could f...
297
       Yeah, I know. I'm just making conversation. Co...
337
                            You just want your toy back.
358
       Okay, everybody, I feel really bad about what ...
424
       Honey, it's not gonna go away just because you...
441
                      Jeez. Did I just hit that ostrich?
488
       You just have to remember that life has its li...
493
       You just knocked out cable TV for the whole town!
506
       I can't believe you just sold out your own dau...
```

outcome

Successfully identified clusters that significantly contained humor based on dialogue length and linguistic cues



```
Episode Time_Stamp \
        Season
35
             1
                           00:02:26
63
                          00:03:48
65
                           00:03:49
94
                           00:05:08
100
                           00:05:30
...
           ...
                                ...
154938
            19
                           00:05:46
155015
            19
                          00:09:34
155132
            19
                           00:15:58
155223
            19
                          00:19:46
155229
            19
                           00:20:08
                                                  Dialogue \
35
                                     Yeah, it's all right.
63
                                               Right here.
65
                                 All right. What do I win?
94
                I almost walked right into that one. God!
100
                                   Mom, are you all right?
...
154938
        All right, we'll find something else. Thank yo...
155015
       I mean, this is the Mafia, not the Laugh Facto...
155132
                                    All right, I'll do it.
155223
                    All right, nobody do anything stupid.
155229
                              All right. So long, you two.
                               NRC_Sentiment AFINN_Sentiment \
35
                                      Neutral
                                                     Positive
63
                                      Neutral
                                                      Neutral
65
                                      Neutral
                                                     Positive
94
        anticipation, fear, joy, positive, trust
                                                     Positive
100
                                      Neutral
                                                      Neutral
...
                                          ...
                                                           ...
154938
                                      Neutral
                                                     Positive
155015
         fear, joy, negative, positive, surprise
                                                     Positive
155132
         anger.disgust.fear.negative.sadness
                                                      Neutral
155223
                                     negative
                                                     Negative
155229
                                                      Neutral
                                 anticipation
```

Results

```
AFINN Sentiment Score BING Sentiment Word Count Polarity \
35
                                    Positive
                                                       4 0.285714
63
                                    Positive
                                                       2 0.285714
65
                                    Positive
                                                       6 0.542857
                                    Positive
                                                       8 0.357143
100
                                     Positive
                                                          0.285714
...
154938
                                    Positive
                                                       10 0.272857
155015
                                    Positive
                                                       11 0.091071
155132
                                    Positive
                                                       5 0.285714
155223
                           -2
                                    Positive
                                                       6 -0.257143
155229
                                    Positive
                                                       6 0.117857
        Subjectivity
35
            0.535714 [(Yeah, UH), (,, ,), (it, PRP), ('s, VBZ), (al...
63
                                      [(Right, RB), (here, RB), (., .)]
            0.535714
            0.467857 [(All, DT), (right, NN), (., .), (What, WP), (...
            0.535714 [(I, PRP), (almost, RB), (walked, VBD), (right...
94
100
            0.535714 [(Mom, NN), (,, ,), (are, VBP), (you, PRP), (a...
...
                     [(All, DT), (right, JJ), (,, ,), (we, PRP), ('...
154938
            0.397857
155015
            0.441071 [(I, PRP), (mean, VBP), (,, ,), (this, DT), (i...
            0.535714 [(All, DT), (right, NN), (,, ,), (I, PRP), ('l...
155132
155223
            0.767857 [(All, DT), (right, NN), (,, ,), (nobody, NN),...
155229
            0.467857 [(All, DT), (right, NN), (., .), (So, RB), (lo...
                          Named Entities Cluster
35
63
                                       []
65
                                      []
94
100
                           [[(Mom, NN)]]
...
154938
155015
        [[(Laugh, NNP), (Factory, NNP)]]
155132
155223
                                      []
                                                3
                                       []
155229
```

IS A JOKE APPROPRIATE?



FAMILY GUY RATING

- The Movie Picture Association (MPAA) rating that has been assigned to Family Guy is: "language, some sexual content and drug use"
- Early seasons are rated 12+
- Season 10 and beyond are rated 14+



IS A JOKE APPROPRIATE?

- Profanity Analysis of classified jokes
 - Does a joke contain profanity?
 - How many of the classified jokes use profanity?
- Check Sentiment Scores
- Determine if a joke is inappropriate or appropriate



PROFANITY ANALYSIS

- Package: better_profanity (0.7.0)
- Steps:
 - Extract profane words
 - Identify words
 - Count the frequency



```
from better_profanity import profanity

profanity.load_censor_words()

def extract_profanity_words(dialogues):
    profane_words_list = []
    for dialogue in dialogues:
        words = word_tokenize(dialogue.lower())

    profane_words = [word for word in words if profanity.contains_profanity(word)]
    profane_words_list.extend(profane_words)
    return profane_words_list
```

NOTABLE MENTIONS

Episode 1,1

- Stupid: 35

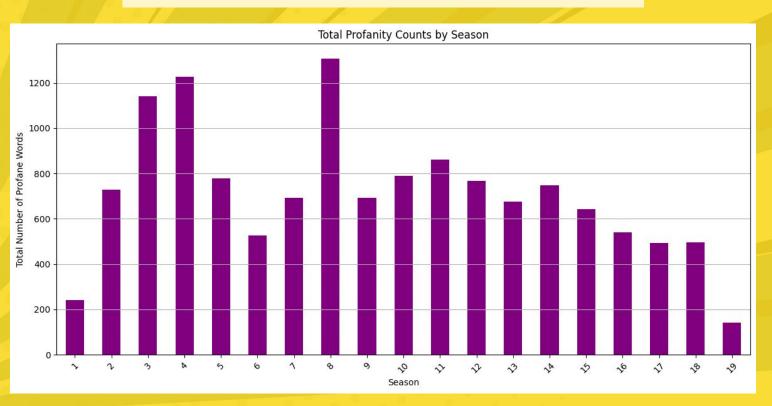
- Fat: 29

- Poop: 13

- F**king: 2



SEASONAL PROFADITY



LOOKING BEYOND PROFANITY

- Using vader_lexicon and SentimentIntensityAnalyzer we analyzed the scores of each joke to determine if it was appropriate
- Positive, Neutral, and Negative

```
Dialogue
     Season Episode
83
        1.0
                                       That's just crazy enough to work.
                 1.0
116
                      Peter, I like you. But I need you to be more t...
135
       1.0
                      Just not as much. So it might get a little com...
146
                      When she worries, she says, I told you so and:...
171
                      Screw this! I just came over to buy some firew...
    Appropriateness
        Appropriate
83
      Inappropriate
135
        Appropriate
146
        Appropriate
      Inappropriate
```

Number of Appropriate dialogues: 12142 Number of Inappropriate dialogues: 989

FUTURE WORK

- Video and Audio Analysis

 Analyze videos and determine how
 does the inputted audio affect a joke
- Time Analysis

 Can we predict if a joke is more likely to be funny depending on when it appears in an episode?
- Edinburgh Fringe Joke detector
- Different Languages



VIDEO ANALYSIS

with quote as source:
 r.adjust_for_ambient_noise(source)
 audio=r.record(source)

result = r.recognize_google(audio)

{'transcript': "do whatever you have to do Brian butt never let him know he's got talent yeah you're right I won't be so hard after all I've certainly done worse I replaced Peter's I Can't Believe It's Not Butter with real butterfly", 'confidence': 0.37292942}.

'Do whatever you have to do, Brian, but never let him know he's got talent.Yeah, you're right.It won't be so hard.After all, I've certainly done worse. I replaced Peter's I Can't Believe It's Not Butter with real butter.Buh... wuh... suh... muh—wuh...I... I can't... I can't believe...Muh... wuh...I do n't know, Doctor.Looking back, I think it may have been real butter.Your husband murdered three children.'

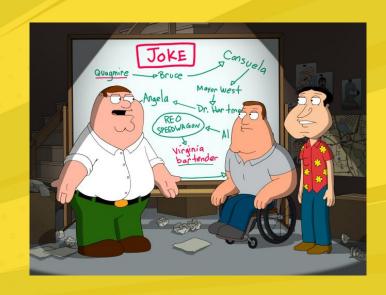
Accuracy: 0.16049382716049382

'Not Funny'



AUDIO ANALYSIS

Audio Emotion Recognition
 We would like to analyze how
 does the background music,
 sound effects, and character's
 tone affect a joke delivery



TIME ANALYSIS

- Given all of the information dialogue's have timestamps, can we determine when a joke is funniest?

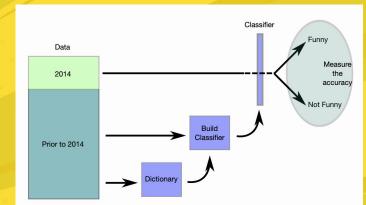


Season	Episode	Time_Stamp	Dialogue
1.0	1.0	00:04:33	That's just crazy enough to work.
1.0	1.0	00:06:14	Peter, I like you. But I need you to be more t
1.0	1.0	00:07:20	Just not as much. So it might get a little com

EDINBURGH FRINGE

 Use an open-source joke detection model inspired by the famous Edinburgh Fringe Festival

- Data includes the top ten jokes of every year since 2011





THANK YOU

