```
In [146]:
           #Installing all the libraries
              import praw
              import pandas as pd
              import nltk
              from nltk.corpus import stopwords
              from nltk.sentiment import SentimentIntensityAnalyzer
              from empath import Empath
              import re
              nltk.download('vader_lexicon')
              nltk.download('stopwords')
              nltk.download('punkt')
              [nltk_data] Downloading package vader_lexicon to
                              C:\Users\sindh\AppData\Roaming\nltk_data...
              [nltk_data]
              [nltk data]
                            Package vader lexicon is already up-to-date!
              [nltk data] Downloading package stopwords to
                              C:\Users\sindh\AppData\Roaming\nltk data...
              [nltk data]
              [nltk_data]
                            Package stopwords is already up-to-date!
              [nltk_data] Downloading package punkt to
              [nltk_data]
                             C:\Users\sindh\AppData\Roaming\nltk_data...
              [nltk_data]
                            Package punkt is already up-to-date!
   Out[146]: True
In [147]: | reddit = praw.Reddit(client_id="HSD2DY-UkJP-XmaT97hwiA",
                                   client secret="e9px65SFRjJTXqNLQbILk lpOa-l Q",
                                   user agent="linux:my reddit bot:v1.0 (by /u/sindhu2111)")
              keyword = "Cincinnati Zoo"
              submissions = reddit.subreddit("all").search(query=keyword, sort="comments", time_filter="all")
              # Create a list of submissions and their comment count
              submissions with comments = [(submission, submission.num comments) for submission in submissions]
              if submissions_with_comments:
                  # Sort the list by the number of comments
                  submissions_with_comments.sort(key=lambda x: x[1], reverse=True)
                  # Print the top 2 submissions and their subreddits sorted by most comments
                  for submission, comments count in submissions with comments[:2]:
                      print(submission.subreddit.display_name, "-", comments_count, "comments")
                      print(submission.permalink)
                  for submission, comments_count in submissions_with_comments[:1]:
                      print("top submission: ",submission.permalink)
                  url1=submission.permalink
                  print("No submissions found for the given keyword.")
              news - 5973 comments
              /r/news/comments/4licqq/cincinnati zoo kills gorilla to protect small/
              nottheonion - 4795 comments
              /r/nottheonion/comments/4z07o6/harambe_stop_making_memes_of_our_dead_gorilla/
              top submission: /r/news/comments/4licqq/cincinnati zoo kills gorilla to protect small/
```

```
In [148]: | url = "https://www.reddit.com"+url1
              submission = reddit.submission(url=url)
              submission.comments.replace more(limit=None)
              comments = [comment for comment in submission.comments.list()]
              # Save raw comments data
              raw_data = {
                  "Author": [comment.author.name if comment.author else "deleted" for comment in comments],
                  "Date": [pd.to_datetime(comment.created_utc, unit="s") for comment in comments],
                  "Comment": [comment.body for comment in comments],
                  "Subreddit": [comment.subreddit.display_name for comment in comments],
              raw_df = pd.DataFrame(raw data)
              raw_df.to_csv("Sentimental_analysis_rawdata.csv", index=False)
In [149]: ▶ import pandas as pd
              import re
              from nltk.corpus import stopwords
              unique_authors = set()
              stop_words = set(stopwords.words("english"))
              def is_comment_from_2016(comment):
                  created year = pd.to datetime(comment.created utc, unit="s").year
                  return created_year == 2016
              def contains_shareable_link(comment):
                  regex = r"http[s]?://(?:[a-zA-Z]|[0-9]|[$-_@.&+]|[!*\\(\\),]|(?:%[0-9a-fA-F][0-9a-fA-F]))+"
                  return bool(re.search(regex, comment))
              def preprocess comment(comment):
                  # Remove URLs
                  comment = re.sub(r'http\S+', '', comment)
                  # Remove punctuation and convert to lowercase
                  comment = re.sub(r'[^\w\s]', '', comment).lower()
                  # Remove stopwords
                  comment = ' '.join([word for word in comment.split() if word not in stop_words])
                  return comment
              filtered_comments = []
              for comment in comments:
                  if (comment.author not in unique_authors and
                      is comment from 2016(comment) and
                      not comment.body == "[deleted]" and
                      not contains_shareable_link(comment.body)):
                      unique_authors.add(comment.author)
                      filtered_comments.append(comment)
              preprocessed comments = [preprocess comment(comment.body) for comment in filtered comments]
              # Save processed comments data
              processed data = {
                  "Author": [comment.author.name if comment.author else "deleted" for comment in filtered_comments],
                  "Date": [pd.to_datetime(comment.created_utc, unit="s") for comment in filtered_comments],
                  "Comment": preprocessed_comments,
                  "Subreddit": [comment.subreddit.display_name for comment in filtered_comments],
              processed_df = pd.DataFrame(processed_data)
```

processed\_df.to\_csv("Sentimental\_analysis\_processeddata.csv", index=False)

```
₩ #Perform VADER and EMPATH sentiment analysis
In [150]:
              sia = SentimentIntensityAnalyzer()
              vader_results = [sia.polarity_scores(comment) for comment in preprocessed_comments]
              lexicon = Empath()
              empath results = []
              selected_categories = ["love", "hate","joy","play","positive_emotion","trust","aggression","anger","fear","pain"
              for comment in preprocessed_comments:
                       categories = {k: v for k, v in lexicon.analyze(comment, normalize=True).items() if k in selected_categor
                       empath_results.append(categories)
              def classify_sentiment(compound_score, very_negative_threshold=-0.5, negative_threshold=-0.05, neutral_threshold
                  if compound_score <= very_negative_threshold:</pre>
                      return "Very Negative'
                  elif very_negative_threshold < compound_score <= negative_threshold:</pre>
                      return "Negative"
                  elif negative_threshold < compound_score <= neutral_threshold:</pre>
                      return "Neutral"
                  elif neutral threshold < compound score < positive threshold:</pre>
                       return "Positive"
                  elif positive_threshold <= compound_score < very_positive_threshold:</pre>
                      return "Very Positive"
                  else:
                       return "Very Positive"
              def classify_empath_sentiment(row, selected_categories):
                  max_category = max(selected_categories, key=lambda x: row[f"EMPATH_{x}"])
                  max_value = row[f"EMPATH_{max_category}"]
                  if max_value == 0:
                      return "Neutral"
                  return max_category.capitalize()
              # Prepare data for DataFrame
              data = {
                   "Author": [comment.author.name if comment.author else "<mark>deleted</mark>" for comment in filtered_comments],
                   "Date": [pd.to_datetime(comment.created_utc, unit="s") for comment in filtered_comments],
                  "Comment": preprocessed comments,
                  "Subreddit": [comment.subreddit.display_name for comment in filtered_comments],
                  "VADER_neg": [result['neg'] for result in vader_results],
                  "VADER_neu": [result['neu'] for result in vader_results],
                  "VADER pos": [result['pos'] for result in vader results],
                  "VADER_compound": [result['compound'] for result in vader_results],
              }
              # Add EMPATH categories to the data dictionary
              for category in selected_categories:
                  data[f"EMPATH_{category}"] = [result[category] for result in empath_results]
              # Create DataFrame
              df = pd.DataFrame(data)
              df["Overall_VADER_sentiment"] = df["VADER_compound"].apply(classify_sentiment)
              df["Overall_EMPATH_sentiment"] = df.apply(lambda row: classify_empath_sentiment(row, selected_categories), axis=
              # Export the DataFrame to a CSV file
              df.to_csv("Sentimental_analysis_VADER_EMPATH.csv", index=False)
```

```
In [151]: ▶
              # Count the occurrences of each sentiment
              sentiment_counts = df["Overall_VADER_sentiment"].value_counts()
              print(sentiment_counts)
              # Calculate the total number of comments
              total\_comments = len(df)
              # Calculate the average sentiment scores
              sentiment_averages = {}
              for sentiment in sentiment counts.index:
                  sentiment_averages[sentiment] = sentiment_counts[sentiment] / total_comments
              print("Average Sentiment Scores:")
              for sentiment, score in sentiment_averages.items():
                  print(f"{sentiment}: {score:.3f}")
              print()
              # Find the most prevalent sentiment
              most_prevalent_sentiment = max(sentiment_averages.items(), key=lambda x: x[1])
              print(f"Most prevalent sentiment: {most_prevalent_sentiment[0]} (Score: {most_prevalent_sentiment[1]:.3f})")
              Very Positive
                               703
              Very Negative
                               651
              Neutral
                               428
              Negative
                               383
              Positive
                                36
              Name: Overall_VADER_sentiment, dtype: int64
              Average Sentiment Scores:
              Very Positive: 0.319
              Very Negative: 0.296
              Neutral: 0.194
              Negative: 0.174
              Positive: 0.016
              Most prevalent sentiment: Very Positive (Score: 0.319)
In [152]: ▶ # Count the occurrences of each EMPATH sentiment
              empath_sentiment_counts = df["Overall_EMPATH_sentiment"].value_counts()
              # Calculate the total number of comments
              total\_comments = len(df)
              # Calculate the average sentiment scores for EMPATH
              empath_sentiment_averages = {}
              for sentiment in empath sentiment counts.index:
                  empath_sentiment_averages[sentiment] = empath_sentiment_counts[sentiment] / total_comments
              print("EMPATH Average Sentiment Scores:")
              for sentiment, score in empath_sentiment_averages.items():
                  print(f"{sentiment}: {score:.3f}")
              most_prevalent_empath_sentiment = max(empath_sentiment_averages.items(), key=lambda x: x[1])
              print(f"Most prevalent EMPATH sentiment: {most_prevalent_empath_sentiment[0]} (Score: {most_prevalent_empath_sen
              EMPATH Average Sentiment Scores:
              Neutral: 0.342
              Play: 0.216
              Negative_emotion: 0.198
              Positive_emotion: 0.066
              Hate: 0.061
              Trust: 0.032
              Love: 0.029
              Pain: 0.026
              Aggression: 0.019
              Fear: 0.006
              Anger: 0.005
              Joy: 0.000
              Most prevalent EMPATH sentiment: Neutral (Score: 0.342)
```

```
In [167]: ▶ from collections import Counter
              import matplotlib.pyplot as plt
              import seaborn as sns
              # Additional insights
              def get top words(df, sentiment, column, n=5):
                  comments = df[df[column] == sentiment]["Comment"]
                  words = [word for comment in comments for word in comment.split()]
                  word_counts = Counter(words)
                  return word_counts.most_common(n)
              df = pd.read_csv("Sentimental_analysis_VADER_EMPATH.csv")
              # Top 5 most common words used in each sentiment category for VADER
              print("Top 5 words for each VADER sentiment category:")
              for sentiment in sentiment averages:
                  top_words = get_top_words(df, sentiment, "Overall_VADER_sentiment")
                  print(f"{sentiment}: {top_words}")
              print()
              # Top 5 most common words used in each sentiment category for EMPATH
              print("Top 5 words for each EMPATH sentiment category:")
              for sentiment in empath_sentiment_averages:
                  top_words = get_top_words(df, sentiment, "Overall_EMPATH_sentiment")
                  print(f"{sentiment}: {top_words}")
              print()
              plt.figure(figsize=(8, 6))
              sns.barplot(x=sentiment_counts.index, y=sentiment_counts.values)
              plt.title("Sentiment Distribution (VADER)", fontsize=16)
              plt.xlabel("Sentiment", fontsize=12)
              plt.ylabel("Count", fontsize=12)
              plt.show()
              plt.figure(figsize=(20, 6))
              sns.barplot(x=empath_sentiment_counts.index, y=empath_sentiment_counts.values)
              plt.title("Sentiment Distribution (EMPATH)", fontsize=16)
              plt.xlabel("Sentiment", fontsize=12)
              plt.ylabel("Count", fontsize=12)
              plt.show()
```

```
Top 5 words for each VADER sentiment category:

Very Positive: [('gorilla', 199), ('like', 162), ('zoo', 161), ('kid', 149), ('parents', 126)]

Very Negative: [('gorilla', 253), ('parents', 223), ('kid', 217), ('child', 183), ('zoo', 171)]

Neutral: [('gorilla', 46), ('zoo', 44), ('parents', 42), ('kid', 36), ('child', 34)]

Negative: [('gorilla', 101), ('kid', 100), ('parents', 74), ('zoo', 69), ('people', 57)]

Positive: [('gorilla', 11), ('parents', 10), ('child', 8), ('going', 7), ('kid', 7)]

Top 5 words for each EMPATH sentiment category:

Neutral: [('gorilla', 77), ('zoo', 76), ('like', 64), ('people', 59), ('dont', 54)]

Play: [('kid', 356), ('child', 249), ('gorilla', 233), ('parents', 187), ('zoo', 130)]

Negative_emotion: [('gorilla', 170), ('parents', 143), ('zoo', 137), ('people', 105), ('fucking', 102)]

Positive_emotion: [('zoo', 35), ('animals', 34), ('people', 32), ('better', 32), ('family', 26)]

Hate: [('gorilla', 40), ('kill', 29), ('zoo', 28), ('bad', 25), ('parents', 23)]

Trust: [('gorilla', 18), ('people', 17), ('agree', 16), ('think', 15), ('zoo', 13)]

Love: [('feel', 25), ('like', 17), ('zoo', 12), ('love', 12), ('gorilla', 12)]

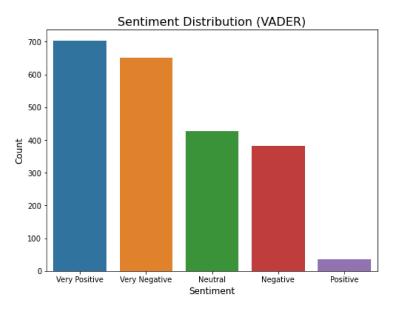
Pain: [('fall', 24), ('gorilla', 19), ('zoo', 13), ('parents', 12), ('screaming', 11)]

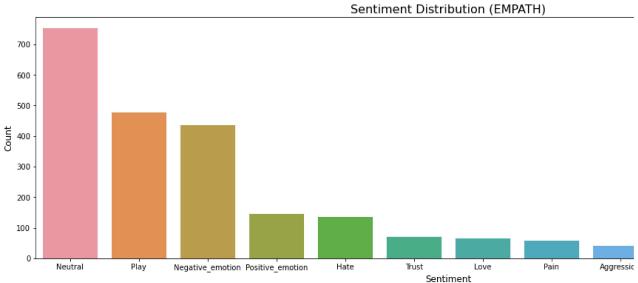
Aggression: [('gorilla', 11), ('animal', 10), ('would', 9), ('gorillas', 8), ('species', 8)]

Fear: [('would', 5), ('face', 5), ('dont', 5), ('think', 4), ('shit', 4)]

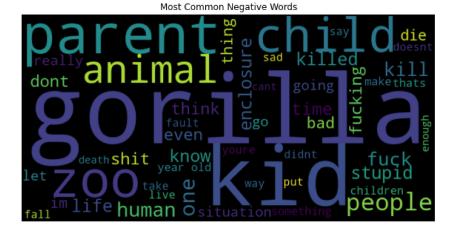
Anger: [('shame', 7), ('parents', 5), ('animal', 3), ('damn', 3), ('childs', 2)]

Joy: [('happened', 1), ('reddit', 1), ('like', 1), ('4chan', 1), ('without', 1)]
```





Tanimal imknow youre life will better some old say 80 love well better some well better some well better some will be the some one don't be the some well better some well bette



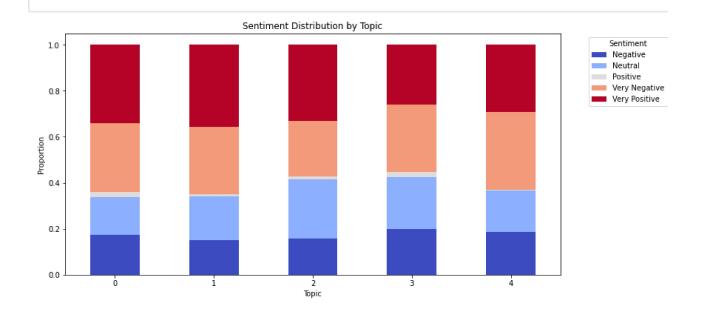
```
    import gensim

In [161]:
              from gensim import corpora
              from gensim.models.ldamodel import LdaModel
              import pandas as pd
              # Load the preprocessed data
              processed_df = pd.read_csv("Sentimental_analysis_processeddata.csv")
              # Tokenize the comments
              tokenized_comments = processed_df["Comment"].apply(lambda x: x.split())
              # Create a dictionary and a corpus from the tokenized comments
              dictionary = corpora.Dictionary(tokenized_comments)
              corpus = [dictionary.doc2bow(comment) for comment in tokenized_comments]
              # Train the LDA model
              num topics = 5
              lda_model = LdaModel(corpus, num_topics=num_topics, id2word=dictionary, passes=20, random_state=42)
              # Display the top keywords for each topic
              for idx, topic in lda_model.print_topics(-1):
                  print(f"Topic: {idx} \nWords: {topic}\n")
              Topic: 0
              Words: 0.019*"parents" + 0.017*"kid" + 0.012*"zoo" + 0.012*"child" + 0.010*"dont" + 0.010*"know" + 0.009*"year"
              Words: 0.014*"people" + 0.008*"animals" + 0.008*"like" + 0.007*"zoo" + 0.007*"take" + 0.007*"one" + 0.006*"reaso
              Words: 0.015*"like" + 0.013*"youre" + 0.011*"kid" + 0.010*"people" + 0.007*"get" + 0.007*"time" + 0.007*"dont" +
              Words: 0.028*"gorilla" + 0.015*"would" + 0.015*"gorillas" + 0.014*"humans" + 0.014*"human" + 0.012*"kill" + 0.01
              Topic: 4
              Words: 0.031*"gorilla" + 0.018*"child" + 0.015*"parents" + 0.013*"zoo" + 0.009*"zoos" + 0.009*"fault" + 0.009*"a
In [156]: ▶
              # Assign the dominant topic for each comment
              def dominant topic(ldamodel, corpus):
                  dominant_topic_list = []
                  for doc in corpus:
                      topic probs = ldamodel.get document topics(doc)
                      dominant_topic = sorted(topic_probs, key=lambda x: x[1], reverse=True)[0][0]
                      dominant_topic_list.append(dominant_topic)
                  return dominant topic list
              processed_df["Dominant_topic"] = dominant_topic(lda_model, corpus)
              # Perform VADER sentiment analysis
              sia = SentimentIntensityAnalyzer()
              processed_df["Sentiment"] = processed_df["Comment"].apply(lambda x: sia.polarity_scores(x)["compound"]).apply(cl
              # Analyze the sentiment distribution for each topic
              sentiment_by_topic = processed_df.groupby(["Dominant_topic", "Sentiment"]).size().unstack()
              sentiment_distribution = sentiment_by_topic.apply(lambda x: x / x.sum(), axis=1)
              print(sentiment distribution)
              Sentiment
                              Negative Neutral Positive Very Negative Very Positive
              Dominant topic
              0
                              0.172161 0.163614 0.023199
                                                                 0.300366
                                                                                0.340659
              1
                              0.150685 0.188356 0.010274
                                                                 0.294521
                                                                                0.356164
                              0.158209 0.256716 0.011940
                                                                 0.241791
                                                                                0.331343
              2
                              0.197917 0.226562 0.020833
                                                                 0.294271
                                                                                0.260417
              3
                              0.185484 0.180108 0.005376
                                                                 0.336022
                                                                                0.293011
              4
```

```
In [162]:

    import matplotlib.pyplot as plt

              import seaborn as sns
              def get_topic_keywords(ldamodel, num_topics):
                  topic_keywords = {}
                  for idx, topic in ldamodel.show_topics(num_topics, formatted=False):
                      topic_keywords[idx] = ", ".join([word[0] for word in topic])
                  return topic_keywords
              topic_keywords = get_topic_keywords(lda_model, num_topics)
              def plot_sentiment_distribution(df, title, topic_keywords):
                  ax = df.plot(kind="bar", stacked=True, figsize=(12, 6), colormap="coolwarm")
                  ax.set_xlabel("Topic")
                  ax.set_ylabel("Proportion")
                  ax.set_title(title)
                  ax.legend(title="Sentiment", bbox_to_anchor=(1.05, 1), loc='upper left')
                  plt.xticks(rotation=0)
                  plt.show()
              plot_sentiment_distribution(sentiment_distribution, "Sentiment Distribution by Topic", topic_keywords)
```



```
In [160]:

    import pyLDAvis

                                                   import pyLDAvis.gensim_models as gensimvis
                                                  # Prepare the visualization data
                                                  vis_data = gensimvis.prepare(lda_model, corpus, dictionary)
                                                  # Display the interactive visualization
                                                  pyLDAvis.display(vis_data)
                                                  {\tt C:\Users\sindh\anaconda3\lib\site-packages\pyLDAvis\prepare.py:243: Future Warning: In a future version of pandal part of the part of
                                                   'labels' will be keyword-only.
                                                         default_term_info = default_term_info.sort_values(
             Out[160]:
                                                     Selected Topic: 1
                                                                                                                             Previous Topic
                                                                                                                                                                                Next Topic
                                                                                                                                                                                                                         Clear Topic
                                                                                                                                                                                                                                                                                                                                                                  Slide to adjust relevance metric:
                                                                                                                                                                                                                                                                                                                                                                                                     \lambda = 0.43
                                                                                                  Intertopic Distance Map (via multidimensional scaling)
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                                                                                                                                                                                                                                                                                                                                       someone
                                                                     Marginal topic distribution
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                                                                                                                                                                                                                                                                                                                                                                                       Estimated term frequency withir
                                                                                                                                2%
                                                                                                                                                                                                                                                                                                                                                               1. saliency(term w) = frequency(w) * [sum
                                                                                                                                5%
                                                                                                                                                                                                                                                                                                                                                              2. relevance(term w | topic t) = \lambda * p(w | t) +
                                                                                                                                10%
```

In [ ]: M