Project Report: Reddit Cryptocurrency Sentiment Analysis

AUTHOR Wenbo Liu (wl2939) PUBLISHED February 19, 2025

1. Introduction

Project Overview

This project aims to analyze sentiment trends in Reddit cryptocurrency discussions. The goal is to collect, clean, preprocess, and analyze textual and numerical data from Reddit posts, focusing on Bitcoin-related discussions. The final analysis highlights patterns in engagement metrics (upvotes, comments) and explores their relationship with sentiment and time-based features.

Data Source

The data is collected from the **r/cryptocurrency** subreddit using the **Reddit API (praw)**. The dataset contains metadata, including: - **Post Title & Content** - **Upvotes & Comments** - **Post Creation Time** - **Post URL**

2. Data Acquisition

Methodology

We used the praw library to fetch Reddit posts related to Bitcoin using the subreddit.search() function. The search query was set to "Bitcoin" to ensure relevance.

Code Implementation

Dataset Sample

Post_ID	Title	Upvotes	Comments	Timestamp
1gqafju	Bitcoin cycle analysis	3577	701	2024-01-01
1h6yoqp	Bitcoin hits 100K	19972	342	2024-01-02

3. Data Cleaning & Preprocessing

Cleaning Steps

- Convert timestamps to datetime format
- Remove special characters & URLs from text
- Handle missing values (content field filled with "No content")
- Remove duplicate posts based on URLs

```
def clean_reddit_data(df):
    df = df.copy()
    df.columns = df.columns.str.lower().str.replace(" ", "_")
    df["timestamp"] = pd.to_datetime(df["timestamp"], unit="s")
    df.drop_duplicates(subset=["url"], inplace=True)
    df["content"].fillna("No content", inplace=True)
    return df
```

Preprocessing Steps

- Feature scaling (Log Transform for Upvotes & Comments)
- Extract time-based features (Hour, Day, Weekend Flag)
- Categorical encoding for content

```
import numpy as np
from sklearn.preprocessing import StandardScaler

def preprocess_data(df):
    df = df.copy()
    df['day_of_week'] = df['timestamp'].dt.dayofweek
    df['hour_of_day'] = df['timestamp'].dt.hour
    df['is_weekend'] = df['day_of_week'].apply(lambda x: 1 if x >= 5 else 0)
    df['log_upvotes'] = np.log1p(df['upvotes'])
    df['log_comments'] = np.log1p(df['comments'])
    scaler = StandardScaler()
```

```
df[['upvotes', 'comments']] = scaler.fit_transform(df[['upvotes', 'comments']])
return df
```

4. Exploratory Data Analysis (EDA)

Key Insights from Visualizations

- Upvotes & comments have a right-skewed distribution.
- Weekday vs. Weekend engagement shows higher upvotes on weekends.
- Correlation between upvotes and comments is positive.

Visualization: Upvotes Distribution

```
import seaborn as sns
import matplotlib.pyplot as plt

sns.histplot(df['upvotes'], bins=30, kde=True, color='blue')
plt.title("Distribution of Upvotes")
plt.show()
```

Visualization: Correlation Heatmap

```
sns.heatmap(df.corr(numeric_only=True), annot=True, cmap='coolwarm')
plt.title("Correlation Matrix")
plt.show()
```

5. Feature Engineering

Added Features

Feature	Description		
log_upvotes	Log transformation of upvotes because of right skewed distribution		
log_comments	Log transformation of comments because of right skewed distribution		
title_len	Number of words in post title		
content_len	Number of words in post content		
is_working_hours	Whether the post was made during business hours (9AM - 5PM)		
engagement_score	Weighted score of upvotes & comments		
upvote_to_comment_ratio	Ratio of upvotes to comments		
has_bitcoin	Whether "Bitcoin" appears in the title		

6. Conclusion & Future Work

Findings

- Sentiment polarity influences engagement (higher upvotes for positive sentiment).
- Posts with "Bitcoin" in the title tend to receive higher upvotes.

• Weekends show higher engagement.

Next Steps

- Apply NLP techniques (TF-IDF, LLM sentiment analysis).
- Build predictive models for Bitcoin price movement based on Reddit sentiment.
- Experiment with fine-tuned LLMs for finance-specific sentiment classification.

This report summarizes the entire process, from data collection to analysis. The next phase will explore predictive modeling using the cleaned dataset.