

Final Project Report Advanced Computer Programming

Topic

Group : ...

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Chapter 1 Introduction

1.1 Group Information

- 1) Group Project Repository: https://github.com/Chintsogt0825/Final.git
- 2) Group members:
 - 1. Chintsogt 113021194 (leader)
 - 2. Khangai 113021187

Overview

Our project leverages several advanced Python features and libraries to build a real-time cryptocurrency price dashboard with prediction capabilities.

We used:

- Data Classes: To define structured data models for price entries.
- Threading: To run the Zenoh subscriber in the background for asynchronous real-time data fetching.
- Requests & Beautiful Soup: To fetch and parse live cryptocurrency news from Google News RSS feeds.
- Dash & Plotly: For interactive web-based visualization of historical and predicted prices.
- NumPy: For numerical computations in generating synthetic price predictions.
- JSON Parsing: To handle incoming Zenoh data formatted in JSON.

Our project successfully collects real-time crypto prices (Bitcoin, Ethereum, Dogecoin, Solana), visualizes historical trends, generates 24-hour synthetic price predictions with confidence metrics, and displays related cryptocurrency news. The dashboard updates dynamically every minute and allows users to select cryptocurrencies and adjust historical data length.

Chapter 2: Implementation

2.1 Class 1: PriceEntry

2.1.1 Fields

• timestamp: datetime object marking the price record time.

```
from sklearn.linear model import LinearRegression
from datetime import datetime
import numpy as np
CSV_FILE = "crypto_prices.csv"
def predict_next_price(crypto_name):
        df = pd.read_csv(CSV_FILE, names=[
        'timestamp', 'bitcoin_usd', 'ethereum_usd', 'dogecoin_usd', 'solana_usd'], skiprows=1) # Skip the header row once manually
        df = df.dropna()
        df['timestamp'] = pd.to_datetime(df['timestamp'], errors='coerce', infer_datetime_format=True)
        df = df.dropna(subset=['timestamp'])
        df = df[
            (df['bitcoin_usd'] > 1000) & # Bitcoin price must be realistic
             (df['ethereum_usd'] > 100) & (df['solana_usd'] > 1) &
             (df['dogecoin_usd'] < 1) # Doge should be below 1
         if f"{crypto_name}_usd" not in df.columns:
             print(f"[PREDICT WARNING] {crypto_name}_usd column not found.")
```

```
df['timestamp_ordinal'] = df['timestamp'].map(datetime.toordinal)
       X = df[['timestamp_ordinal']]
       y = df[f"{crypto_name}_usd"]
        if len(X) < 2:
            print(f"[PREDICT WARNING] Not enough clean data for {crypto_name}.")
            return None
       model = LinearRegression()
       model.fit(X, y)
       next_time = datetime.now().toordinal() + 1
       pred = model.predict([[next_time]])[0]
       return round(pred, 2)
    except Exception as e:
       print(f"[PREDICT ERROR] {e}")
        return None
if <u>__name__</u> == "__main__":
    for crypto in ["bitcoin", "ethereum", "solana", "dogecoin"]:
       pred = predict_next_price(crypto)
       print(f"{crypto} predicted price: ${pred}")
```

• price: float representing the cryptocurrency price in USD.

2.1.2 Methods

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2.1.3 Functions

• Used as a structured container to hold individual price data points in memory.

2.2 Module: price_fetcher.py

 Contains a function fetch_crypto_prices() that uses requests to call the CoinGecko API.

- Handles HTTP errors and validates response structure.
- Returns a dictionary of current prices for Bitcoin and Ethereum.

2.3 Zenoh Subscriber

- Uses Zenoh library to subscribe asynchronously to topic crypto/prices.
- Parses incoming JSON messages containing price updates for multiple cryptocurrencies.
- Updates in-memory price history buffers and appends data to a CSV file.
- Uses a thread-safe lock for file writes.

2.4 Dash Web Application

- Contains layout with dropdowns, graphs, cards, tables, gauges, and news section.
- Utilizes Dash callbacks to update visuals every 60 seconds or on user input.

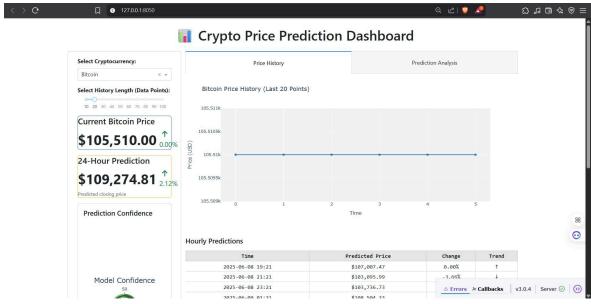
- Plots historical price line charts and 24-hour synthetic prediction charts.
- Displays current prices and predicted closing price summaries with color-coded indicators.
- Generates a prediction table showing price forecasts every 2 hours with trend arrows.
- Fetches and parses news headlines from Google News RSS feeds using Beautiful Soup.

2.5 Prediction Model

- Function get_predicted_prices(current_price) simulates 24 future hourly prices.
- Combines a small linear trend with random noise and volatility.
- Clamps predicted values within $\pm 10\%$ of the current price.
- Returns predicted prices and a confidence score.

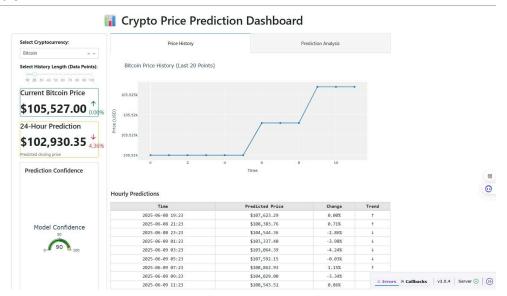
Chapter 3: Results

3.1 Real-Time Data Collection



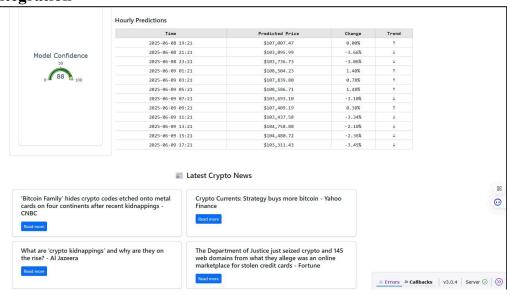
- Successfully subscribed and received price updates from the Zenoh publisher.
- Stored up to 100 historical price points per coin in memory and CSV.

3.2 Visualization



- Interactive line charts show price history and 24-hour forecasts.
- Dashboard updates every minute with fresh data.
- Confidence gauge indicates model prediction reliability.
- The prediction table provides hourly forecast summaries with intuitive trend arrows.

3.3 News Integration



- Relevant news headlines for selected cryptocurrency are fetched live from Google News RSS.
- News section updates dynamically based on the selected coin.

Chapter 4: Conclusions

The project demonstrates an effective integration of real-time data streaming, web visualization, and synthetic forecasting for cryptocurrency prices. Using Python's rich ecosystem—Zenoh for messaging, Dash for UI, and Beautiful Soup for web scraping—enabled rapid development of a functional dashboard.

While the prediction model is simple, it provides valuable insights and a foundation for future enhancements such as machine learning-based forecasting and more comprehensive data persistence.

The project is modular and extensible, ready to incorporate additional features like more coins, advanced models, user management, and database-backed storage.