



## Final Project

# Clustering Analysis on Gold Price (XAU/USD) Using K-Means

Arranged by:  
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## I INTRODUCTION

This project applies unsupervised machine learning using K-Means clustering to segment hourly gold price (XAU/USD) data. The goal is to uncover patterns in market behavior based on price movement, volatility, trading volume, and technical indicators.

### **Objective:**

To explore and cluster hourly gold price data (XAU/USD) from May 1–19, 2025, using unsupervised machine learning (K-Means).

### **Tools Used:**

- Python (pandas, matplotlib, seaborn, sklearn)
- Jupyter Notebooks



## II DATASET DESCRIPTION

This dataset provides hourly historical price data for XAU/USD (Gold vs. US Dollar) from May 1, 2025, to May 19, 2025.

**Each row represents a single hour and includes:**

- Date and Time
- Open, High, Low, Close ( OHLC ) prices
- Tick Volume (Number of price changes)
- Spread

**Our Dataset Link :**

**Dataset XAUUSD\_Hourly.csv**

**Resulting File**

Cleaned dataset saved as:

**XAUUSD\_Hourly\_Cleaned.csv**

## III DATA CLEANING

### 1 Checked for Missing Values

To make sure there are no gaps or null values in important columns.

- `df.isnull().sum()` → Result: No missing values

### 2 Removed Irrelevant & Redundant Columns

To simplify the dataset and keep only useful information.

- Volume column: always 0 → no useful data → removed.
- Date and Time columns: already combined into datetime → now redundant → removed.

### 3 Datetime Standardization

To ensure the datetime column is in a usable format for time-based analysis.

- `pd.to_datetime()` to convert the datetime column.

### 4 Spread Standardization

To make sure trading conditions in the dataset are consistent.

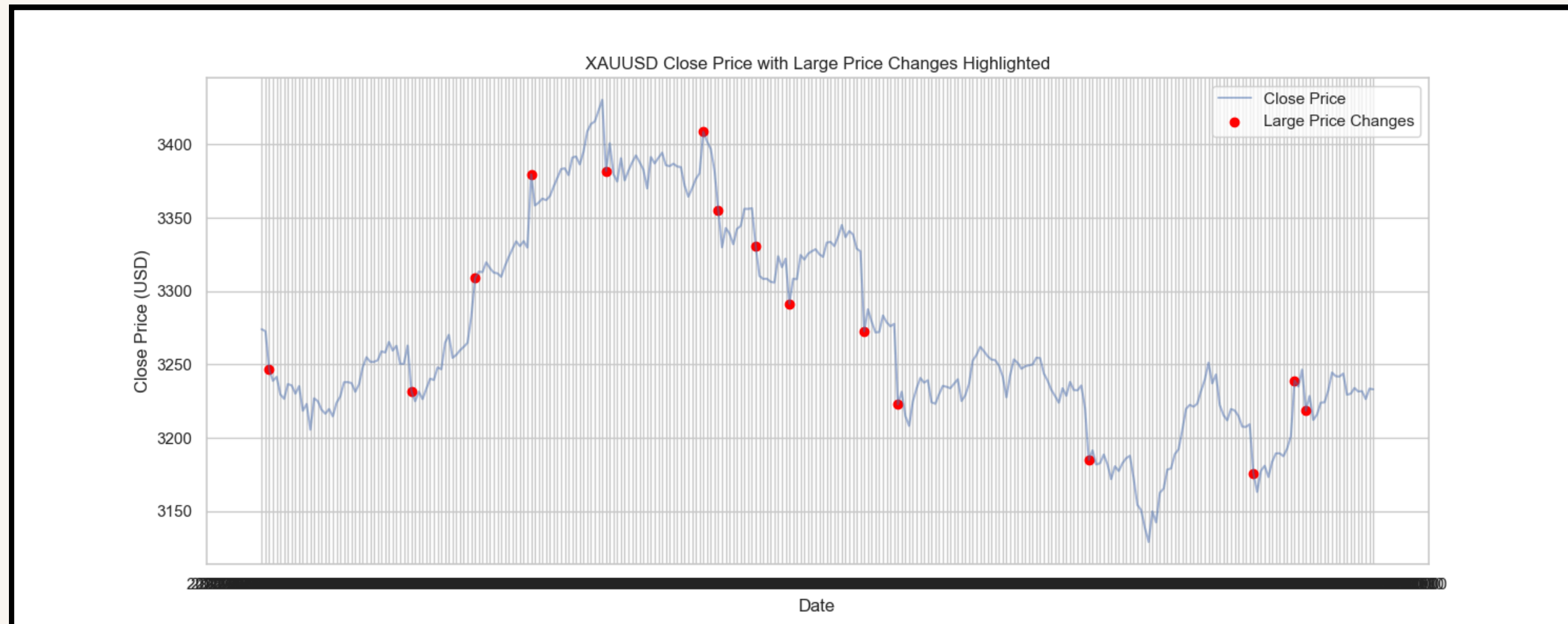
- The Spread column mostly had value 7, but some values were 6 or 8.
- To keep consistency, all spread values were replaced with the mode = 7.

### 5 Duplicate & Outlier Check

To remove any repeated or extreme data that could skew analysis.

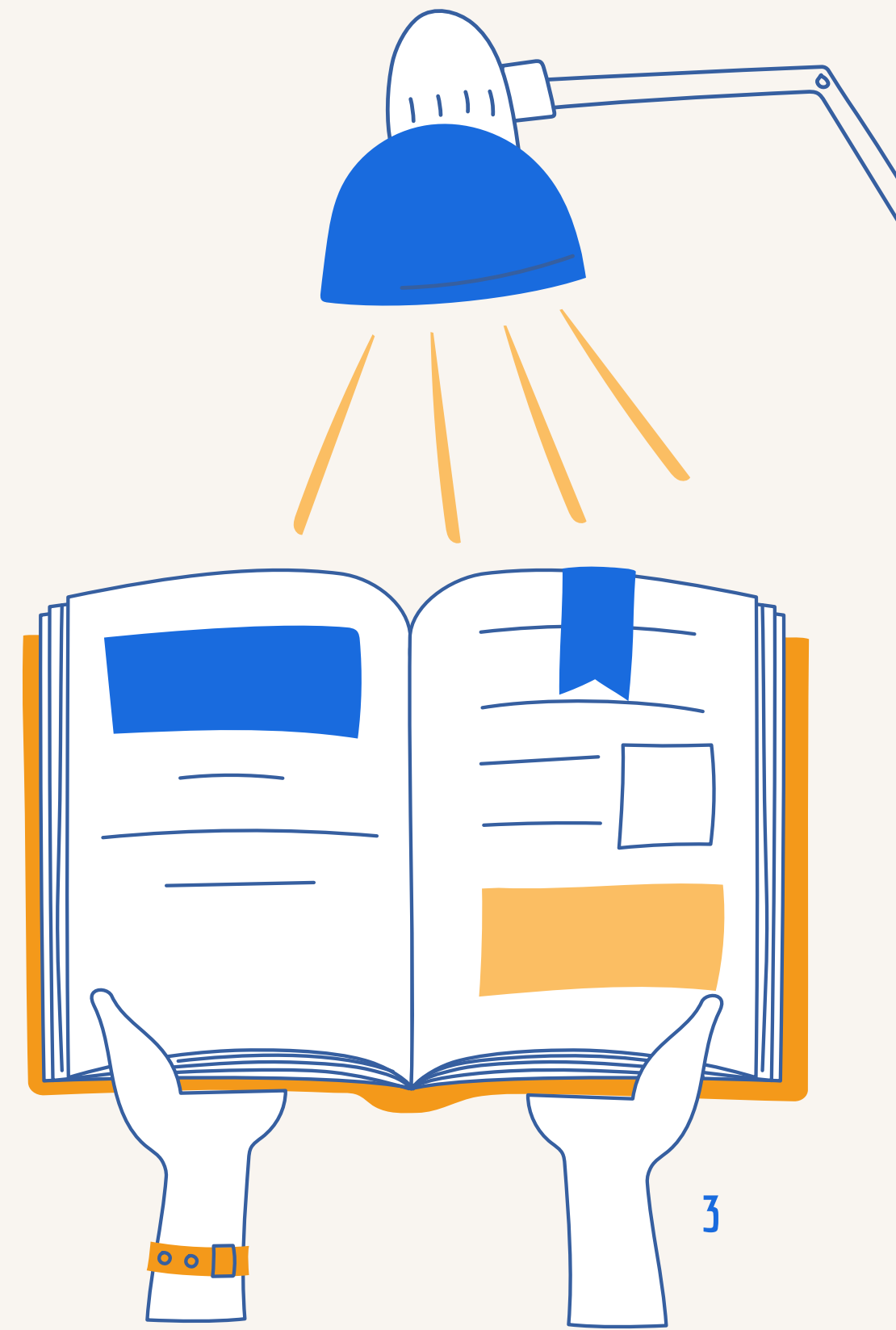
- Used `drop_duplicates()` → No duplicates
- Outlier check using IQR → No extreme outliers found in price

## IV EXPLORATORY DATA ANALYSIS (EDA)



The Exploratory Data Analysis revealed key patterns in price behavior, volatility, trading activity, and technical indicators, offering insights into market trends and potential predictive signals.

- Avg. close price: \$3271.96 with moderate volatility
- High correlation across price columns
- 15 major price swings linked to volatility peaks
- RSI and SMA indicate potential trend changes

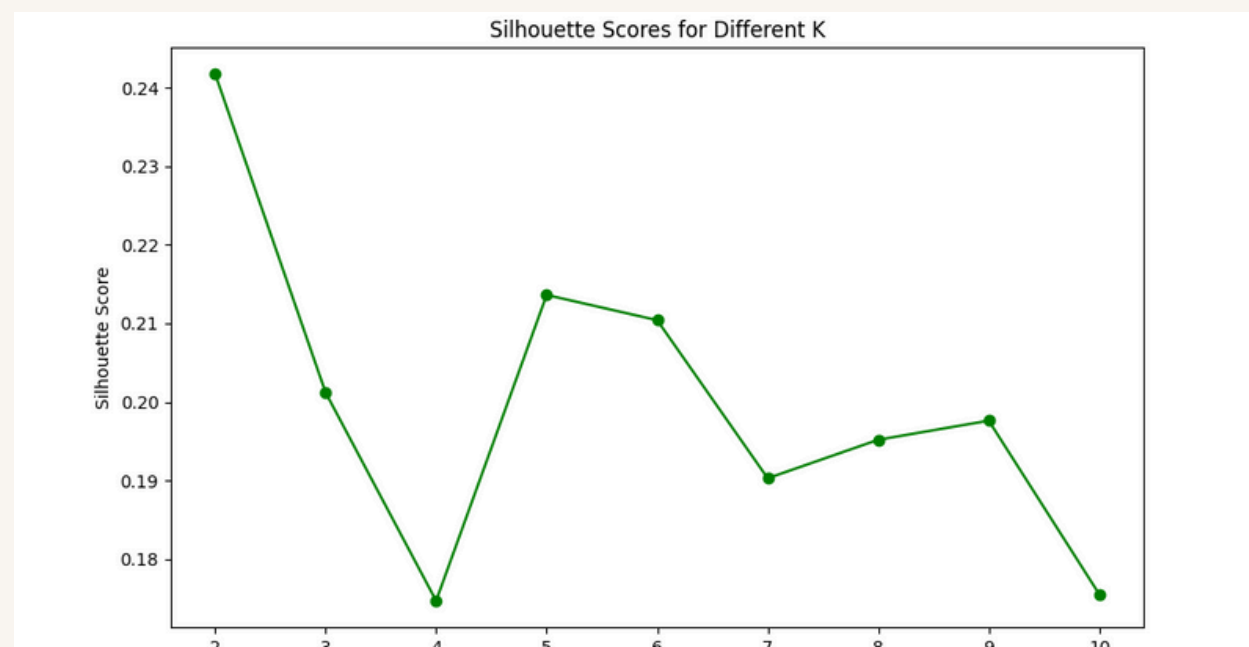


## V UNSUPERVISED LEARNING – K-MEANS CLUSTERING

### SILHOUETTE SCORE PLOT

BEST CLUSTERING QUALITY AT  $K = 2$

WE USED  $K = 4$  TO CAPTURE RICHER MARKET REGIMES.

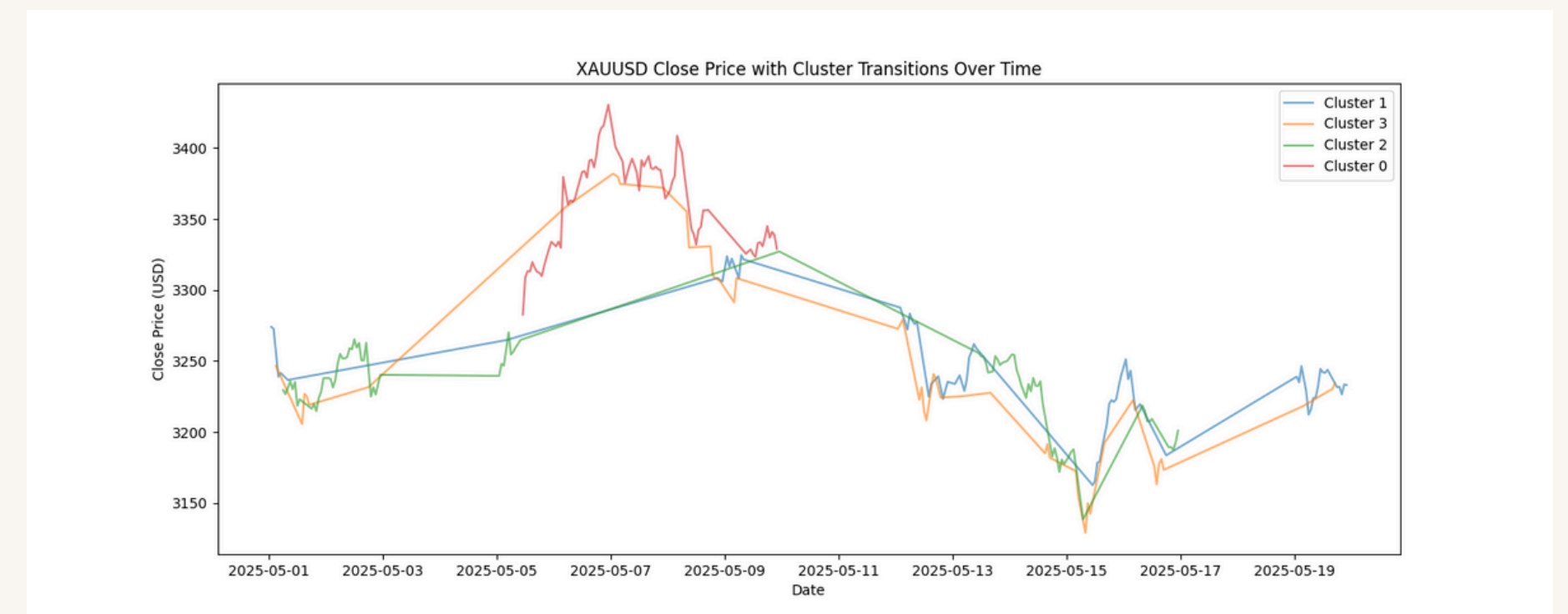


### CLUSTER CHARACTERISTICS

0	High Price, High RSI → Bullish Trend
1	Mid Price, High Return, Mixed Momentum
2	Low Price, Low Volatility → Stable
3	High Volatility, Low RSI → Bearish/Correction Periods

### Features Used:

- **Close, Returns, Volatility, High\_Low\_Range, TickVolume, RSI**
- **All features scaled with StandardScaler.**

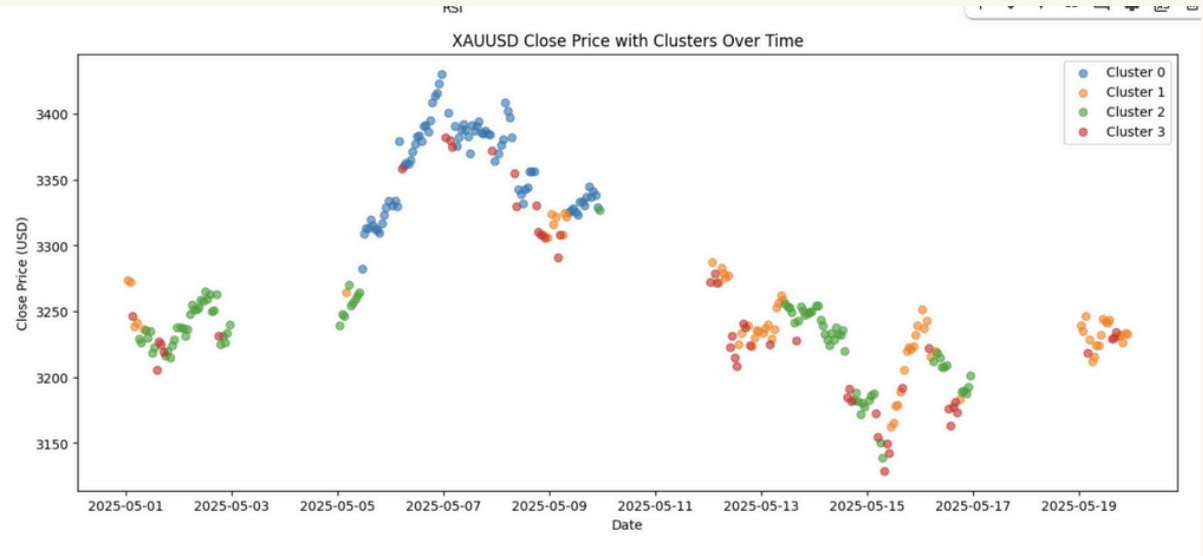
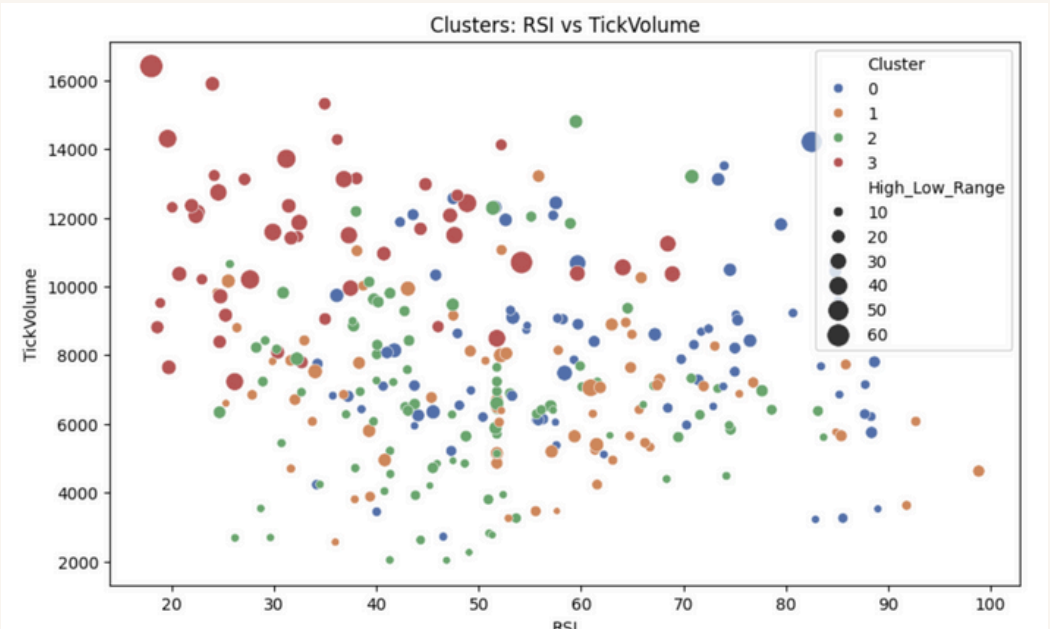
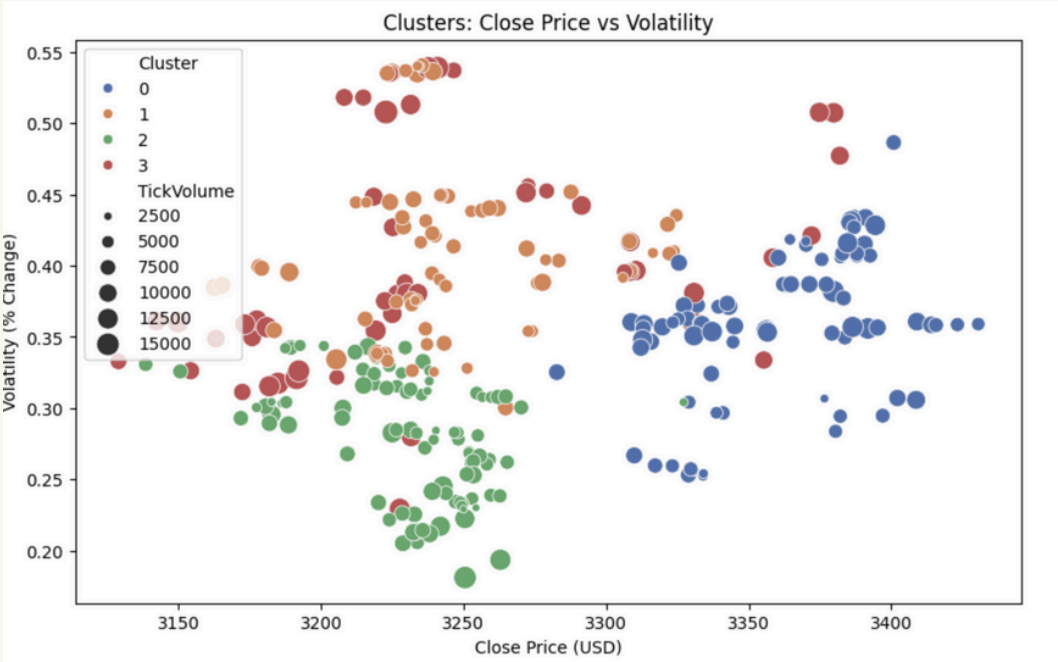


- Cluster Transition Plot (cluster\_transitions\_timeseries.png):
- Shows how the market transitioned between clusters over time.
- Clear periods of bullish (Cluster 0) and bearish (Cluster 3) trends.



VI CLUSTER SUMMARY

Cluster	Price Level	Volatility	RSI	Activity
0	High Price	Moderate	Bullish	High
1	High Price	Moderate	Bullish	High
2	High Price	Moderate	Bullish	High
3	High Price	Moderate	Bullish	High



- **CLUSTER 0:** HIGHEST PRICES, MODERATE VOLATILITY, STRONG TRADING ACTIVITY, BULLISH TREND.
- **CLUSTER 1:** MID PRICES, HIGHEST RETURNS, HIGHER VOLATILITY, MIXED MOMENTUM.
- **CLUSTER 2:** LOWER PRICES, STABLE RETURNS, LOWEST VOLATILITY, NEUTRAL MARKET.
- **CLUSTER 3:** WIDE PRICE RANGE, NEGATIVE RETURNS, HIGH VOLATILITY, BEARISH TREND.

# VII BUSINESS / TRADING INSIGHT

We identified 4 distinct market clusters using unsupervised learning:

Cluster	Market Behavior	Suggested Strategy
0	High price, bullish, high RSI	Trend-following (Buy & Hold)
1	Mid price, high return, volatile	Short-term Momentum Trading
2	Low volatility, stable price	Low-risk or Hold strategy
3	High volatility, bearish, low RSI	Mean Reversion (Buy Low, Sell Fast)

**Note:** Each cluster gives insight into the best action traders can take depending on market behavior.

## Key Insights & Real-World Use

Content:

 Indicators:

- $RSI > 70 \rightarrow$  Overbought  $\rightarrow$  Sell
- $RSI < 30 \rightarrow$  Oversold  $\rightarrow$  Buy
- High Volume & Volatility  $\rightarrow$  Market reacting to news/events

 Business Use:

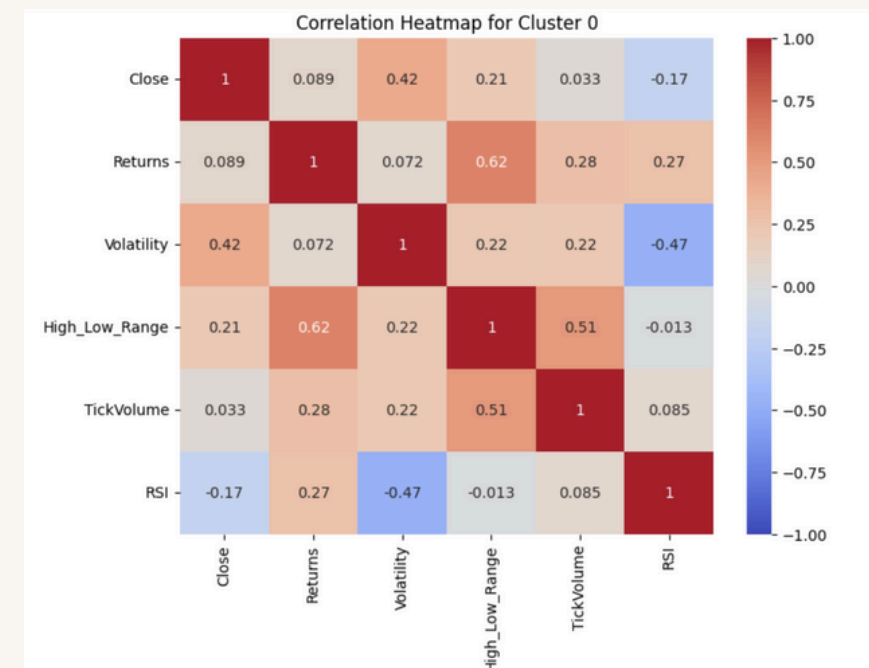
- Fast market classification
- Supports trading alerts & strategies
- Useful for fintech dashboards & trading bots

## VIII CORRELATION ANALYSIS

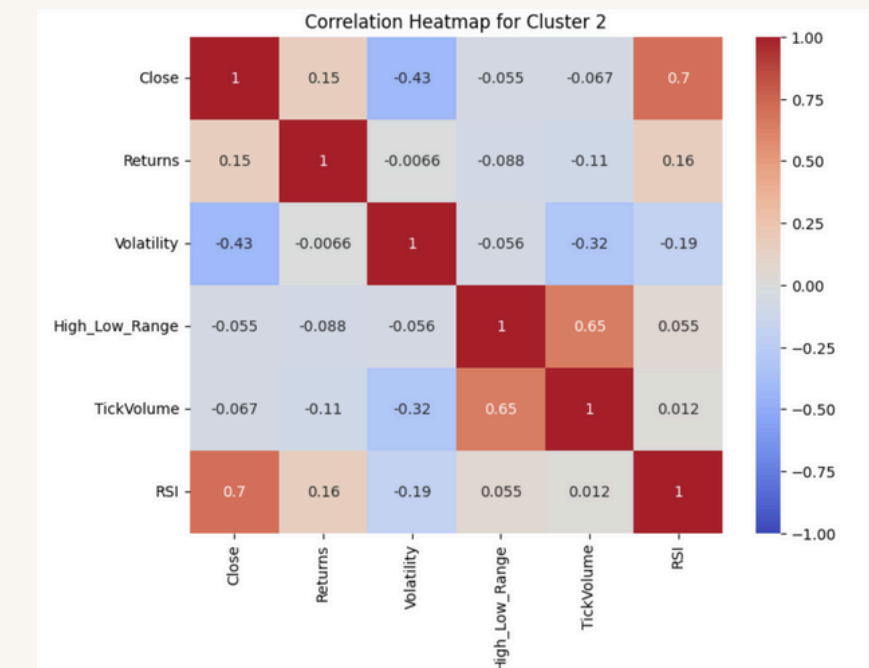
THE CORRELATION ANALYSIS REVEALED KEY RELATIONSHIPS BETWEEN MARKET INDICATORS, BOTH OVERALL AND WITHIN SPECIFIC CLUSTERS.

- CLOSE AND RSI SHOW A MODERATE POSITIVE CORRELATION, LINKING PRICE WITH MOMENTUM
- TICKVOLUME STRONGLY CORRELATES WITH HIGH\_LOW\_RANGE, SUGGESTING ACTIVE MARKETS WIDEN SPREADS
- RETURNS AND VOLATILITY SHOW WEAK NEGATIVE CORRELATION
- CLUSTER HEATMAPS HIGHLIGHT UNIQUE INTRA-CLUSTER DYNAMICS FOR DEEPER INSIGHTS

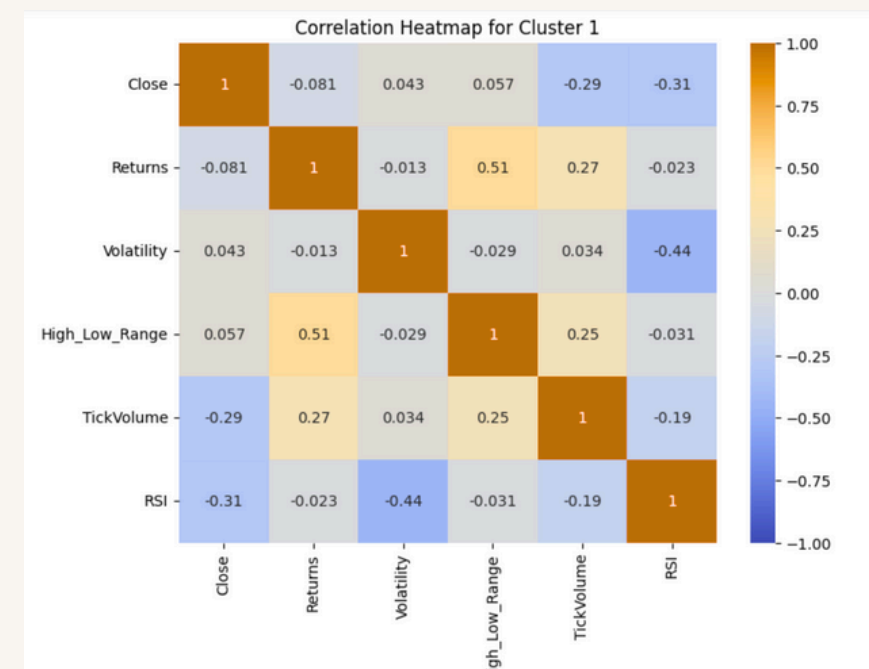
Cluster 0



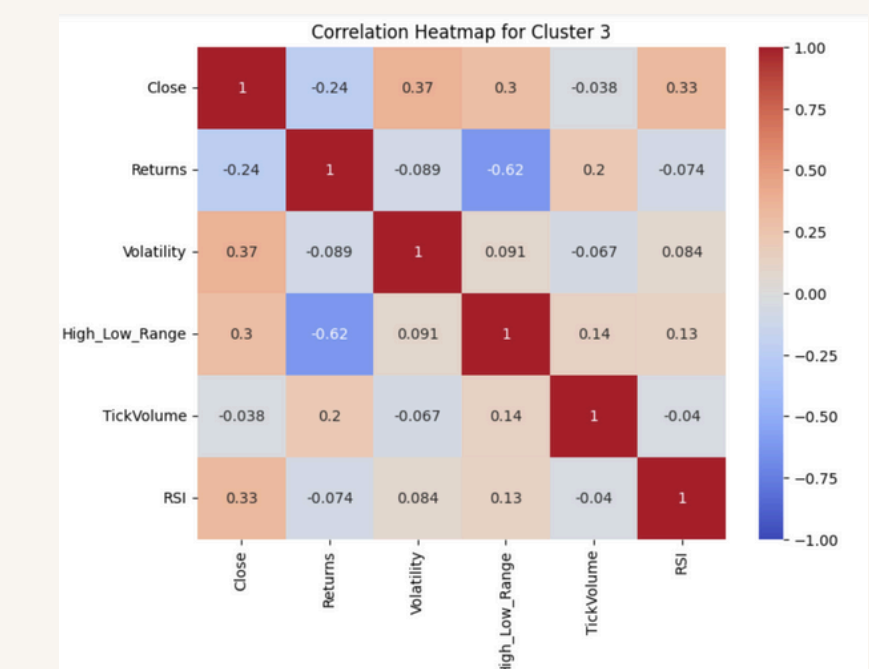
Cluster 2



Cluster 1



Cluster 3





## VIII CONCLUSION & NEXT STEPS 🐸



This project showed that K-Means clustering is effective in analyzing hourly gold price data by identifying four distinct market patterns: bullish, bearish, stable, and volatile. The combination of data cleaning, EDA, and unsupervised learning helped simplify complex financial data and reveal valuable insights into price behavior and market momentum.

### ***Next Steps:***

- Extend the dataset for broader, long-term analysis
- Use Elbow and Silhouette methods to refine cluster accuracy
- Try other clustering methods (e.g., DBSCAN, HDBSCAN) for deeper insights
- Explore supervised models to predict market trends from cluster features



Welcome for any questions and

**Thank you for  
Attention**

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