

### **Overview**

This project utilizes predictive modeling by analyzing historical data to predict whether an investor in the foreign exchange market should invest in the S&P 500 using their desired currency for a specified period. The ultimate objective is to enhance algorithmic trading strategies by forecasting market trends.



# Hypothesis

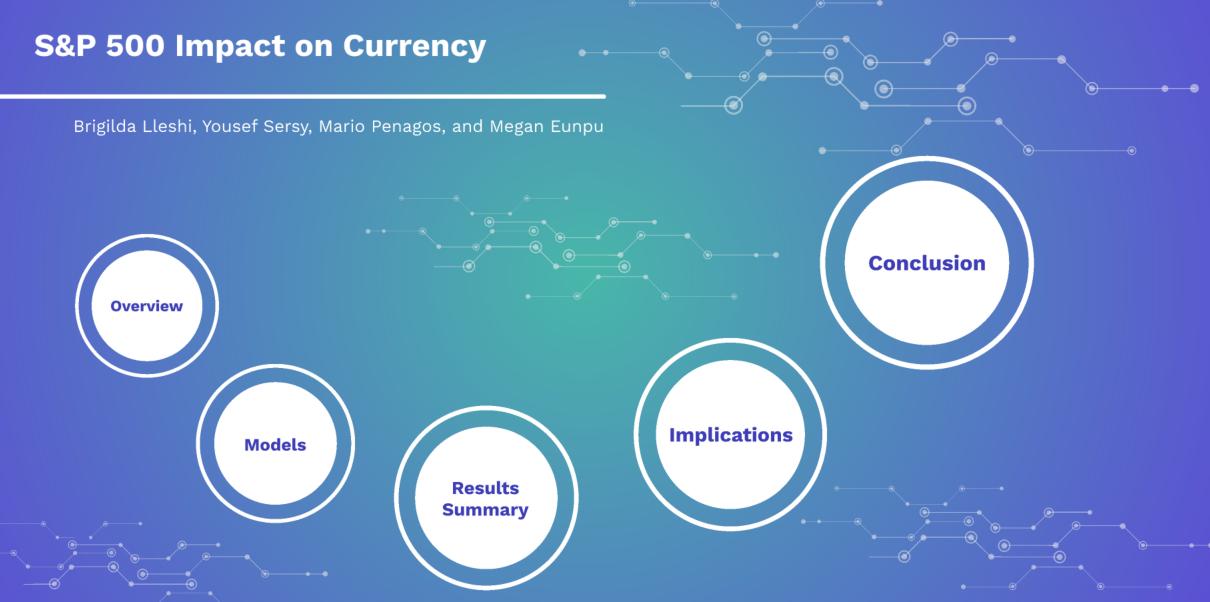
Currency rates are significantly influenced by the movements of the S&P 500, and this relationship can be accurately predicted using machine learning models.

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### **Selected Models**

Simple, interpretable, good for binary classification.

Logistic Regression

Robust to overfitting, handles non-linear data.

Random Forest

## **Currencies Used in Model**



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## Data

#### **Data Collection**

 Yahoo Finance for currencies and SPY (S&P 500 ETF) for time period of 05/01/2019 - 05/01/2024 (MM/DD/YYYY)

### **Data Processing**

- · Download and clean data
- Calculate percentage changes
- Create target signals (SPY direction)

### **Balancing Dataset**

SMOTE (Synthetic Minority Over-Sampling Technique)

# **Model Training**

#### **Process**

- Split data into training and testing sets (60% and 40%, respectively)
- Train Logistic Regression and Random Forest models

#### **Performance Evaluation (Metrics)**

- Accuracy
- Classification Report (Precision, Recall, F1-Score)

#### Results

- Logistic Regression
- Random Forest





# **CODE DEMO**



### **Selected Models**

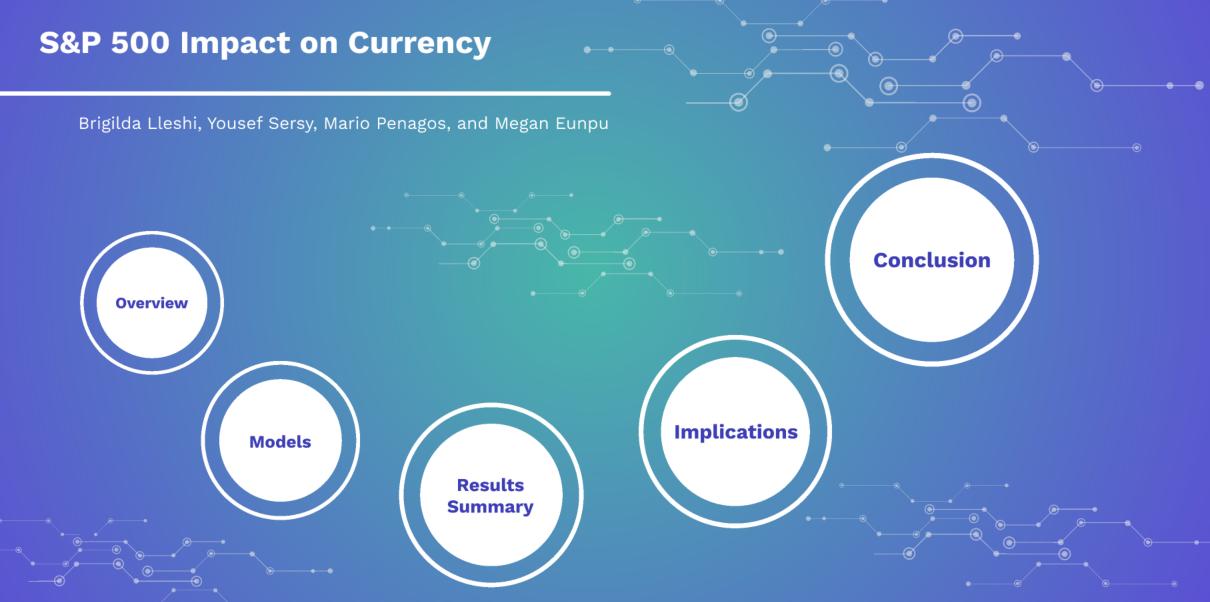
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### **Results Summary**

Both models have similar accuracy (52%), indicating they are performing only slightly better than random guessing.

#### Logistic Regression

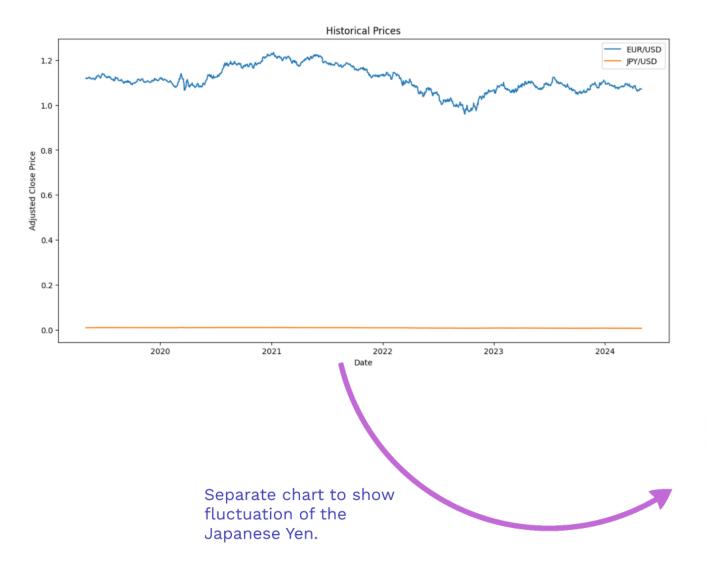
- Better at identifying when the S&P 500 does not go up (higher recall for class 0 at 72%).
- Less effective at identifying when the S&P 500 goes up (lower recall for class 1 at 32%).

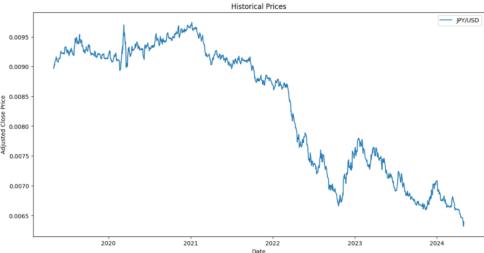
#### Random Forest

- More balanced in identifying both classes compared to Logistic Regression.
- Slightly better recall for class 1 at 48%, indicating it is better at identifying days when the S&P 500 goes up compared to Logistic Regression.

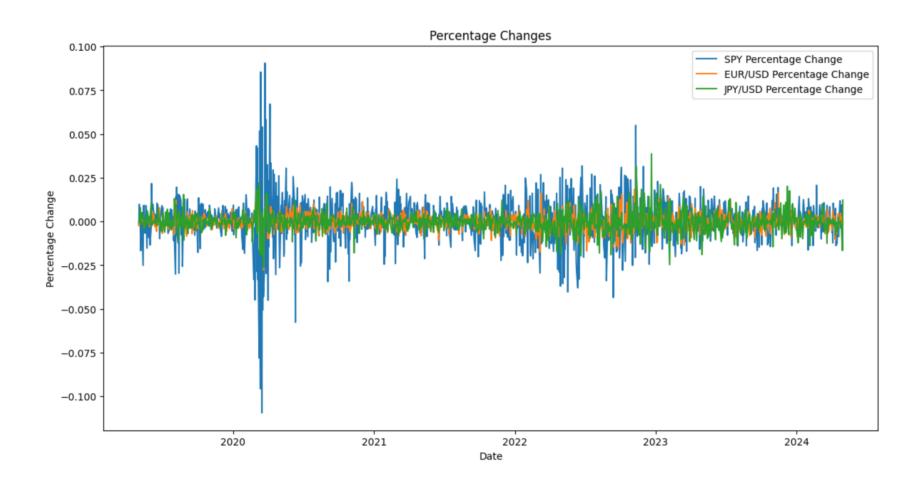
Logistic Regression accuracy: 0.52 Random Forest accuracy: 0.50 Logistic Regression Classification Report:				
	precision	recall	f1-score	support
0	0.51	0.72	0.60	268
1	0.55	0.32	0.41	276
accuracy			0.52	544
macro avg	0.53	0.52	0.50	544
weighted avg	0.53	0.52	0.50	544
Random Forest Classification Report:				
	precision	recall	f1-score	support
0	0.50	0.55	0.52	268
1	0.51	0.46	0.48	276
accuracy			0.50	544
macro avg	0.50	0.50	0.50	544
weighted avg	0.50	0.50	0.50	544

# Visualizations (Part I)

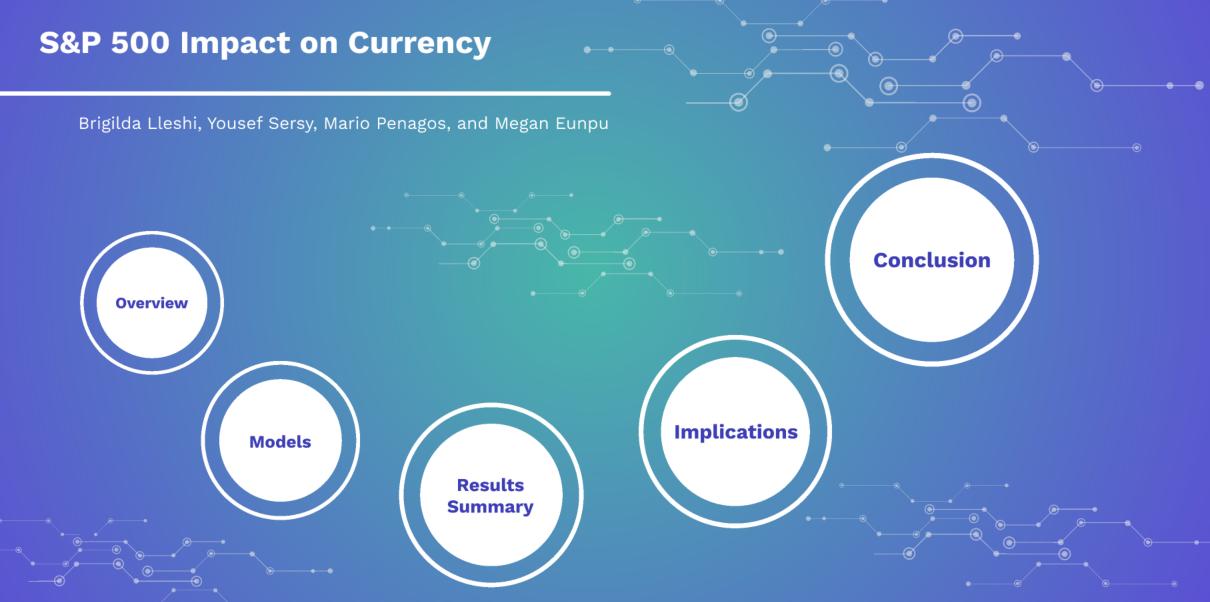




# Visualizations (Part II)







## **Implications**

It is understood that the current version of this project requires additional features in order to capture the complex relationships between the SPY and exchange rates. Features, like engineered or financial indicators/market sentiment data, might show a stronger relationship with SPY movements.

Implications include:

Model Improvement

(more sophisticated models, etc.)

Strategies for Market Stress

## Model Improvements

### **Hyperparameter Tuning**

 Techniques, like Grid Search or Random Search, to find the optimal hyperparameters for each model

### **Addressing Overfitting/Underfitting**

- Regularization
- Pruning



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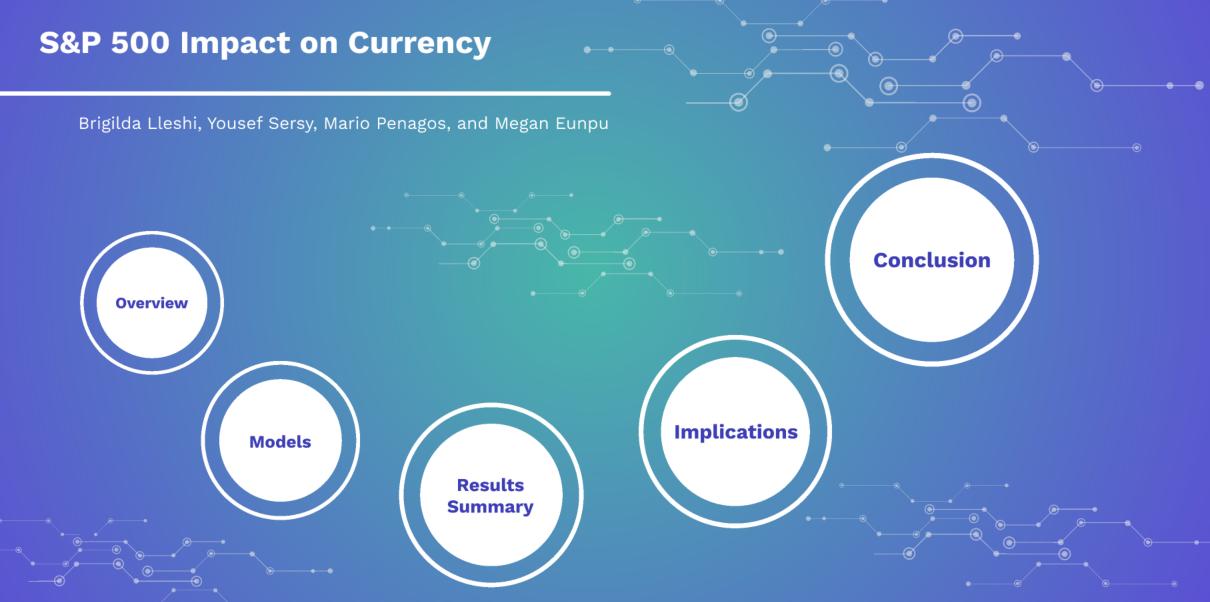
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### Conclusion

The current models show limited predictive power with an accuracy of 52%.

Future work should focus on improving feature engineering, exploring more sophisticated models, and handling periods of market stress to enhance prediction accuracy.

Questions?



**S&P 500 Impact on Currency** Brigilda Lleshi, Yousef Sersy, Mario Penagos, and Megan Eunpu Conclusion Overview **Implications** Models Results **Summary**