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## **ARIMA Time Series Forecasting - The S&P 500 Stock Market**

In this project, we perform time series forecasting on the S&P 500 Stock Market data using the Autoregressive Integrated Moving Average (ARIMA) model. The following is a step-by-step process of how we achieved the results:

### **Libraries and Data Loading**

- Import libraries such as Numpy, Pandas, Matplotlib, and Statsmodels
- Load the time series data of the S&P 500 Stock Market

### **Data Visualization and Stationarity Tests**

- Check the time series visually to ensure it is stationary
- Perform an Augmented Dickey-Fuller (ADF) test to formally test for stationarity
- If the time series is not stationary, apply the Differencing method to make it stationary

### **Determining 'p' and 'q' values**

- Use the Autocorrelation Function (ACF) and Partial Autocorrelation Function (PACF) plots to determine the values of 'p' (number of autoregressive terms) and 'q' (number of moving average terms)

### **Grid Search for Optimal 'p', 'd', and 'q' values**

- Use the grid search method to find the optimal values of 'p', 'd', and 'q'

### **Model Building**

- Build the ARIMA model using the optimal values of 'p', 'd', and 'q'

### **Model Evaluation**

- Fit the model on the training data
- Use the model to make predictions on the test data
- Plot the predictions against the actual values to evaluate the model's performance
- Use the root mean squared error (RMSE) to evaluate the model's performance

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## **Model Selection**

- Use the Akaike Information Criterion (AIC) and Bayesian Information Criterion (BIC) to compare different models and select the best one

## **Tools and Models used**

- Libraries: Numpy, Pandas, Matplotlib, Statsmodels, and Sklearn
- Models: Autoregressive Integrated Moving Average (ARIMA)
- Metrics: Augmented Dickey-Fuller (ADF) test, Differencing method, Autocorrelation Function (ACF), Partial Autocorrelation Function (PACF), root mean squared error (RMSE), Akaike Information Criterion (AIC), and Bayesian Information Criterion (BIC).