AAPL R Code

```
# Load necessary libraries
library(forecast)
library(tidyverse)
library(readr)
library(ggplot2)
library(zoo)
library(TSA)
library(rugarch)
library(forecast)
library(PerformanceAnalytics)
library(xts)
library(quantmod)
```

AAPL Monthly Data 2016-2024

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# Load & Inspect the Data
# Load the data
aapl_monthly_data <- read.csv("~/Documents/GitHub/MA-641-Course-Project/AAPL_Monthly2016.csv")</pre>
# Convert the date column to Date type
aapl_monthly_data$Date <- as.Date(aapl_monthly_data$Date, format="%Y-%m-%d")</pre>
# Inspect the data
head(aapl_monthly_data)
summary(aapl_monthly_data)
# Create a time series object
aaplmonthly_ts <- ts(aapl_monthly_data$Close, start=c(2016, 01), end = c(2024, 05), frequency=12)
# Descriptive Analysis
plot(aaplmonthly_ts, main="Monthly Apple Stock Prices", ylab="Close Price", xlab="Time")
summary(aaplmonthly ts)
# ACF and PACF Plots
par(mar=c(5, 5, 4, 2) + 0.1)
acf(aaplmonthly_ts, main="ACF of Monthly Apple Stock Prices", lag.max = 72)
pacf(aaplmonthly_ts, main="PACF of Monthly Apple Stock Prices", lag.max = 72)
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eacf(aaplmonthly_ts)
# Augmented Dickey-Fuller Test
adf_test <- adf.test(aaplmonthly_ts, alternative="stationary")</pre>
print(adf_test)
# Differencing the series if it is not stationary
if (adf_test$p.value > 0.05) {
 ts_data_diff <- diff(aaplmonthly_ts, differences=1)</pre>
  adf_test_diff <- adf.test(ts_data_diff, alternative="stationary")</pre>
 print(adf_test_diff)
  # Update the time series data to the differenced series
  aaplmonthly_ts <- ts_data_diff</pre>
}
# Time Series Plot after Differencing
plot(aaplmonthly_ts, main="Monthly Apple Stock Prices", ylab="Close Price", xlab="Time")
# ACF and PACF Plots
par(mar=c(5, 5, 4, 2) + 0.1)
acf(aaplmonthly_ts, main="ACF of Monthly Apple Stock Prices", lag.max = 72)
pacf(aaplmonthly_ts, main="PACF of Monthly Apple Stock Prices", lag.max = 72)
eacf(aaplmonthly_ts)
# ARIMA Models
# Fit AR model
ar_model <- Arima(aaplmonthly_ts, order=c(2,0,0))</pre>
# Fit MA model
ma_model <- Arima(aaplmonthly_ts, order=c(0,0,2))</pre>
# Fit ARMA(1,1,1) model
arma_model1 <- Arima(aaplmonthly_ts, order=c(1,1,1))</pre>
# ARMA(2,1,1) Model
arma_model2 <- Arima(aaplmonthly_ts, order=c(2,1,1))</pre>
# ARMA(2,1,2) Model
arma_model3 <- Arima(aaplmonthly_ts, order=c(2,1,2))</pre>
# Comparing Models using AIC and BIC
models <- list(ar_model, ma_model, arma_model1, arma_model2, arma_model3)</pre>
model_names <- c("AR", "MA", "ARIMA(1,1,1)", "ARIMA(2,1,1)", "ARIMA(2,1,2)")
aic_values <- c(sapply(models, AIC))</pre>
bic_values <- c(sapply(models, BIC))</pre>
comparison <- data.frame(Model=model_names, AIC=aic_values, BIC=bic_values)</pre>
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print(comparison)
# Perform diagnostics for ARIMA(2,1,1)
par(mar=c(5, 5, 4, 2) + 0.1)
tsdiag(arma_model2, gof.lag = 10, main = "Diagnostics for ARIMA(2,1,1)")
checkresiduals(arma_model2)
# Q-Q plot for ARIMA(2,1,1)
residuals_arma211 <- residuals(arma_model2)</pre>
qqnorm(residuals_arma211, main = "Q-Q Plot of Residuals for ARIMA(2,1,1)")
qqline(residuals_arma211, col = "red")
# Forecasting with the ARIMA(2,1,1) model
arma2_forecast <- forecast(arma_model2, h=12)</pre>
plot(arma2_forecast, main="Forecasts from ARIMA(2,1,1)")
# GARCH Models
# Create a time series object
aaplmonthly ts2 <- ts(aapl monthly data$Close, start=c(2016, 01), end = c(2024, 05), frequency=12)
# Calculate returns for modeling
returns <- diff(log(aaplmonthly_ts2))</pre>
returns <- returns[!is.na(returns)]</pre>
plot(returns, main="Monthly Apple Stock Prices", ylab="Close Price", xlab="Time", type="1")
# Specify ARMA(1,0)-GARCH(1,1) model
spec_garch <- ugarchspec(variance.model = list(model = "sGARCH", garchOrder = c(1, 1)),</pre>
                         mean.model = list(armaOrder = c(1, 0)),
                         distribution.model = "norm")
fit garchAR1 <- ugarchfit(spec = spec garch, data = returns)</pre>
# Specify ARMA(2,0)-GARCH(1,1) model
spec_garch <- ugarchspec(variance.model = list(model = "sGARCH", garchOrder = c(1, 1)),</pre>
                          mean.model = list(armaOrder = c(2, 0)),
                         distribution.model = "norm")
fit_garchAR2 <- ugarchfit(spec = spec_garch, data = returns)</pre>
# Specify ARMA(0,2)-GARCH(1,1) model
spec_garch <- ugarchspec(variance.model = list(model = "sGARCH", garchOrder = c(1, 1)),</pre>
                         mean.model = list(armaOrder = c(0, 2)),
                         distribution.model = "norm")
fit_garchMA2 <- ugarchfit(spec = spec_garch, data = returns)</pre>
# Specify ARMA(2,1)-GARCH(1,1) model
spec garch <- ugarchspec(variance.model = list(model = "sGARCH", garchOrder = c(1, 1)),</pre>
                         mean.model = list(armaOrder = c(2, 1)),
                          distribution.model = "norm")
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fit_garchARMA21 <- ugarchfit(spec = spec_garch, data = returns)</pre>
# Specify ARMA(2,2)-GARCH(1,1) model
spec_garch <- ugarchspec(variance.model = list(model = "sGARCH", garchOrder = c(1, 1)),</pre>
                                                                  mean.model = list(armaOrder = c(2, 2)),
                                                                  distribution.model = "norm")
fit_garchARMA22 <- ugarchfit(spec = spec_garch, data = returns)</pre>
# Comparing Models using AIC and BIC
model_names \leftarrow c("ARFIMA(1,0)-GARCH(1,1)", "ARFIMA(2,0)-GARCH(1,1)", "ARFIMA(0,2)-GARCH(1,1)", "ARFIMA(1,0)-GARCH(1,1)", "ARFIMA(1,0)-GARCH(1,0)-GARCH(1,0)-GARCH(1,0)-GARCH(1,0)-GARCH(1,0)-GARCH(1,0)-GARCH(1,0)-GARCH(1,0)-GARCH(1,0)-GARCH(1,0)-GARCH(1,0)-GARCH(1,0)-GARCH(1,0)-
aic_values \leftarrow c(-2.0775, -2.0822, -2.0873, -2.0705, -2.0843)
bic_values <- c(-1.9473, -1.9259, -1.9310, -1.8882, -1.8759)
comparison <- data.frame(Model=model_names, AIC=aic_values, BIC=bic_values)</pre>
print(comparison)
# GARCH Model Assumptions Check
# Extract standardized residuals
std_residuals <- residuals(fit_garchMA2, standardize = TRUE)</pre>
# Remove time index, if present
std_residuals <- as.numeric(std_residuals)</pre>
checkresiduals(std_residuals)
# Plot diagnostics
plot(fit_garchMA2, which = 1)
plot(fit_garchMA2, which = 2)
plot(fit_garchMA2, which = 3)
plot(fit_garchMA2, which = 4)
plot(fit_garchMA2, which = 5)
plot(fit_garchMA2, which = 6)
plot(fit_garchMA2, which = 7)
plot(fit_garchMA2, which = 8)
plot(fit_garchMA2, which = 9)
plot(fit_garchMA2, which = 10)
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plot(fit_garchMA2, which = 11)

plot(fit_garchMA2, which = 12)

# Forecasting with the GARCH model
forecast_garch <- ugarchforecast(fit_garchMA2, n.ahead=12)
plot(forecast_garch, which=1) # Forecast series</pre>
```