ARCH Models on S&P 500 Stock Returns

TIØ4317 Project

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Problem statement

Which forecasting model provides the most accurate out-of-sample volatility forecasts for stocks in the S&P 500, and does the best model vary across different market conditions?

Week	Date	Weekday	Time	Room	Teacher	Topic		
Part I: Emp	irical finance							
2	10.01.2025	Friday	13:15-16:00	GL-RFB R9	Morten	Introduction		
3	17.01.2025	Friday	13:15-16:00	GL-RFB R9	Maria	Statistical inference Descriptive statistics Cross-sectional data		
4	24.01.2025	Friday	13:15-16:00	GL-RFB R9	Maria	Linear regression		
5	31.01.2025	Friday	13:15-16:00	GL-RFB R9	Morten	Post estimation diagnostic tests		
6	07.02.2025	Friday	13:15-16:00	GL-RFB R9	Morten	Introduction to time series data and models		
7	14.02.2025	Friday	13:15-16:00	GL-RFB R9	Morten	Univariate time series models Forecasting Guest lecture: Hafslun		
8	21.02.2025	Friday	13:15-16:00	GL-RFB R9	Morten	Volatility models		
9	27.02.2025	Thursday	14:15-16:00	GL-RFB R10	#N/A	Guest lecture: NBIM		
9	28.02.2025	Friday	13:15-16:00	GL-RFB R9	Morten	Multivariate time series models		
10	07.03.2025	Friday	13:15-16:00	GL-RFB R9	#N/A	Winter holiday		
11	14.03.2025	Friday	13:15-16:00	GL-RFB R9	Maria	Panel data Q&A Part 1		
Dart II: Eina	ancial optimization	and rick ma	aggament					
12	21.03.2025	Friday	13:15-16:00	GL-RFB R9	Stein-Erik	Risk Measures I		
13	28.03.2025	Friday	13:15-16:00	GL-RFB R9	Stein-Erik	Risk Measures II		
14	04.04.2025	Friday	13:15-16:00	GL-RFB R9	Stein-Erik	Portfolio optimization		
15	11.04.2025	Friday	13:15-16:00	GL-RFB R9	#N/A	Ind-øk trip break		
16	18.04.2025	Friday	13:15-16:00	GL-RFB R9	#N/A	Easter break		
Student pro	oject presentation	s						
19	08.05.2025	Thursday	08:00-19:00	F5	Morten	Presentations and peer-review		
	09.05.2025	Friday	08:00-15:00	F5	Morten	Presentations and peer-review		

	Week	Date	Weekday	Time	Room	Teacher	Topic
i	8	21.02.2025	Friday	13:15-16:00	GL-RFB R9	Morten	Volatility models

Using theory from "Volatility models" chapter in the curriculum



Data

XOM Oil & Gas

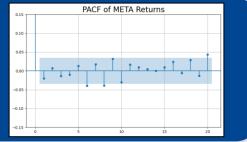
$$\sigma^2 = 1.61$$



META

Technology

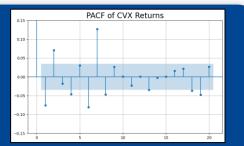
 σ^2 2.5



CVX

Oil & Gas

$$\sigma^2 = 1.73$$



TSLA

Technology

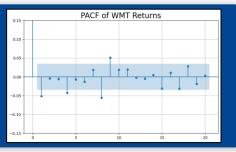
$$\sigma^2 = 3.58$$



WMT

Retail

$$\sigma^2 = 1.25$$



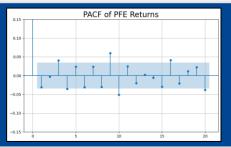
Selected 10 S&P 500 companies with different ACF and different volatility distributions



Data

PFEPharmaceuticals

$$\sigma^2 = 1.37$$



JNJ

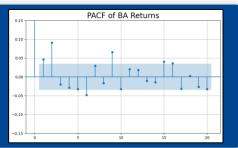
Pharmaceuticals

$$\sigma^2 = 1.08$$



BAIndustrials

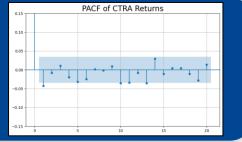
$$\sigma^2 = 2.33$$



CTRA

Industrials

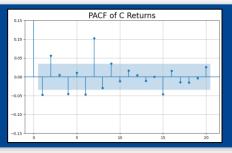
$$\sigma^2 = 2.26$$



C

Finance

$$\sigma^2 = 1.98$$



ARCH Models

ARCH (1)

- Models volatility as a function of past squared returns
- Captures volatility clustering

GARCH (1,1)

- Extends ARCH by adding lagged conditional variance terms
- Captures volatility persistence

GJR-GARCH (1,1)

- Incorporates
 asymmetry for
 negative returns
- Captures leverage effects



Approach

Select 5 S&P stocks

Fit ARCH models on absolute log returns

Evaluate fitness across models

Evaluate fitness across volatility levels

Cross-evaluate models across volatility levels

Models to be fitted

GARCH(1,1)

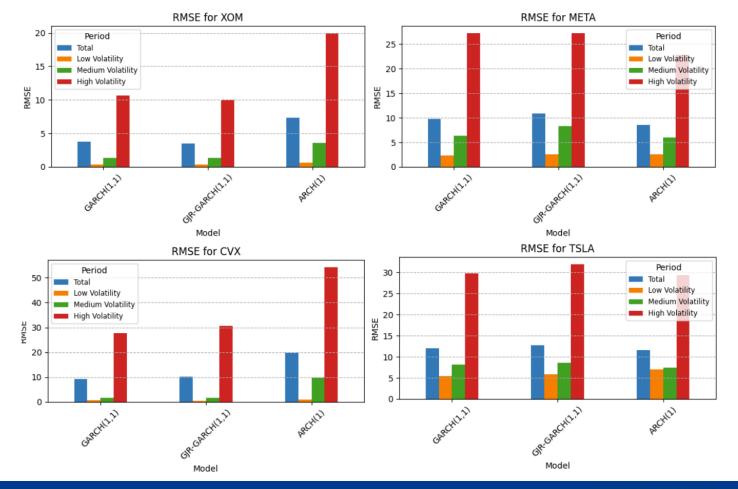
GJR-GARCH(1,1,1)

ARCH(1)

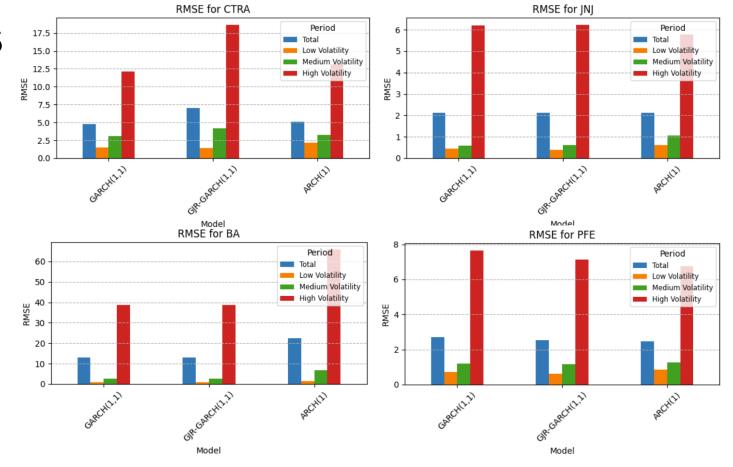
The models are fitted on a rolling window of size 21 days.
When forecasting we use the volatilities in the window to predict the next-day volatility



Results

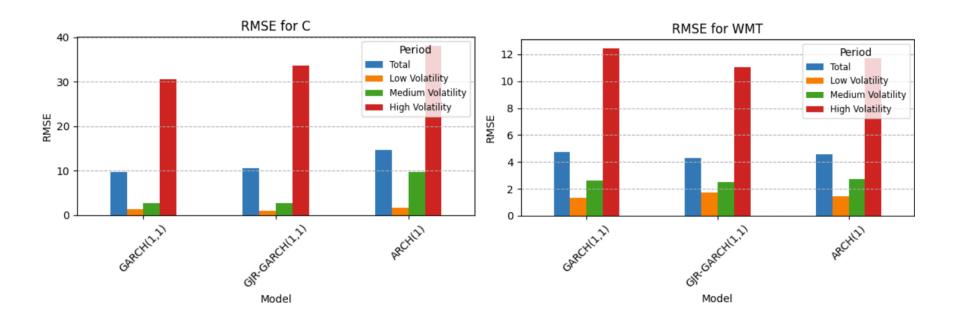


Results





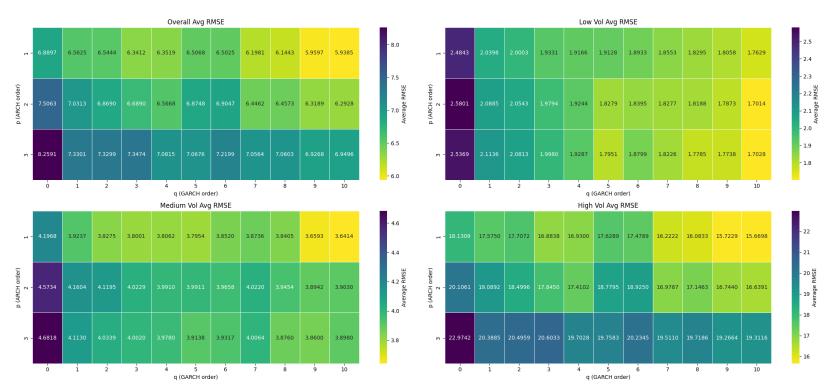
Results



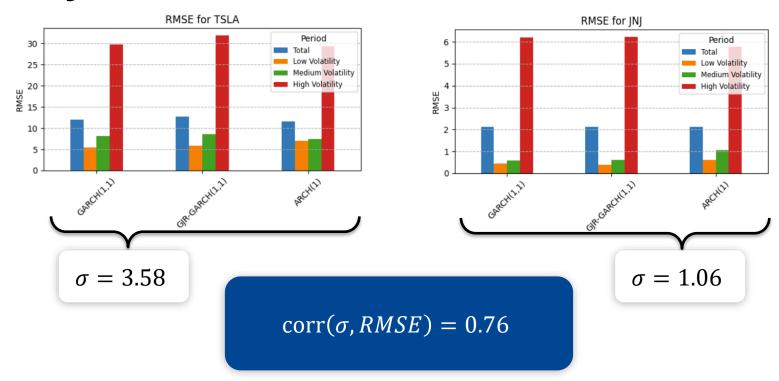


Results – GARCH parameters

Average RMSE for GARCH(p, q) Combinations Across All Stocks



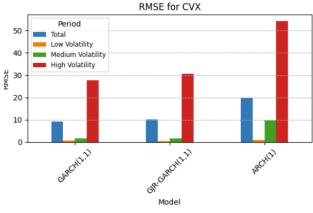
Analysis – RMSE value

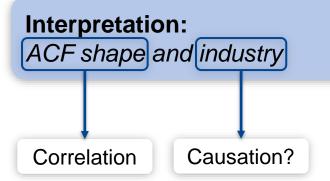


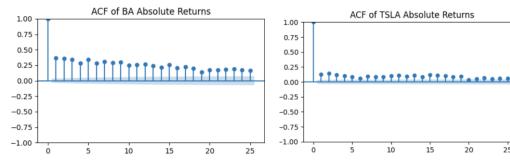
Analysis – ARCH performance

4 stocks have this property:

XOM, CVX, C, BA







20

25

Analysis – GARCH parameters

Insight:

Shocks do not persist over multiple lags, volatility does

