



Time Series Statistical Models

FinTech
Lesson 10.2



Class objectives

By the end of today's class you will understand:



Stationary vs Non-stationary data



Augmented Dickey-Fuller Test



Autoregressive Moving Average Model (ARMA)



AutoRegressive Integrated Moving Average (ARIMA)

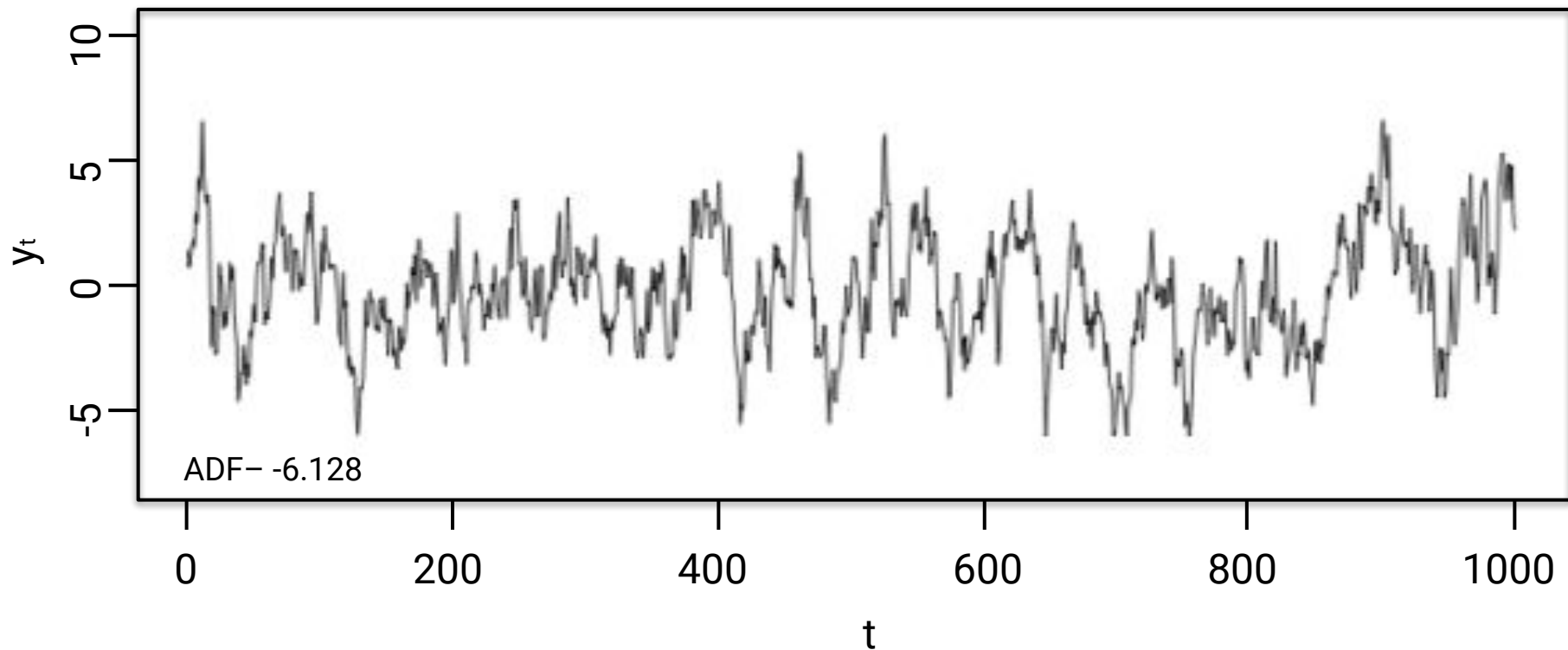


Generalized Autoregressive Conditional Heteroskedasticity (GARCH)

Stationarity

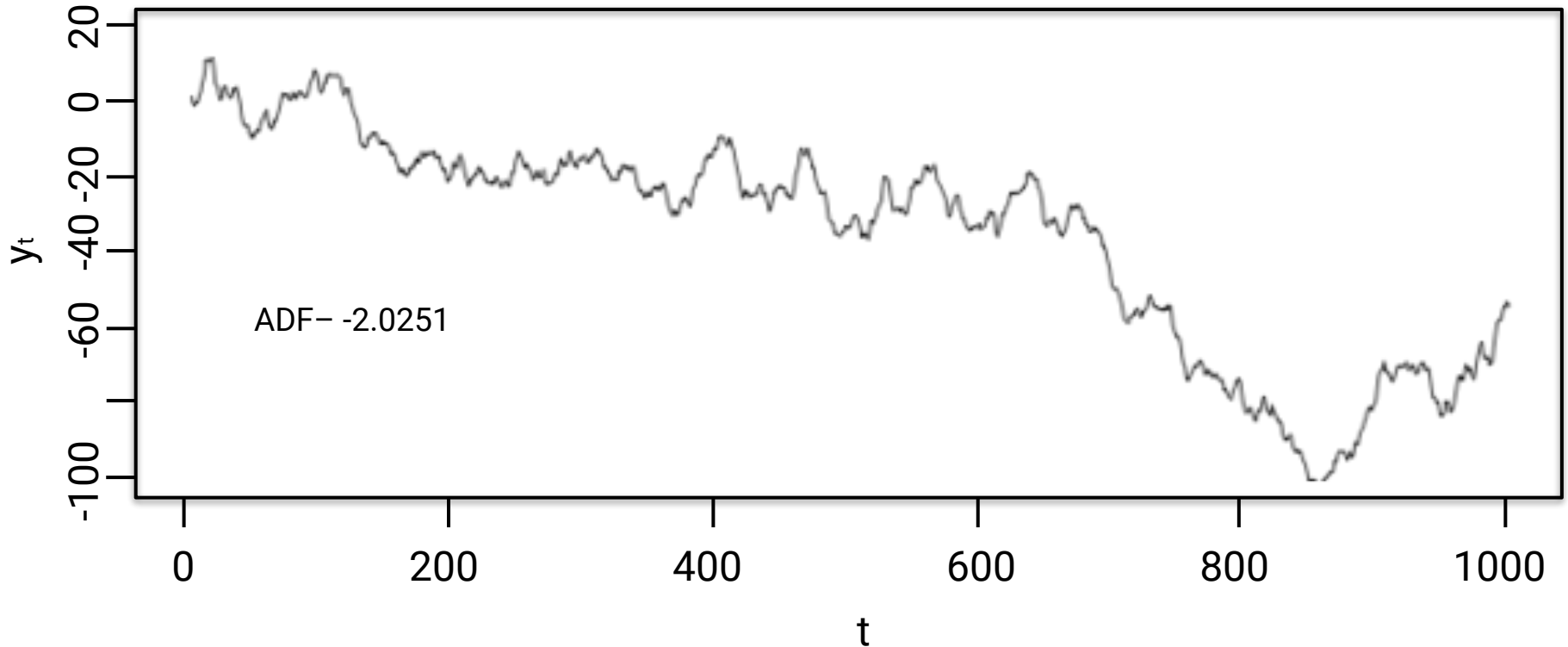
Stationarity

In a stationary process, the mean and variance are constant across time.



Non-stationary

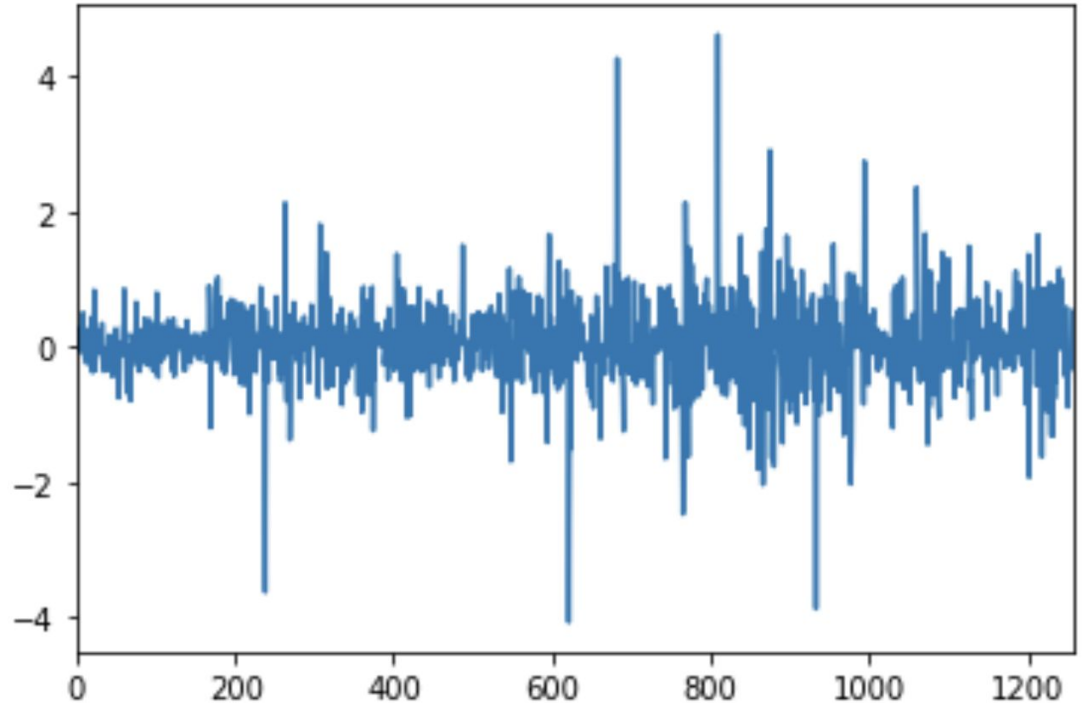
A time series with an upward or downward trend is **not stationary**.



Stationarity

Important in selecting a time series model.
Makes data easier to model.

There are strategies to transform a non-stationary time series into a stationary one.





ARMA

Auto-Regressive Model

$$y_t = \mu + a_1 y_{t-1} + \epsilon_t$$

Auto-Regressive (AR) Models

01

Past values are used to predict future values.

02

Therefore assumes some degree of autocorrelation.

03

An AR model may have one significant lag, or it may have multiple.

Second-order AR model

$$y_t = \mu + a_1 y_{t-1} + a_2 y_{t-2} + \epsilon_t$$

AR Model Summary

An AR model predicts future values based on:

01

Past values at a specified lag.

02

The number of significant lags.

Moving Average Model

$$y_t = m\epsilon_{t-1} + \epsilon_t$$



Past **errors** (plus current error) are used to predict future values.



ARIMA

ARIMA Model

$$y_t = \mu + a_1 y_{t-1} + a_2 y_{t-2} + m \epsilon_{t-1} + \epsilon_t$$



Combines features of AR and MA models.



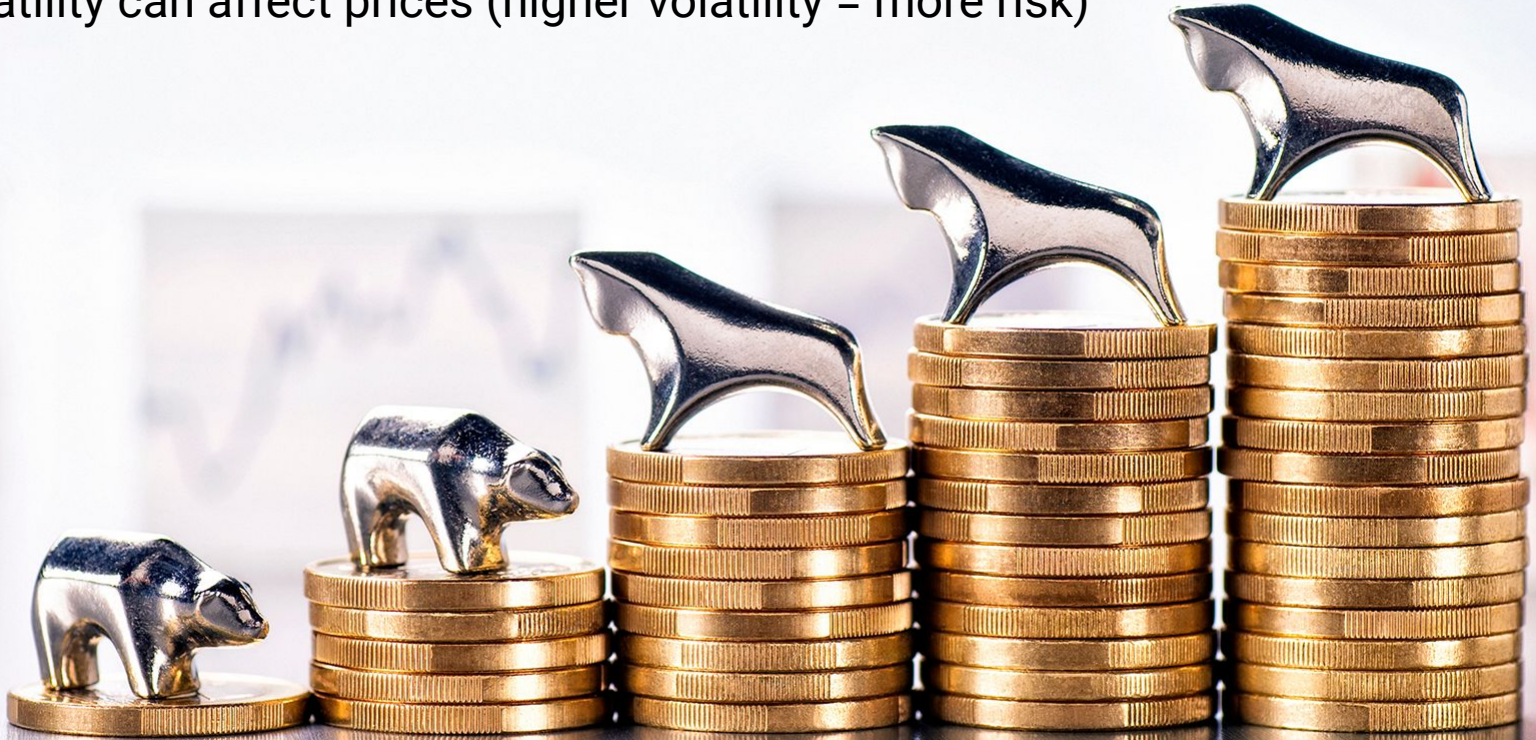
Past values and errors are used to predict future values.



Why is Volatility Important to Understand?

Higher volatility = More Risk

High volatility can affect prices (higher volatility = more risk)



Diversified Portfolio


By understanding volatility of individual assets (stocks, bonds, etc), a more diversified portfolio can be constructed



Derivatives

Some assets are particularly sensitive to volatility, e.g. derivatives.



The background of the image is a blurred financial chart. It features a grid with various data points, including percentages like +2.11%, -1.11%, +7.14%, and -3.12% on the left, and numerical values like -4.28, +13.28, -11.28, +17.28, -2.28, +13.28, -11.28, and +17.28 on the right. A hand is visible in the center, holding a pen and pointing at a smartphone screen. The overall theme is financial volatility and data analysis.

Volatility
can beget
volatility,
i.e. cluster.

AIC & BIC



Akaike Information Criterion, Bayesian Information Criterion.



Assess how well a model fits the data (goodness of fit), and complexity.



Higher-order models are penalized for complexity.



Lower scores are better.



GARCH

ARMA

Auto-Regressive component:

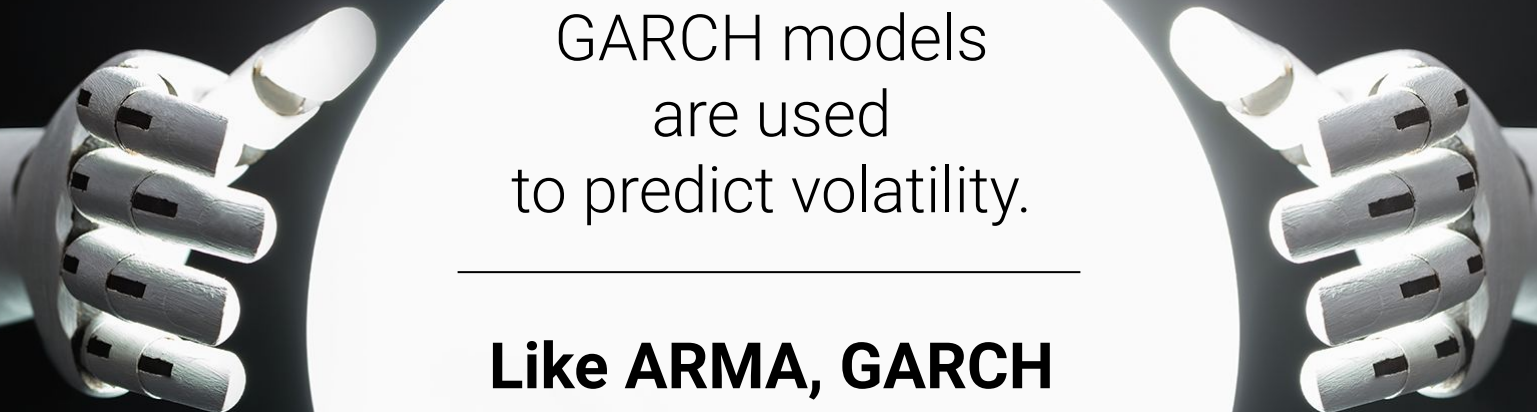
Future values predicted
based on **past values**.

Moving Average component:

Future values predicted based
on **past errors**.



GARCH



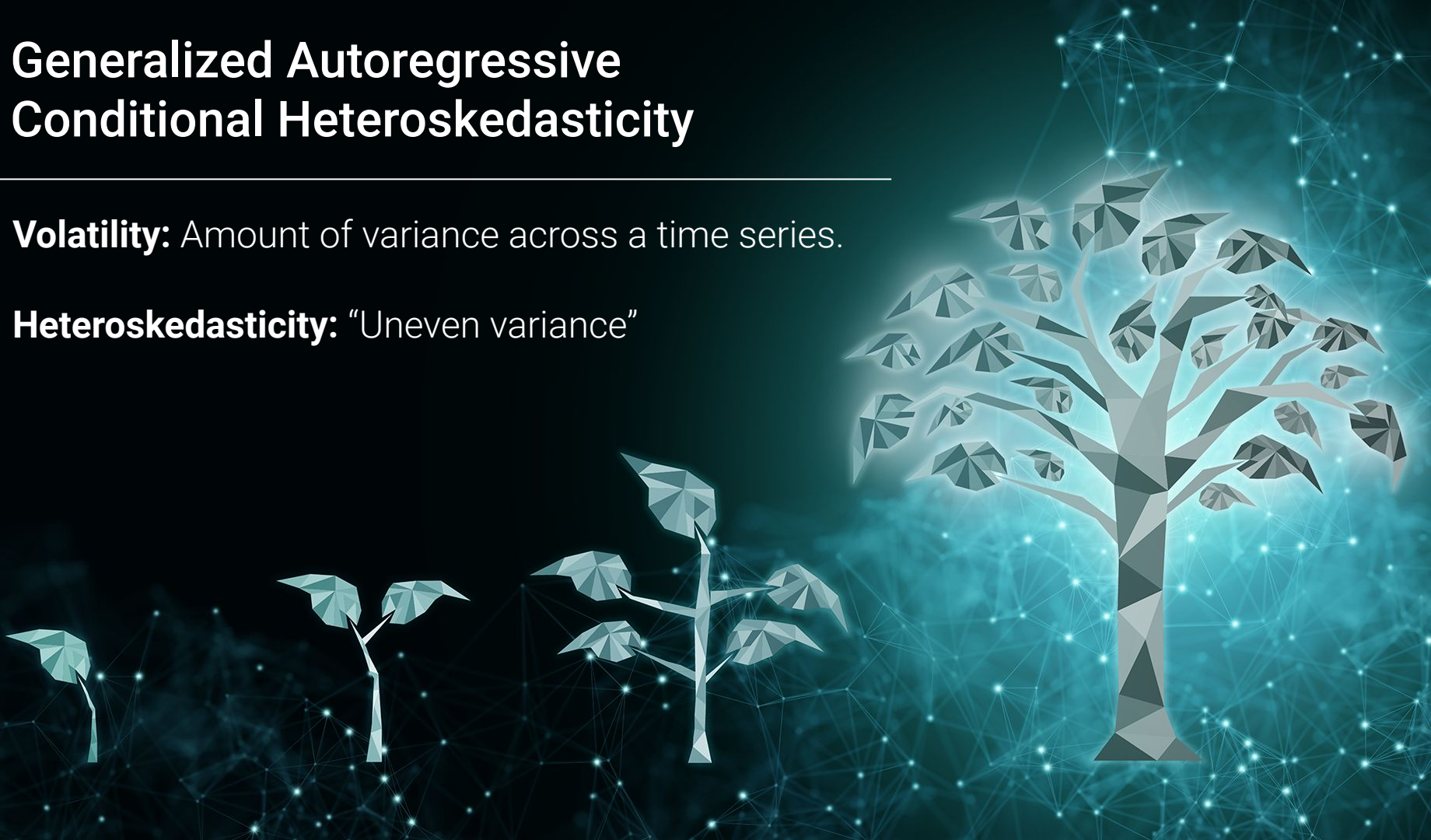
GARCH models
are used
to predict volatility.

**Like ARMA, GARCH
also has auto-regressive
and moving average
components.**

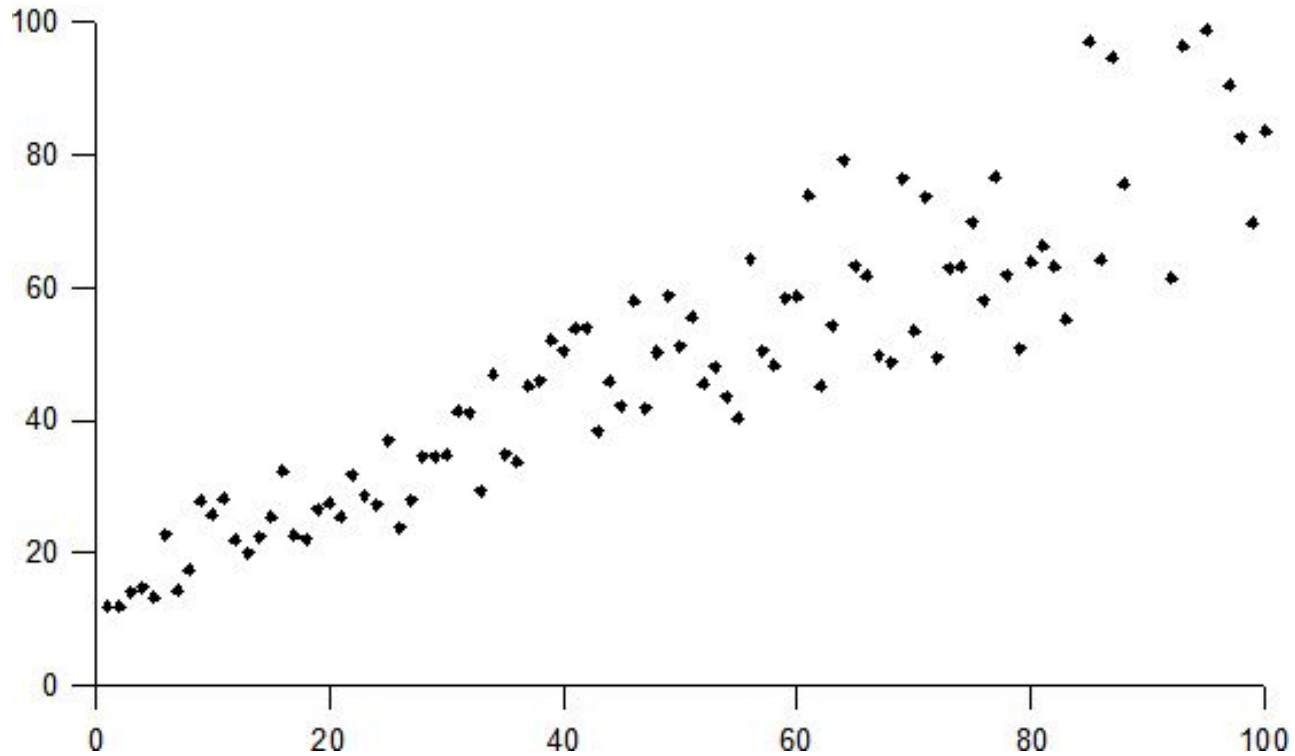
Generalized Autoregressive Conditional Heteroskedasticity

Volatility: Amount of variance across a time series.

Heteroskedasticity: “Uneven variance”



Heteroskedasticity



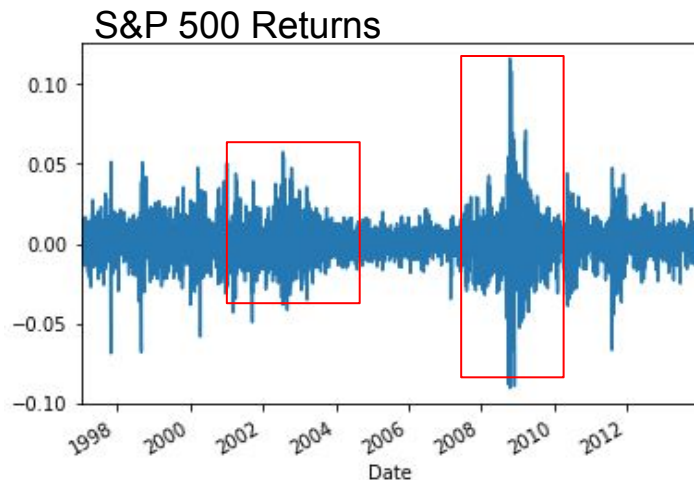
Volatile Periods in the US Stock Market



Volatility and returns tend to cluster.



GARCH is a model designed to take specific advantage of that.





Questions?