

## GARCH(r,s)

- is used to reflect the fact that periods with high volatility are followed by periods of low volatility.

- Assumption: (log) returns are W.N.

$$\begin{cases} r_t = \mu + \varepsilon_t \\ \varepsilon_t \sim N(0, \sigma_t^2) \end{cases} \rightarrow \text{depends on } t$$

- Equation:

$$\begin{cases} \sigma_t^2 = a_0 + \sum_{j=1}^s a_j \varepsilon_{t-j}^2 + \sum_{j=1}^r b_j \sigma_{t-j}^2 \\ \sum_{j=1}^s a_j + \sum_{j=1}^r b_j < 1, a_j, b_j > 0 \end{cases}$$

previous errors      prev. variance

## IGARCH(r,s)

- GARCH(r,s) with  $\sum_{j=1}^s a_j + \sum_{j=1}^r b_j = 1$

## ARCH(s) = GARCH(0,s)

- Current variance is explained only by previous errors

## t-GARCH

- Used to better model fat tails

- Assumption:

$$\begin{cases} r_t = \mu + \varepsilon_t \\ \varepsilon_t \sim t(df > 2) \end{cases}$$

instead of  $N(\mu, \sigma^2)$

## ARCH-in-mean [ARCH-M]

- Investors are risk averse  $\Rightarrow$  they want higher average return in uncertain periods: DIFFERENT model for returns assumed

$$\begin{cases} r_t = \mu_0 + \mu_1 \sigma_t^2 + \varepsilon_t \\ \mu_1 > 0 \\ \varepsilon_t \sim N(0, \sigma_t^2) \end{cases} \rightarrow \text{capture higher risk}$$

## Asymmetric GARCH [threshold GARCH, GJR model]

- Good and Bad news effects differently

Equation (for GARCH(1,1))

$$\begin{cases} \sigma_t^2 = a_0 + a_1 \varepsilon_{t-1}^2 + b_1 \sigma_{t-1}^2 + j D_{t-1} \varepsilon_{t-1}^2 \\ D - \text{dummy, positive shock} \end{cases}$$

## Exponential GARCH [EGARCH]

- Guaranties positive variance:  $\log(\sigma_t^2)$  is modeled:

Equation:

$$\begin{cases} \log(\sigma_t^2) = a_0 + a_1 \frac{|\varepsilon_{t-1}|}{\sigma_{t-1}} + b_1 \log(\sigma_{t-1}^2) + j \frac{\varepsilon_{t-1}}{\sigma_{t-1}^2} \\ \text{if } j < 0, \text{ positive shocks generate less volatility than negative} \end{cases}$$

## HOW IT WORKS!

use e.g. ARIMA

- From original ts.  $\rightarrow$  stationary with errors = W.N.
- We still can have big errors in prediction because of non-constant volatility  $\rightarrow$  use GARCH models  $\rightarrow$  now we can predict not only "trend" but also capture all volatility patterns, making the whole model more precise