

# Advanced Risk Management

## Computer Lab 2

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# Assignment 1 (Lab)

# Question 1

# Tips

- Install and import relevant libraries
- Import data: Excel file “raw\_data\_2021”.
  - ▷ Remove irrelevant parts (seconds) from “Date” string: `str[]`
  - ▷ Transform “Date” string to format datetime: `to_datetime()`
  - ▷ Set index of Pandas DataFrame to “Date” column, allows to refer to a string using time index: `set_index()`
  - ▷ Calculate close-to-close returns (multiply by 100 to define % growth) and select for one particular index: `loc[]`
  - ▷ Select RV for one particular index: `loc[]`; multiply  $100^2$
- Define new Pandas DataFrame: first column - returns, second column - RV
  - ▷ Plot: (i) return series, (ii) autocorr of return series, (iii) autocorr of squared return series

## Question 2

## Tips (1/2)

- Model with constant mean is given by:

$$\begin{aligned}r_t &= \mu + \varepsilon_t \\ \sigma_t^2 &= \omega + \alpha \varepsilon_{t-1}^2 + \beta \sigma_{t-1}^2 \\ \varepsilon_t &= \sigma_t e_t, \quad e_t \sim N(0, 1)\end{aligned}\tag{1}$$

- ▷ stationarity condition:  $\alpha + \beta < 1$
- Use `arch_model("Returns", p = ..., o = ..., q = ...)`
- Obtain results: `fit(last_obs = , disp = "off")`
  - ▷ "last\_obs" defines the last observation of sub-period to estimate the model
  - ▷ "disp" controls whether convergence information is shown

## Tips (2/2)

- Construct standardised residuals:

$$e_t = \frac{r_t - \mu}{\sigma_t} \quad (2)$$

- Obtain non-standardised residuals ( $\varepsilon_t$ ) using method `resid()` applied to model estimation results
- Obtain conditional volatility using method `conditional_volatility()` applied to model estimation results
- Alternatively, `plot()` applied to results after fitting the model: plots  $e_t$  and  $\sigma_t$
- Conduct diagnostic tests based on standardised residuals

## Question 3



## Tips (1/2)

- GJR-GARCH(1,1):

$$\sigma_t^2 = \omega + (\alpha + \gamma \mathbf{1}_{\varepsilon_{t-1} \leq 0}) \varepsilon_{t-1}^2 + \beta \sigma_{t-1}^2 \quad (3)$$

- ▷ NIC is asymmetric iff  $\gamma \neq 0$
- ▷ present leverage effect if  $\gamma > 0$
- ▷ stationarity condition:  $\alpha + \frac{1}{2}\gamma + \beta < 1$
- `arch_model()`:  $o = 1$  (asymmetric terms)
- Conduct diagnostic tests based on standardised residuals

## Tips(2/2)

- Is the log likelihood of the GJR model significantly higher?
- Does the LR test for the GARCH model, as a restricted version of the GJR-GARCH model, reject the null hypothesis?
  - ▷  $H_0 : \gamma = 0$
  - ▷  $LR = 2(LL_{GJR} - LL_{GARCH}) \sim \chi_1^2$
  - ▷ One degree of freedom in chi-squared distribution because we test only one restriction

## HAR model (additional)

## Tips

- HAR model: heterogeneous autoregressive model for the realised variance

$$RV_{t+1} = \phi_0 + \phi_d RV_{d,t} + \phi_w RV_{w,t} + \phi_m RV_{m,t} + \varepsilon_{t+1}, \quad (4)$$

where

- ▷  $RV_{d,t}$  is daily mean
- ▷  $RV_{w,t}$  is weakly mean
- ▷  $RV_{m,t}$  is monthly mean
- Estimate rolling averages using `.rolling(window = ...).mean()` in pandas
- Estimate model using OLS: `smf.ols()`: shift regressors by one!

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