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## (G)ARCH models

- (G)ARCH models are employed b/c sometimes having a constant variance limits interpretation for some very volatile series. They drop this assumption by modelling  $\epsilon_t$  as an ARMA process.
  - ↳ If it's reasonable to adopt a GARCH model, the squared correlograms will look very AR, indicating the series has a substantial dependence in the volatility of the returns.

The process for estimating a model w/ GARCH errors:

- Estimate the mean eqn (mult. regress. or ARIMA) for  $y_t$
- Take the squared residuals,  $e_t^2$ , and estimate variance eqn

Once you specify the mean eqn you can formally test for conditional heteroskedasticity using the Lagrange Multiplier (LM) test.

$H_0$ : no ARCH effects of order 1, ...,  $g$ .  $H_a$ : some ARCH effects of same order.

Finalising this hypothesis test is a two-step process:

- Estimate  $e_t^2$  w/ OLS:  $e_t^2 = \alpha_0 + \alpha_1 e_{t-1}^2 + \dots + \alpha_g e_{t-g}^2 + \epsilon_t$ 
    - ↳ No ARCH effects of order 1 to  $g$  if insig.
  - Take  $R^2$  from regression & compute LM statistic
    - ↳  $LM = TR^2$   $\rightarrow T$  = usable sample size
    - ↳ Under  $H_0$ , LM converges to a chi-square dist. w/ df =  $g$
- $\rightarrow$  Reject  $H_0$  if LM is sufficiently large (use test to check if GARCH spec. is sufficient)

