

Mean - Vol Relationship

VAR

A reduced form VAR of some number of variables, k , and some number of lags, p , is written as,

$$y_t = A_1 y_{t-1} + \dots + A_p y_{t-p} + u_t \quad (1)$$

where A_i is a $k \times k$ matrix of coefficients and,

$$y_t = \begin{bmatrix} \text{output}_t \\ \text{unemployment}_t \\ \text{inflation}_t \\ \dots \\ \text{vfc}_t \end{bmatrix}, \quad u_t = \begin{bmatrix} u_t^{\text{output}} \\ u_t^{\text{unemployment}} \\ u_t^{\text{inflation}} \\ \dots \\ u_t^{\text{vfc}} \end{bmatrix} \quad (2)$$

Each u_t are the reduced form residuals for the accompanying data series.

With this, we can estimate the predicted values, \hat{y}_t , and the residuals, \hat{u}_t .

Volatility

We can then define the log-variance as

$$\text{Var}_t \equiv \log(\hat{u}_t^2) \quad (3)$$

This can be estimated with the regression,

$$\text{Var}_t = \alpha_1 y_{t-1} + \dots + \alpha_p y_{t-p} + \epsilon_t \quad (4)$$

Then the VFCI is defined as,

$$\text{VFCI}_t \equiv \widehat{\text{Var}_t} \quad (5)$$

The VFCI series is then rescaled to $N(0,1)$.

External VFCI

We will compare to the externally estimated VFCI on forward GDP growth (forwarded 1 quarter) and financial principal components.

variable — External VFCI — VFCI from VAR (one for each variable)



