

Diagnostics for Constant Parameters

Recall, in the CER model

$$R_t \sim \text{iid } N(\mu, \sigma^2), \quad t = 1, \dots, T$$

μ is constant over time

σ^2 is constant over time

Q: Is the assumption of constant means and variances consistent with the observed data?

Q: How can you determine if μ and σ^2 are constant over time?

Rolling Means

Idea: compute estimate of μ over rolling windows of length $n < T$

$$\begin{aligned}\hat{\mu}_t(n) &= \frac{1}{n} \sum_{i=0}^{n-1} R_{t-i} \\ &= \frac{1}{n} (R_t + R_{t-1} + \cdots + R_{t-n+1}) \\ t &= n, n+1, \dots, T\end{aligned}$$

Plot $\hat{\mu}_t(n)$ over time. If $\hat{\mu}_n(n) \approx \hat{\mu}_{n+1}(n) \approx \cdots \hat{\mu}_T(n)$ then the data support μ constant

Note: The S-PLUS function

`aggregateSeries`

can be used to easily compute rolling means.

Rolling Variances and Standard Deviations

Idea: Compute estimates of σ^2 and σ over rolling windows of length $n < T$

$$\hat{\sigma}_t^2(n) = \frac{1}{n-1} \sum_{i=0}^{n-1} (R_{t-i} - \hat{\mu}_t(n))^2$$

$$\hat{\sigma}_t(n) = \sqrt{\hat{\sigma}_t^2(n)}$$

$$t = n, n+1, \dots, T$$

Plot $\hat{\sigma}_t(n)$ over time. If $\hat{\sigma}_n(n) \approx \hat{\sigma}_{n+1}(n) \approx \dots \approx \hat{\sigma}_T(n)$ then the data support σ constant

Rolling Covariances and Correlations

Idea: Compute estimates of σ_{jk} and ρ_{jk} over rolling windows of length $n < T$

$$\hat{\sigma}_{jk,t}(n) = \frac{1}{n-1} \sum_{i=0}^{n-1} (R_{jt-i} - \hat{\mu}_j(n))(R_{kt-i} - \hat{\mu}_k(n))$$
$$\hat{\rho}_{jk,t}(n) = \frac{\hat{\sigma}_{jk,t}(n)}{\hat{\sigma}_{jt}(n)\hat{\sigma}_{kt}(n)}$$
$$t = n, n+1, \dots, T$$

Plot $\hat{\rho}_{jk,t}(n)$ over time. If $\hat{\rho}_{jk,n}(n) \approx \hat{\rho}_{jk,n+1}(n) \approx \dots \approx \hat{\rho}_{jk,T}(n)$ then the data support ρ_{ij} constant

Rolling Regression

Recall the Single Index Model

$$R_{it} = \alpha_i + \beta_i R_{Mt} + \varepsilon_{it}$$

α_i is constant over time

β_i is constant over time

Idea: Compute least squares estimates of α_i and β_i from SI model over rolling windows of length $n < T$

$$R_{it}(n) = \alpha_i(n) + \beta_i(n)R_{Mt}(n) + \varepsilon_{it}(n)$$

$$\hat{\beta}_{in}(n) = \frac{\hat{\sigma}_{jM,t}(n)}{\hat{\sigma}_{Mt}^2(n)},$$

$$\hat{\alpha}_{in}(n) = \hat{\mu}_{in}(n) - \hat{\beta}_{in}(n)\hat{\mu}_{Mn}(n)$$

$$t = n, n + 1, \dots, T$$

Plot $\hat{\beta}_{in}(n)$ over time. If $\hat{\beta}_{in}(n) \approx \hat{\beta}_{in+1}(n) \approx \dots \approx \hat{\beta}_{iT}(n)$ then the data support β constant

Rolling Efficient Portfolios

Idea: Using rolling estimates of μ and Σ compute rolling efficient portfolios

- global minimum variance portfolio
- tangency portfolio
- efficient frontier