

$$3.2) \quad Y_t = \beta_0 + \beta_1 t + X_t$$

$$q(\hat{\beta}_0, \hat{\beta}_1) = \sum_{t=1}^n [Y_t - (\hat{\beta}_0 + \hat{\beta}_1 t)]^2$$

Goal:

$$\underset{q(\hat{\beta}_0, \hat{\beta}_1)}{\text{minimize}} = \sum_{t=1}^n (Y_t - \hat{\beta}_0 - \hat{\beta}_1 t)^2$$

$$\frac{\partial q(\hat{\beta}_0, \hat{\beta}_1)}{\partial \hat{\beta}_0} = -2 \sum_{t=1}^n (Y_t - \hat{\beta}_0 - \hat{\beta}_1 t)$$

$$\frac{\partial q(\hat{\beta}_0, \hat{\beta}_1)}{\partial \hat{\beta}_1} = -2 \sum_{t=1}^n t(Y_t - \hat{\beta}_0 - \hat{\beta}_1 t)$$

$$\text{set } \frac{\partial q(\hat{\beta}_0, \hat{\beta}_1)}{\partial \hat{\beta}_0}, \frac{\partial q(\hat{\beta}_0, \hat{\beta}_1)}{\partial \hat{\beta}_1} = 0$$

$$\frac{\partial q(\hat{\beta}_0, \hat{\beta}_1)}{\partial \hat{\beta}_0} = -2 \sum_{t=1}^n (Y_t - \hat{\beta}_0 - \hat{\beta}_1 t) = 0$$

$$= \sum_{t=1}^n (Y_t - \hat{\beta}_0 - \hat{\beta}_1 t) = 0 \quad \bar{X} = \frac{1}{n} \sum_{t=1}^n X_t$$

$$= \sum_{t=1}^n Y_t - \sum_{t=1}^n \hat{\beta}_0 - \sum_{t=1}^n \hat{\beta}_1 t = 0 \quad \sum_{t=1}^n X_t = n \bar{X}$$

$$= n\bar{Y} - n\hat{\beta}_0 - n\hat{\beta}_1 \bar{t} = 0$$

$$= \bar{Y} - \hat{\beta}_0 - \hat{\beta}_1 \bar{t} = 0$$

$$\Rightarrow \bar{Y} = \hat{\beta}_0 + \hat{\beta}_1 \bar{t} \rightarrow \boxed{\hat{\beta}_0 = \bar{Y} - \hat{\beta}_1 \bar{t}}$$

$$\frac{\partial q(\hat{\beta}_0, \hat{\beta}_1)}{\partial \hat{\beta}_1} = -2 \sum_{t=1}^n t(Y_t - \hat{\beta}_0 - \hat{\beta}_1 t) = 0$$

$$= \sum_{t=1}^n tY_t - \sum_{t=1}^n \hat{\beta}_0 t - \sum_{t=1}^n \hat{\beta}_1 t^2 = 0$$

$$= \sum_{t=1}^n tY_t - n\hat{\beta}_0 \bar{t} - \sum_{t=1}^n \hat{\beta}_1 t^2 = 0$$

$$= \sum_{t=1}^n tY_t - n(\bar{Y} - \hat{\beta}_1 \bar{t}) \bar{t} - \hat{\beta}_1 \sum_{t=1}^n t^2 = 0$$

$$= \sum_{t=1}^n tY_t - n\bar{Y} \bar{t} = \hat{\beta}_1 \sum_{t=1}^n t^2 - n\hat{\beta}_1 \bar{t}^2$$

$$\hat{\beta}_1 = \frac{\sum_{t=1}^n tY_t - n\bar{Y} \bar{t}}{\sum_{t=1}^n t^2 - n\bar{t}^2} = \boxed{\frac{\sum_{t=1}^n (Y_t - \bar{Y})(t - \bar{t})}{\sum_{t=1}^n (t - \bar{t})^2}}$$

3.2) $y_t = \mu + e_t - e_{t-1}$

Find $\text{Var}(\bar{y}_t)$ \rightarrow compare to $\text{Var}(\bar{y}_t)$ when $y_t = \mu + e_t$

First, find \bar{y}_t .

$$\begin{aligned}\bar{y}_t &= \frac{1}{n} \sum_{t=1}^n y_t = \frac{1}{n} \sum_{t=1}^n (\mu + e_t - e_{t-1}) \\ &= \mu + \frac{1}{n} \sum_{t=1}^n (e_t - e_{t-1}) \\ &= \mu + \frac{1}{n} (e_n - e_0)\end{aligned}$$

$$\begin{aligned}\text{Var}(\bar{y}_t) &= \text{Var}\left(\mu + \frac{1}{n}(e_n - e_0)\right) \\ &= \cancel{\text{Var}(\mu)} + \text{Var}\left(\frac{1}{n}(e_n - e_0)\right) \\ &= \frac{1}{n^2} (\sigma^2 + \sigma^2) \\ &= \frac{2\sigma^2}{n^2}\end{aligned}$$

When $y_t = \mu + e_t$

$$\begin{aligned}\bar{y}_t &= \frac{1}{n} \sum_{t=1}^n y_t = \mu + \frac{1}{n} \sum_{t=1}^n e_t \\ \text{Var}(\bar{y}) &= \frac{1}{n^2} (n\sigma^2) = \frac{\sigma^2}{n}\end{aligned}$$

So, $y_t = \mu + e_t - e_{t-1}$ $\left\{ \begin{array}{l} y_t = \mu + e_t \\ \text{Var}(\bar{y}_t) = \frac{2\sigma^2}{n^2} \end{array} \right. \quad \left\{ \begin{array}{l} y_t = \mu + e_t \\ \text{Var}(\bar{y}) = \frac{\sigma^2}{n} \end{array} \right.$

Primary difference is in the denominator
When $y_t = \mu + e_t - e_{t-1}$, the sample mean has a significantly larger impact on the variance of the mean.

3.16)

$\{Y_t\}$ stationary

$$P(k) = \phi^k \quad k \geq 0 \quad \phi: (-1, +1)$$

a) Show that $V(\bar{Y}) = \frac{\sigma_0}{n} \left[\frac{1+\phi}{1-\phi} - \frac{2\phi}{n} \left(\frac{1-\phi^n}{(1-\phi)^2} \right) \right]$

Hint: $\sum_{k=0}^n \phi^k = \frac{1-\phi^{n+1}}{1-\phi}$

$$\sum_{k=0}^n k \phi^{k-1} = \frac{d}{d\phi} \left(\sum_{k=0}^n \phi^k \right) = \frac{d}{d\phi} \left(\frac{1-\phi^{n+1}}{1-\phi} \right) \quad f'(\phi) = -(n+1)\phi^n \quad g'(\phi) = -1$$

quotient rule: $h'(x) = \frac{f'(x)g(x) - f(x)g'(x)}{[g(x)]^2}$

$$\Rightarrow \frac{d}{d\phi} \left(\frac{1-\phi^{n+1}}{1-\phi} \right) = \frac{(-1)(n+1)\phi^n (1-\phi) - (1-\phi^{n+1})(-1)}{(1-\phi)^2}$$

numerator: $(1-\phi)(-1)(n+1)\phi^n + (1-\phi^{n+1})(1)$
 $= (-1)(n+1)\phi^n + (1-\phi^{n+1})$
 $= (-1)(n+1)\phi^n + 1 - \phi^{n+1}$

$$\frac{d}{d\phi} \left(\frac{1-\phi^{n+1}}{1-\phi} \right) = \frac{-1(n+1)\phi^n + 1 - \phi^{n+1}}{(1-\phi)^2}$$

$$\text{Var}(\bar{Y}) = \frac{\sigma_0}{n} \left(1 + 2 \sum_{k=1}^{n-1} \left(1 - \frac{k}{n} \right) \rho_k \right)$$

$$= \frac{\sigma_0}{n} \left(1 + 2 \sum_{k=1}^{n-1} \left(1 - \frac{k}{n} \right) \phi^k \right)$$

$$= \frac{\sigma_0}{n} \left(1 + 2 \sum_{k=1}^{n-1} \phi^k - \sum_{k=1}^{n-1} \frac{k}{n} \phi^k \right)$$

$$= \frac{\sigma_0}{n} \left(1 + 2 \sum_{k=1}^{n-1} \phi^k - \frac{2}{n} \sum_{k=1}^{n-1} k \phi^k \right)$$

$$= \frac{\sigma_0}{n} \left(1 + 2 \left(\frac{1-\phi^n}{1-\phi} \right) - \frac{2}{n} \left(\frac{1 + (n-1)(\phi^n - n\phi^{n-1})}{(1-\phi)^2} \right) \right)$$

$$= \frac{\sigma_0}{n} \left[\frac{1+\phi}{1-\phi} - \frac{2\phi}{n} \left(\frac{1-\phi^n}{(1-\phi)^2} \right) \right]$$

b) for large n , $\text{Var}(\bar{Y}) \approx \frac{\sigma_0}{n} \left(\frac{1+\phi}{1-\phi} \right)$

because the $\frac{2\phi}{n} \left(\frac{1-\phi^n}{(1-\phi)^2} \right)$ term goes towards zero
 this only leaves the first term