

$$\hat{\beta} = \beta + \sum z_i \varepsilon_i$$

$$\begin{aligned} \text{Var}(\hat{\beta}) &= \text{Var}(\beta) + \text{Var}(\sum z_i \varepsilon_i) + 2\text{Cov}(\beta, \sum z_i \varepsilon_i) \\ &= 0 + \text{Var}(\sum z_i \varepsilon_i) + 0 \end{aligned}$$

$$= \text{Var}(z_1 \varepsilon_1 + z_2 \varepsilon_2 + \dots + z_n \varepsilon_n)$$

$$\sum_{i=1}^n \text{Var} z_i \varepsilon_i + 2 \sum_{i \neq j} \text{Cov}(z_i \varepsilon_i, z_j \varepsilon_j)$$

$$= \sum_{i=1}^n z_i^2 \text{Var}(\varepsilon_i) + 2 \sum_{i \neq j} z_i z_j \text{Cov}(\varepsilon_i, \varepsilon_j)$$

$$\sum_{i=1}^n z_i^2 \sigma^2 + 2 \sum_{i \neq j} 0$$

$$\cancel{\sigma^2 \sum_{i=1}^n z_i^2} + 0$$

$$= \sigma^2 \sum_{i=1}^n z_i^2 + 0 = \sigma^2 \sum_{i=1}^n z_i^2$$

where  $z_i = \frac{x_i}{\sum_{j=1}^n x_j^2}$

Note:- ~~Cov(z\_i, z\_j)~~

$\text{Cov}(\varepsilon_i, \varepsilon_j) = 0$  since  $\varepsilon_i, \varepsilon_j$  are i.i.d.s.

$\beta$  is a constant (unknown) so  $\text{Var}(\beta) = 0$ ,  $E[\beta] = \beta \Rightarrow \beta - E[\beta] = 0$   
 $\text{Cov}(\beta, \sum z_i \varepsilon_i) = E[(\beta - E[\beta]) \cdot (\sum z_i \varepsilon_i - E[\sum z_i \varepsilon_i])]$   
 $= E[0] = 0$

$$\hat{\beta} = \beta + (\cancel{\sum z_i^2}) \sum z_i \varepsilon_i$$

$$\beta = \sum z_i \varepsilon_i$$

$$= \frac{(z_1^2 + z_2^2 + \dots + z_n^2)}{10^2}$$

$$E[(\varepsilon_i - 0)(\varepsilon_j)]$$

$$E(\varepsilon_i \varepsilon_j)$$

$$\text{Cov}(\varepsilon_i, \varepsilon_j) = E[(\varepsilon_i - E[\varepsilon_i])(\varepsilon_j - E[\varepsilon_j])]$$

$$= E[\varepsilon_i \varepsilon_j] = E[\varepsilon_i] E[\varepsilon_j]$$

$$= 0 \times 0 = 0$$

since iid's