

The Solution to the regularized LS problem is given by  $\hat{\beta}_z = (X^TX + \tau II)^T X^TY$ We can write  $Y = X\beta + \varepsilon$ , giving us,  $\hat{\beta} = (X^TX + zI)^T X^T (X\beta + \varepsilon)$   $= (X^TX + zI)^T X^T (X\beta) + (X^TX + zII)^T X^T \varepsilon$  $E[\hat{\beta}_{z}|x] = E[(x^{T}X + zII)^{T}X^{T}(x\beta + c) | X]$   $= E[(x^{T}X + zII)^{T}X^{T}E[x] + E[(x^{T}X + zII)^{T}x^{T}x\beta]$   $= (x^{T}X + zII)^{T}X^{T}E[E|x] + (x^{T}X + zII)^{T}x^{T}x\beta$ = (XTX+ZII) XTXB = SZ'SZB whose S = (XTX+TII), S = XTX  $Cov(\hat{\beta}_{\epsilon}) = Cov((x^{T}X + zII)^{-1}X^{T}(x\beta + \epsilon))$   $= Cov((x^{T}X + zII)^{-1}X^{T}\epsilon)$   $= (x^{T}X + zII)^{-1}X^{T}Cov(\epsilon)X(x^{T}X + zII)^{-1}$ Cov(e) = 02, thosefose

 $Cov(\hat{B}_{\tau}) = \sigma^{2}(x^{T}x + \tau T)^{-1}x^{T}x(x^{T}x + \tau T)^{-1}$   $= \sigma^{2}S_{\tau}^{-1}SS_{\tau}^{-1}$