Given model as in (5), let  $\tilde{B} = (X'X + KI)'X'Y$ ,  $\tilde{X}_{j} = \frac{\lambda_{j}}{\lambda_{j}^{2} + K} \cdot Y^{*}J$  be canonical ridge regression estimates.

Recall: 
$$X = ULV'$$
  
 $X' = VLU'$ 

Note: Ur' L U by definition.

V'V = U'U = I since cols of V, U form orthonormal basis

Let & = VB

(L'L+ KI) is square, and diag, so inverse can be expressed element-wise

(L2+KI) and L2 can commute,

 $X'XV = VL^2$  by definition

$$x'x = VLU' \cdot ULV' = VL^{2}V'$$
 $(x'x)^{-1} = (VL^{2}V')^{-1}$ 
 $V' = V^{-1}, (VV')^{-1} = VV'$ 

Setup: 
$$Y = XB + E$$
,  $U_* = [U, Ur]$ ,  $Y_* = U_*'Y = [U']Y$ ,  $E_X = U_*'E$ 
 $Y = ULV'B + E$ 
 $Y_* = U_*'(ULV'B) + E_*$ 
 $Y_* = [U']ULV'B + E_* = [U'UL]V'B + E_*$ 
 $Y_* = [U] \times + E_*$ 

This is the canonical regression model of  $Y_*$  with parameter  $Y_*$ , and design matrix  $Y_*$ .

Use the ridge regression estimate for the regression parameter 8:

$$\widetilde{\delta} = \left( [L', \emptyset] \begin{bmatrix} L \\ 0 \end{bmatrix} + KI \right)^{-1} [L', \emptyset] Y_{*}$$

$$= \left( L'L + KI \right)^{-1} \left[ L', \emptyset \right] \begin{bmatrix} L \\ \emptyset \end{bmatrix} \widehat{\delta}$$

$$= \left( L'L + KI \right)^{-1} \left( L'L \right) \widehat{\delta} = \left( L^{2} + KI \right)^{-1} L^{2} \widehat{\delta}$$

$$V_{\widetilde{\delta}} = V \left[ \left( L^{2} + KI \right)^{-1} L^{2} \widehat{\delta} \right]$$

$$= V \left( L^{2} + KI \right)^{-1} L^{2} \cdot L^{-1} U'Y$$

 $= V(L^2 + KI)^{-1}V'X'Y$   $= (VL^2V' + KI)^{-1}X'Y$ 

= V (L2+KI)-1 LU'Y

From above: 
$$Y_* = \begin{bmatrix} L \\ 0 \end{bmatrix} \times + \times +$$

$$\Rightarrow \hat{X} = \begin{bmatrix} L^{-1} & 0 \end{bmatrix} Y_+ = \begin{bmatrix} L^{-1} & 0 \end{bmatrix} \begin{bmatrix} U' \\ U' \end{pmatrix} Y$$

$$\hat{X} = L^{-1} U' Y$$

$$\hat{X} = \frac{M_3' Y}{\lambda_3'} = \frac{y_* y_*}{\lambda_3'}$$