

5.8) (i) $y = \begin{bmatrix} 0 \\ 1 \\ 0 \\ 0 \\ 1 \end{bmatrix}$ (ii) $v = \begin{bmatrix} 0 & 1 \\ 1 & 0 \\ \vdots & \vdots \\ 1 & 0 \end{bmatrix} = [y, 1-y]$

$\hat{\beta}_{OLS} = (X^T X)^{-1} X^T y$ $\hat{\beta}_{LASSO} = (X^T X)^{-1} X^T y = \left[\hat{\beta}_0, \underbrace{(X^T X)^{-1} X^T (y - \hat{y})}_{\substack{\text{shrinkage} \\ \text{penalty}}} \right]$

Penalization: x_0^T
 $\hat{y}_0 = x_0^T \hat{\beta}_{LASSO} > 0.5$ $0 < \hat{y}_1 - \hat{y}_0 = x_1^T \hat{\beta}_1 - x_0^T \left(\begin{bmatrix} \hat{\beta}_0 \\ \hat{\beta}_1 \end{bmatrix} - \hat{\beta}_0 \right)$
 $= 2 x_1^T \hat{\beta}_1 - 1$
 $\Rightarrow x_0^T \hat{\beta}_{LASSO} > 0.5$

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4.2) a) This is just (4.11) and (4.10)
 $\hat{\sigma}_2(x) > \hat{\sigma}_1(x)$, or you can get this from (4.4)

c) $\hat{\Sigma}_0 \hat{\beta} = (\hat{\mu}_2 - \hat{\mu}_1) \frac{(\hat{\mu}_2 - \hat{\mu}_1)^T \hat{\beta}}{\lambda} = \lambda (\hat{\mu}_2 - \hat{\mu}_1)$

(4.5c) $\Rightarrow (N-2) \hat{\Sigma} \hat{\beta} + \frac{N \lambda}{N-2} (\hat{\mu}_2 - \hat{\mu}_1) = N (\hat{\mu}_2 - \hat{\mu}_1)$
 $\Rightarrow \hat{\Sigma} \hat{\beta} = (\hat{\mu}_2 - \hat{\mu}_1) \left(N - \frac{N \lambda}{N-2} \right) \frac{1}{N-2} = \gamma (\hat{\mu}_2 - \hat{\mu}_1)$
 $\hat{\beta} = \gamma \hat{\Sigma}^{-1} (\hat{\mu}_2 - \hat{\mu}_1)$

$x_0 = \hat{\mu}_2$
 $\hat{\Sigma}^{-1} (\hat{\mu}_2 - \hat{\mu}_1) = c$
 $\hat{\Sigma}^{-1} x_0 = \hat{\Sigma}^{-1} \hat{\mu}_2 = c$
 $\hat{\Sigma}^{-1} c^T (\hat{\mu}_2 - \hat{\mu}_1) = c$
 $\Rightarrow \hat{\Sigma}^{-1} (\hat{\mu}_2 - \hat{\mu}_1) = c$

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