

Ex. 3.30

$$\hat{\beta}^{\text{lasso}} = \underset{\beta}{\operatorname{argmin}} \left\{ \frac{1}{2} \sum_{i=1}^N (y_i - \beta_0 - \sum_{j=1}^P x_{ij} \beta_j)^2 + \lambda \sum_{j=1}^P |\beta_j| \right\}$$

Now, taking a hint from exercise 3.12 we will augment the X matrix and y vector as follows:

$$X^* = \begin{bmatrix} X \\ a \mathbb{I}_P \end{bmatrix}; \text{ where } a \text{ is some constant and } \mathbb{I}_P \text{ is the identity matrix of dim } P \times P$$

$$y^* = \begin{bmatrix} y \\ 0_P \end{bmatrix}, \text{ where } 0_P \text{ is a vector of } 0\text{'s of length } P.$$

Now,

$$\hat{\beta}^{\text{lasso}} = (y^* - X^* \beta)^T (y^* - X^* \beta) + \lambda \sum_j |\beta_j|$$

$$= (y - X\beta)^T (y - X\beta) + a^2 \beta^T \beta + \lambda \sum_j |\beta_j|$$

$$= (y - X\beta)^T (y - X\beta) + a^2 \sum_j \beta_j^2 + \lambda \sum_j |\beta_j|$$

and setting $a^2 = \lambda^* \alpha$ and $\lambda = \lambda^* (1 - \alpha)$ we obtain:

$$\hat{\beta}^{\text{lasso}} = \underset{\beta}{\operatorname{argmin}} \left\{ \sum_{i=1}^N (y_i - \beta_0 - \sum_{j=1}^P x_{ij} \beta_j)^2 + \lambda^* \sum_{j=1}^P (\alpha \beta_j^2 + (1 - \alpha) |\beta_j|) \right\}$$