

Here we give the ADMM algorithm for fitting the lasso problem:

$$\min_{\beta} \frac{1}{2n} (X\beta - y)^2 + \lambda |\beta|$$

We frame this question in ADMM way:

$$\min_{\beta, \alpha} \frac{1}{2n} (X\beta - y)^2 + \lambda |\alpha|$$

$$\text{under condition } \beta - \alpha = 0$$

We write the augmented Lagrangian

$$\frac{1}{2n} (X\beta - y)^2 + \lambda |\alpha| + y^T (\beta - \alpha) + \frac{\rho}{2} (\beta - \alpha)^2$$

The algorithm follows

$$\beta^{k+1} = \arg \min_{\beta} \frac{1}{2n} (X\beta - y)^2 + \frac{\rho}{2} (\beta - v^k)^2$$

$$\alpha^{k+1} = \arg \min_{\alpha} \lambda |\alpha| + \frac{\rho}{2} (\alpha - w^k)^2$$

$$u^{k+1} = u^k + \beta^{k+1} - \alpha^{k+1}$$

$$v^k = \alpha^k - u^k$$

$$w^k = \beta^{k+1} + u^k$$

Update for both  $\beta$  and  $\alpha$  are solvable problems

$$\beta^{k+1} = \arg \min_{\beta} \frac{1}{2n} (X\beta - y)^2 + \frac{\rho}{2} (\beta + v^k)^2 = (\frac{1}{n} X^T X + \rho I)^{-1} (\frac{1}{n} X^T y + \rho v^k)$$

$$\alpha^{k+1} = \arg \min_{\alpha} \lambda |\alpha| + \frac{\rho}{2} (\alpha - w^k)^2 = S_{\frac{\lambda}{\rho}}(w^k)$$