

# 1 FISTA

Our function:

$$\frac{1}{\sqrt{n}} \sum_{i=1}^n \sum_{t=1}^T [\hat{p}_t^{(k)}(1|S_t)(1 - \hat{p}_t^{(k)}(1|S_t))(\tilde{Y}_{i,t+1}^{(DR)} - f_t(S_t)^T \beta)^2] + \lambda \|\beta\|_1 - w^t \beta \quad (1)$$

We can let

$$f(\beta) = \frac{1}{\sqrt{n}} \sum_{i=1}^n \sum_{t=1}^T [\hat{p}_t^{(k)}(1|S_t)(1 - \hat{p}_t^{(k)}(1|S_t))(\tilde{Y}_{i,t+1}^{(DR)} - f_t(S_t)^T \beta)^2] - \omega^t \beta$$

$$g(\beta) = \lambda \|\beta\|_1$$

Then

$$\nabla f(\beta) = \frac{-2}{\sqrt{n}} \sum_{i=1}^n \sum_{t=1}^T \hat{p}_t^{(k)}(1|S_t)(1 - \hat{p}_t^{(k)}(1|S_t))(\tilde{Y}_{i,t+1}^{(DR)} - f_t(S_t)^T \beta) f_t(S_t) - \omega$$

$$\nabla^2 f(\beta) = \frac{2}{\sqrt{n}} \sum_{i=1}^n \sum_{t=1}^T \hat{p}_t^{(k)}(1|S_t)(1 - \hat{p}_t^{(k)}(1|S_t)) f_t(S_t) f_t(S_t)^T$$

Then  $L$  the Lipschitz constant of  $\nabla f$  is the  $L = \sup \|\nabla^2 f(\beta)\|$  the largest eigenvalue.

In the algorithm we also need to get  $p_L(y_k)$  and

$$p_L(y_k) = \underset{\beta}{\operatorname{argmin}} \left\{ \lambda \|\beta\|_1 + \frac{L}{2} (\beta - [y_k - \frac{1}{L} \nabla f(y_k)])^2 \right\}$$

Then  $x_k = (|y_k - \frac{1}{L} \nabla f(y_k)| - \frac{\lambda}{L})_+ \operatorname{sgn}(y_k - \frac{1}{L} \nabla f(y_k))$

The rest is just following the FISTA algorithm in page 12. Better to use FISTA with backtracking, and you even don't need  $L$ .