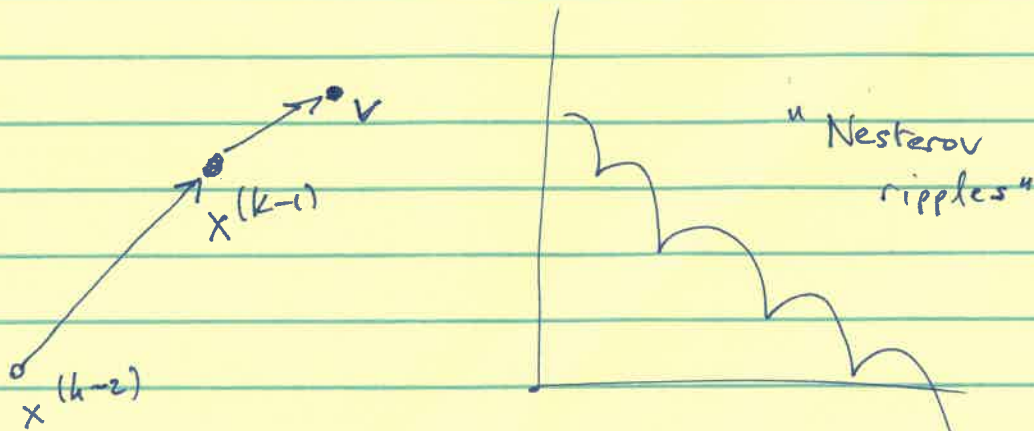
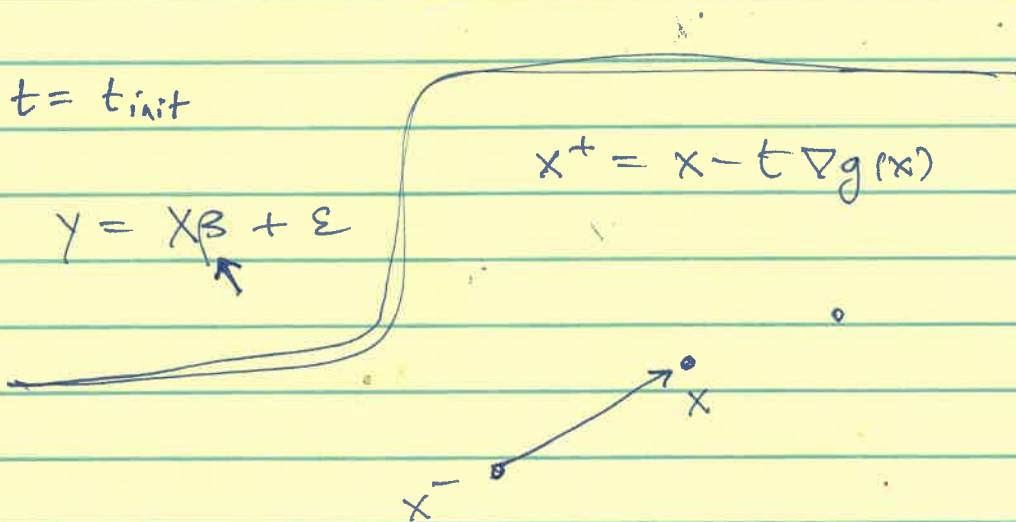


$$\text{prox}_t(x) = \underset{z}{\text{argmin}} \frac{1}{2t} \|x - z\|_2^2 + h(z)$$



NOT a descent method



$$\underset{B}{\text{min}} \frac{1}{2} \sum_{\substack{(i,j) \\ \in \Omega}} (Y_{ij} - B_{ij})^2 + \lambda (\|B\|_* + \|B\|_{\text{tr}})$$

$$\begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} \quad \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}$$

$$\frac{1}{n} \|y - X\beta\|_2^2 = \frac{1}{n} \sum_{i=1}^n \underbrace{(y_i - x_i^T \beta)^2}_{f_i(\beta)}$$

X $n \times p$.

x_i rows of x

$i=1, \dots, n$

$$\mathbb{E}[\nabla f_{i_k}(x)] = \frac{1}{m} \sum_{i=1}^m \nabla f_i(x)$$

$$p_i(\beta) = \frac{1}{1 + \exp(-x_i^T \beta)} \quad \text{logistic.}$$

$$\begin{aligned} x^{(k+2)} &= x^{(k+1)} - \epsilon \nabla f_2(x^{(k+1)}) \\ &= x^{(k)} - \epsilon [\nabla f_1(x^{(k)}) + \nabla f_2(x^{(k+1)})] \end{aligned}$$

$$F(x, \frac{\epsilon}{2}) = f(x) + \frac{\epsilon^2}{2} f''(x - \frac{\epsilon}{2})$$

SGE

Experience (Duchi)