

LINMA2171

Homework I

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Projected Gradient Method

(a) Computation of $\nabla f_l(\cdot)$

$$\begin{aligned}\nabla f_l(X)_{kl} &= \frac{\partial}{\partial_{kl}} \frac{1}{2} \sum_{i=1}^m \sum_{j=1}^n (X_{ij} - Y_{ij})^2 + \frac{\partial}{\partial_{kl}} \lambda \sum_{i=2}^{m-1} \sum_{j=2}^{n-1} (X_{i+1,j} - X_{i,j})^2 + (X_{i,j+1} - X_{i,j})^2 \\ &= \frac{1}{2} \sum_{i=1}^m \sum_{j=1}^n \frac{\partial}{\partial_{kl}} (X_{ij} - Y_{ij})^2 + \lambda \sum_{i=2}^{m-1} \sum_{j=2}^{n-1} \frac{\partial}{\partial_{kl}} (X_{i+1,j} - X_{i,j})^2 + \frac{\partial}{\partial_{kl}} (X_{i,j+1} - X_{i,j})^2 \\ &= (X_{kl} - Y_{kl}) + \lambda [2(X_{k,l} - X_{k-1,l}) - 2(X_{k+1,l} - X_{k,l}) + 2(X_{k,l} - X_{k,l-1}) - 2(X_{k,l+1} - X_{k,l})] \\ &= (X_{k,l} - Y_{k,l}) + 8\lambda X_{kl} - 2\lambda X_{k-1,l} - 2\lambda X_{k+1,l} - 2\lambda X_{k,l-1} - 2\lambda X_{k,l+1}\end{aligned}$$

This is a general expression for the kl component of the gradient, but we give a more precise expression hereafter to consider the "corner cases"

$$\nabla f_l(X)_{kl} = \begin{cases} (X_{kl} - Y_{kl}) & \text{if } i = 1 \text{ and/or } j = 1 \text{ or if } k = m \text{ and } l = n \\ (X_{kl} - Y_{kl}) + 4\lambda X_{kl} - 2\lambda X_{k+1,l} - 2\lambda X_{k,l+1} & \text{if } i=2 \text{ and } j=2 \\ (X_{kl} - Y_{kl}) + 6\lambda X_{kl} - 2\lambda X_{k+1,l} - 2\lambda X_{k,l-1} - 2\lambda X_{k,l+1} & \text{k=2 and l geq 2} \\ (X_{kl} - Y_{kl}) + 6\lambda X_{kl} - 2\lambda X_{k,l+1} - 2\lambda X_{k-1,l} - 2\lambda X_{k+1,l} & \text{l=2 and k geq 2} \\ (X_{kl} - Y_{kl}) + 2\lambda X_{kl} - 2\lambda X_{k-1,l} & \text{k=m and } 2 \leq l \leq n \\ (X_{kl} - Y_{kl}) + 2\lambda X_{kl} - 2\lambda X_{k,l-1} & \text{l = n and } 2 \leq k \leq m \\ (X_{kl} - Y_{kl}) + 8\lambda X_{kl} - 2\lambda X_{k-1,l} - 2\lambda X_{k+1,l} - 2\lambda X_{k,l-1} - 2\lambda X_{k,l+1} & \text{else} \end{cases}$$

(b) $\nabla f_l(\cdot)$ L -Lipschitz derivation

We recall that a function $F : \mathbb{R}^n \rightarrow \mathbb{R}^n$ is L -Lipschitz if there exists a constant L such that $\|F(x) - F(y)\| \leq L\|x - y\| \forall x, y \in \mathbb{R}^n$. To find L for $\nabla f_l(\cdot)$, we will consider the general case of the derivation above, since it will lead to upper bound for the corner cases and we will assume a 0 value for out of bounds index. This leads to:

$$\begin{aligned}
\|F(x) - F(y)\|^2 &\leq \sum_{i=1}^m \sum_{j=1}^n n(|X_{i,j} - Y_{i,j}| + 8\lambda|X_{i,j} - Y_{i,j}| + 2\lambda|X_{i,j+1} - Y_{i,j+1}| + 2\lambda|X_{i,j-1} - Y_{i,j-1}| + 2\lambda|X_{i+1,j} - Y_{i+1,j}|) \\
&= \sum_{i=1}^m \sum_{j=1}^n (|\Delta_{i,j}| + 8\lambda|\Delta_{i,j}| + 2\lambda|\Delta_{i-1,j}| + 2\lambda|\Delta_{i+1,j}| + 2\lambda|\Delta_{i,j-1}| + 2\lambda|\Delta_{i,j+1}|)^2 \\
&\leq 5 \sum_{i=1}^m \sum_{j=1}^n (|\Delta_{i,j}| + 16\lambda|\Delta_{i,j}|)^2 \\
&= \sum_{i=1}^m \sum_{j=1}^n (\sqrt{5}(1 + 16\lambda))^2 (\Delta_{i,j})^2
\end{aligned}$$

Where we can identify the L constant,

(c) $f_l(\cdot)$ convexity derivation

(e) Result of PGM

(f) Relation between λ and the quality

Proximal Gradient Method