

HWg-a

To prove $Aw-b \in \partial g(\lambda)$, we should show that

$Aw-b$ satisfy the $\partial g(\lambda)$ definition,

$$g(\lambda_1) \leq g(\lambda_2) + \partial g(\lambda_2)^T (\lambda_1 - \lambda_2)$$

Since g is concave, the sign here is wrong.

$$\times \rightarrow f(w) + \lambda_1^T (Aw-b) \geq f(w) + \lambda_2^T (Aw-b) + \partial g(\lambda_2)^T (\lambda_1 - \lambda_2)$$

Since $g(\lambda) = \min_w (f(w) + \lambda^T (Aw-b))$, w on the two sides are not the same!

$$\Rightarrow (\lambda_1 - \lambda_2)^T (Aw-b) \geq \partial g(\lambda_2)^T (\lambda_1 - \lambda_2)$$

now, we check if $Aw-b$ satisfies this condition $\rightarrow (\lambda_1 - \lambda_2)^T [Aw-b] \geq [Aw-b]^T (\lambda_1 - \lambda_2)$?

\leadsto This inequality is true & actually the equality holds, therefore ∂g can be

equal to $Aw-b$, which means

$$Aw-b \in \partial g(\lambda)$$

There
HW 3-b

There are 2 theorems as below:

① If f is closed & strong convex with parameter μ , then f^* has a Lipschitz continuous gradient with parameter $\frac{1}{\mu}$

② If f is convex and has a Lipschitz continuous with parameter L , then f^* is strong convex with parameter $\frac{1}{L}$.

proof of (1): By implication of strong convexity,
we have $\|z_n - z_y\| \geq \mu \|x - y\| \quad \begin{matrix} \checkmark z_n \in \partial f(x) \\ \checkmark z_y \in \partial f(y) \end{matrix}$

which implies

$$\|z_n - z_y\| \geq \mu \|\nabla f^*(z_n) - \nabla f^*(z_y)\|$$

Hence, f^* has a Lipschitz continuous gradient with $\frac{1}{\mu}$

proof of (2): By implication of Lipschitz continuous

gradient for convex f , we have:

continue of thw 3-b \rightarrow

$$(\nabla f(x) - \nabla f(y))^T (x - y) \geq \frac{1}{L} \|\nabla f(x) - \nabla f(y)\|^2$$

which implies:

$$(z_n - z_j)^T (x - y) \geq \frac{1}{L} (z_n - z_j)^T \nabla f^*(z_n) - \frac{1}{L} (z_n - z_j)^T \nabla f^*(z_j)$$

Hence, f^* is strongly convex with parameter $1/L$. \square

Therefore, the convergence rate is same as primal problem gradient descent problem with

$$\alpha_k = \frac{2}{\frac{1}{\mu} + \frac{1}{L}}$$

linear rate.

\rightarrow Primal feasible?

The solution is feasible, because the dual problem converges. Accordingly, the primal solution would be feasible.

In general, the statement does not necessarily hold.

Hint: check strong duality.

HW 3-C

$$\text{minimize } \frac{1}{N} \sum_{i \in N} f_i(w_i)$$

$$\text{s.t. } w_i = w_j \text{ for all } j \in N_i$$

$$L(w, \lambda) = \frac{1}{N} \sum_i \left(f_i(w_i) - \sum_{j \in N_i} \lambda_{ij} (w_i - w_j) \right)$$

$$l(w, \lambda) = \frac{1}{N} \sum_i \left(f_i(w_i) - \sum_{j \in N_i} \lambda_{ij} w_i + \sum_{j \in N_i} w_j \right)$$

lambda_{ji}

Step ① node i computes %

$$w_i^k = \arg \min_w \frac{1}{N} f_i(w) - \sum_{j \in N_i} \lambda_{ij}^k w_i$$

Step ②: node i sends w_i to N_i neighbors
 Communication cost = number of nodes \times average node degree.

Then compute:

$$\lambda_{ij}^{k+1} = \lambda_{ij}^k + \alpha^k (w_i - w_j) \text{ for all } j \in N_i$$

Convergence rate of the dual fornet is lower than primal
 because it is based on one hop communication.