

Consider that we have the following problem

$$f(x) = \frac{1}{n} \sum_{i=1}^n f_i(x) + \frac{\lambda}{2} \|x\|^2. \quad (1)$$

Each i -th node has its own dataset. That's why each f_i has the same form

$$f_i(x) = \frac{1}{m_i} \sum_{j=1}^{m_i} f_{ij}(x) + \frac{\lambda}{2} \|x\|^2. \quad (2)$$

We assume that all of f_{ij} are convex and f_i is μ -strongly convex. Moreover each f_{ij} has H-Lipshitz Hessian. Let's slightly specify each f_{ij} . We assume that

$$f_{ij}(x) = \varphi_{ij}(a_{ij}^T x) \quad (3)$$

and φ_{ij} is third times continuously differentiable. We can easily compute Hessian of f_{ij} at point x

$$H_{ij}(x) = \alpha_{ij}(x) a_{ij} a_{ij}^T, \quad (4)$$

where $\alpha_{ij}(x) = \varphi_{ij}''(a_{ij}^T x)$. We approximate Hessian $H_i(x)$ of f_i by

$$B_i(x) = \frac{\max_{j \in [m_i]} \alpha_{ij}(x)}{m_i} \sum_{j=1}^{m_i} a_{ij} a_{ij}^T \quad (5)$$

Since $\max_{j \in [m_i]} \alpha_{ij}(x) \geq \alpha_{ij}(x) \geq 0$ for all j , the $B_i(x) \succeq H_i(x) \succeq \mu I$. For such problem we want to run the following algorithm.

Algorithm 1 Max Coefficient Newton Method

Initialize: Choose starting iterates $x^0 \in \mathbb{R}^d$

for $k = 0, 1, 2, \dots$ **do** in parallel

 broadcast x^k to all workers \leftarrow **master node**

for $i = 0, 1, \dots, n$ **do** \leftarrow **i -th node**

 compute $\alpha_{ij}^k = \varphi_{ij}''(a_{ij}^T x^k)$

 compute $\beta_i^k = \max_{j \in [m_i]} \alpha_{ij}^k$

 broadcast β_i^k to master node

end for

$$B_i^k = \frac{\beta_i^k}{m_i} \sum_{j=1}^{m_i} a_{ij} a_{ij}^T + \lambda I$$

$$x^{k+1} = x^k - \left[\frac{1}{n} \sum_{i=1}^n B_i^k \right]^{-1} \left[\frac{1}{n} \sum_{i=1}^n \nabla f_i(x^k) \right]$$

end for

Let's work in one node case: $n = 1$. Let's denote $H(x) = \frac{1}{n} \sum_{i=1}^n \alpha_i(x) a_i a_i^T$, this is the true Hessian and $B(x) = \frac{\max_i \alpha_i(x)}{n} \sum_{i=1}^n a_i a_i^T = \frac{\beta(x)}{n} \sum_{i=1}^n a_i a_i^T$, this is the estimator of true Hessian. We know from mathematical analyzes the following Theorem of calculus

$$\nabla f(x) - \nabla f(y) = \int_0^1 \nabla^2 f(y + \tau(x - y))(x - y) d\tau. \quad (6)$$

Using this and other properties we obtain

$$\begin{aligned} \|x^{k+1} - x^*\| &= \|x^k - x^* - (B(x^k) + \lambda I)^{-1}[\nabla f(x^k) + \lambda x^k]\| \\ &= \|(B(x^k) + \lambda I)^{-1}[(B(x^k) + \lambda I)(x^k - x^*) - [\nabla f(x^k) + \lambda x^k]]\| \\ &\stackrel{\mu\text{-conv.}}{\leq} \frac{1}{\mu} \|(B(x^k) + \lambda I)(x^k - x^*) - [\nabla f(x^k) + \lambda x^k]\| \\ &= \frac{1}{\mu} \left\| \frac{\beta(x^k)}{n} \sum_{i=1}^n a_i a_i^T (x^k - x^*) + \lambda(x^k - x^*) - \frac{1}{n} \sum_{i=1}^n [\nabla f_i(x^k) + \lambda x^k] \right\| \\ &\stackrel{\nabla f(x^*) + \lambda x^* = 0}{=} \frac{1}{\mu n} \left\| \beta(x^k) \sum_{i=1}^n a_i a_i^T (x^k - x^*) - \sum_{i=1}^n (\nabla f_i(x^k) - \nabla f_i(x^*)) \right\| \\ &\stackrel{(6)}{=} \frac{1}{\mu n} \left\| \beta(x^k) \sum_{i=1}^n a_i a_i^T (x^k - x^*) - \sum_{i=1}^n \int_0^1 H_i(x^* + \tau(x^k - x^*))(x^k - x^*) d\tau \right\| \\ &= \frac{1}{\mu n} \left\| \beta(x^k) \sum_{i=1}^n a_i a_i^T (x^k - x^*) - \sum_{i=1}^n \int_0^1 \alpha_i(x^* + \tau(x^k - x^*)) a_i a_i^T (x^k - x^*) d\tau \right\| \\ &= \frac{1}{\mu n} \left\| \sum_{i=1}^n a_i a_i^T (x^k - x^*) \left[\beta(x^k) - \int_0^1 \alpha_i[x^* + \tau(x^k - x^*)] d\tau \right] \right\| \\ &\stackrel{\text{Jensen's ineq.}}{\leq} \frac{\|x^k - x^*\|}{\mu n} \sum_{i=1}^n \|a_i\|^2 \left| \beta(x^k) - \int_0^1 \alpha_i[x^* + \tau(x^k - x^*)] d\tau \right| \end{aligned} \quad (8)$$

We know that all f_i has Lipshitz Hessian

$$\begin{aligned} \|\alpha_i(x) a_i a_i^T - \alpha_i(y) a_i a_i^T\| &= |\alpha_i(x) - \alpha_i(y)| \|a_i\|^2 \leq H \|x - y\| \\ |\alpha_i(x) - \alpha_i(y)| &\leq \frac{H}{\|a_i\|^2} \|x - y\|. \end{aligned} \quad (9)$$

That's why we know that $\alpha_i(x)$ is Lipshitz function. Let's fix some $\tau_0 \in [0, 1]$, then we obtain

$$\begin{aligned} \left| \alpha_i[x^* + \tau_0 \cdot (x - x^*)] - \alpha_i[x^* + \tau(x - x^*)] \right| &\leq \frac{H}{\|a_i\|^2} \left| \{x^* + \tau_0(x - x^*)\} - \{x^* + \tau(x - x^*)\} \right| \\ &= \frac{H}{\|a_i\|^2} \|x - x^*\| |\tau_0 - \tau| \end{aligned} \quad (10)$$

Now let's use Lagrangian theorem for the difference between α_i as a function of τ and obtain

$$\left| \alpha_i[x^* + \tau_0 \cdot (x - x^*)] - \alpha_i[x^* + \tau(x - x^*)] \right| = \left| \alpha'_i[x^* + \tau(x - x^*)] \right|_{\tau=\tau^*} |\tau_0 - \tau|, \quad \tau^* \in [\tau_0, \tau]. \quad (11)$$

Let's use (11) in (10) and obtain

$$\left| \alpha'_i[x^* + \tau(x - x^*)] \right|_{\tau=\tau^*} |\tau_0 - \tau| \leq \frac{H}{\|a_i\|^2} \|x - x^*\| |\tau_0 - \tau| \quad (12)$$

$$\left| \alpha'_i[x^* + \tau(x - x^*)] \right|_{\tau=\tau^*} \leq \frac{H}{\|a_i\|^2} \|x - x^*\| \quad (13)$$

We see that the right in (13) doesn't depend on τ . Now if $\tau \rightarrow \tau_0$ then from continuity of α' we obtain that $\tau^* \rightarrow \tau_0$ and

$$\left| \alpha'_i[x^* + \tau(x - x^*)]_{\tau=\tau_0} \right| \leq \frac{H}{\|a_i\|^2} \|x - x^*\| \quad (14)$$

One can show that for $x = x^*$ this inequality still true because

$$\alpha_i(x^* + \tau(x - x^*)) = \varphi''_i(a_i^T(x^* + \tau(x - x^*))) \quad (15)$$

$$\left[\alpha_i(x^* + \tau(x - x^*)) \right]'_{\tau} = \left[\varphi''_i(a_i^T(x^* + \tau(x - x^*))) \right]'_{\tau} \quad (16)$$

$$\left[\alpha_i(x^* + \tau(x - x^*)) \right]'_{\tau} = \varphi'''_i(a_i^T(x^* + \tau(x - x^*))) \times [a_i^T(x - x^*)] \quad (17)$$

That's why if $x = x^*$ inequality (14) is still true. We have that

$$\begin{aligned} \left| \beta(x^k) - \int_0^1 \alpha_i[x^* + \tau(x^k - x^*)] d\tau \right| &= \left| \beta(x^k) - \alpha_i(x^k) + \alpha_i(x^k) - \int_0^1 \alpha_i[x^* + \tau(x^k - x^*)] d\tau \right| \\ &\leq \left| \beta(x^k) - \alpha_i(x^k) \right| + \left| \alpha_i(x^k) - \int_0^1 \alpha_i[x^* + \tau(x^k - x^*)] d\tau \right| \\ &= \left| \beta(x^k) - \alpha_i(x^k) \right| + \left| \int_0^1 \left\{ \alpha_i(x^k) - \alpha_i[x^* + \tau(x^k - x^*)] \right\} d\tau \right| \\ &= \left| \beta(x^k) - \alpha_i(x^k) \right| + \left| \int_0^1 \frac{H}{\|a_i\|^2} \|x^k - (x^* + \tau(x^k - x^*))\| d\tau \right| \\ &= \left| \beta(x^k) - \alpha_i(x^k) \right| + \int_0^1 \frac{H}{\|a_i\|^2} \|x^k - (x^* + \tau(x^k - x^*))\| d\tau \\ &= \left| \beta(x^k) - \alpha_i(x^k) \right| + \frac{H}{2\|a_i\|^2} \|x^k - x^*\| \end{aligned} \quad (18)$$

It looks like that we can't prove that $\|B^k - H^k\|$ is upper bounded enough. Other way how we can prove is

$$x^{k+1} - x^* = x^k - x^* - (B^k + \lambda I)^{-1} (\nabla f(x^k) + \lambda x^k) \quad (19)$$

$$= x^k - x^* - (B^k + \lambda I)^{-1} \int_0^1 \left[\nabla^2 f(x^* + \tau(x^k - x^*)) + \lambda I \right] (x^k - x^*) d\tau$$

$$= x^k - x^* - (B^k + \lambda I)^{-1} \underbrace{\int_0^1 \left[\nabla^2 f(x^* + \tau(x^k - x^*)) + \lambda I \right] d\tau}_{G^k} (x^k - x^*)$$

$$= \left(I - (B^k + \lambda I)^{-1} G^k \right) (x^k - x^*) \quad (20)$$

$$= (B^k + \lambda I)^{-1} \left(B^k - \int_0^1 \nabla^2 f(x^* + \tau(x^k - x^*)) d\tau \right) (x^k - x^*). \quad (21)$$

0.1 New notes

Let's work with scaled version of Max. Coefficient Method. Hessian estimator is

$$B^k = \frac{1}{n} \max_i \left\{ \frac{\alpha_i(x^k)}{\alpha_i(x^*)} \right\} \sum_{i=1}^n \alpha_i(x^*) a_i a_i^T = \frac{\beta(x^k)}{n} \sum_{i=1}^n \alpha_i(x^*) a_i a_i^T. \quad (22)$$

Using this and other properties we obtain

$$\begin{aligned} \|x^{k+1} - x^*\| &= \|x^k - x^* - (B(x^k) + \lambda I)^{-1} [\nabla f(x^k) + \lambda x^k]\| \\ &= \|(B(x^k) + \lambda I)^{-1} [B(x^k) + \lambda I](x^k - x^*) - [\nabla f(x^k) + \lambda x^k]\| \\ &\stackrel{\lambda\text{-conv.}}{\leq} \frac{1}{\lambda} \| [B(x^k) + \lambda I](x^k - x^*) - [\nabla f(x^k) + \lambda x^k] \| \\ &= \frac{1}{\lambda} \left\| \frac{\beta(x^k)}{n} \sum_{i=1}^n \alpha_i(x^*) a_i a_i^T (x^k - x^*) + \lambda(x^k - x^*) - \frac{1}{n} \sum_{i=1}^n [\nabla f_i(x^k) + \lambda x^k] \right\| \\ &= \left| \nabla f(x^*) + \lambda x^* = 0 \right| \\ &= \frac{1}{\lambda n} \left\| \beta(x^k) \sum_{i=1}^n \alpha_i(x^*) a_i a_i^T (x^k - x^*) - \sum_{i=1}^n (\nabla f_i(x^k) - \nabla f_i(x^*)) \right\| \\ &\stackrel{(6)}{=} \frac{1}{\lambda n} \left\| \beta(x^k) \sum_{i=1}^n \alpha_i(x^*) a_i a_i^T (x^k - x^*) - \sum_{i=1}^n \int_0^1 H_i(x^* + \tau(x^k - x^*)) (x^k - x^*) d\tau \right\| \\ &= \frac{1}{\lambda n} \left\| \beta(x^k) \sum_{i=1}^n \alpha_i(x^*) a_i a_i^T (x^k - x^*) - \sum_{i=1}^n \alpha_i(x^*) \int_0^1 \frac{\alpha_i(x^* + \tau(x^k - x^*))}{\alpha_i(x^*)} a_i a_i^T (x^k - x^*) d\tau \right\| \\ &= \frac{1}{\lambda n} \left\| \sum_{i=1}^n \alpha_i(x^*) a_i a_i^T (x^k - x^*) \left[\beta(x^k) - \int_0^1 \frac{\alpha_i[x^* + \tau(x^k - x^*)]}{\alpha_i(x^*)} d\tau \right] \right\| \\ &\stackrel{\text{Jensen}}{\leq} \frac{\|x^k - x^*\|}{\lambda n} \sum_{i=1}^n \alpha_i(x^*) \|a_i\|^2 \left| \beta(x^k) - \int_0^1 \frac{\alpha_i[x^* + \tau(x^k - x^*)]}{\alpha_i(x^*)} d\tau \right|. \end{aligned} \quad (23)$$

We know that $\alpha_i(x)$ is Lipschitz continuous with constant $\frac{H}{\|a_i\|^2}$ from (9). That's why we have can use it for $x = x^* + \tau(x^k - x^*)$ and $y = x^*$

$$\alpha_i(x^*) - \frac{H}{\|a_i\|^2} \tau \|x^k - x^*\| \leq \alpha_i[x^* + \tau(x^k - x^*)] \leq \alpha_i(x^*) + \frac{H}{\|a_i\|^2} \tau \|x^k - x^*\| \quad (24)$$

$$\begin{aligned} 1 - \frac{H}{\alpha_i(x^*) \|a_i\|^2} \tau \|x^k - x^*\| &\leq \frac{\alpha_i[x^* + \tau(x^k - x^*)]}{\alpha_i(x^*)} \leq 1 + \frac{H}{\alpha_i(x^*) \|a_i\|^2} \tau \|x^k - x^*\| \\ 1 - \frac{H}{2\alpha_i(x^*) \|a_i\|^2} \|x^k - x^*\| &\leq \int_0^1 \frac{\alpha_i[x^* + \tau(x^k - x^*)]}{\alpha_i(x^*)} d\tau \leq 1 + \frac{H}{2\alpha_i(x^*) \|a_i\|^2} \|x^k - x^*\|. \end{aligned} \quad (25)$$

Let's $j_k = \arg \max_i \left\{ \frac{\alpha_i(x^k)}{\alpha_i(x^*)} \right\}$, then we have $\beta(x^k) = \frac{\alpha_{j_k}(x^k)}{\alpha_{j_k}(x^*)}$. We can write for $\beta(x^k)$ the consequences of Lipschitzness

$$\alpha_{j_k}(x^*) - \frac{H}{\|a_{j_k}\|^2} \|x^k - x^*\| \leq \alpha_{j_k}(x^k) \leq \alpha_{j_k}(x^*) + \frac{H}{\|a_{j_k}\|^2} \|x^k - x^*\| \quad (26)$$

$$\begin{aligned} 1 - \frac{H}{\alpha_{j_k}(x^*) \|a_{j_k}\|^2} \|x^k - x^*\| &\leq \frac{\alpha_{j_k}(x^k)}{\alpha_{j_k}(x^*)} \leq 1 + \frac{H}{\alpha_{j_k}(x^*) \|a_{j_k}\|^2} \|x^k - x^*\| \\ 1 - \frac{H}{\min_i \{\alpha_i(x^*) \|a_i\|^2\}} \|x^k - x^*\| &\leq \beta(x^k) \leq 1 + \frac{H}{\min_i \{\alpha_i(x^*) \|a_i\|^2\}} \|x^k - x^*\|. \end{aligned} \quad (27)$$

Using (25) and (27) we obtain

$$\begin{aligned}
\left(1 - \frac{H}{\min_i \{\alpha_i(x^*) \|a_i\|^2\}} \|x^k - x^*\|\right) &- \left(1 + \frac{H}{2\alpha_i(x^*) \|a_i\|^2} \|x^k - x^*\|\right) \leq \beta(x^k) - \int_0^1 \frac{\alpha_i[x^* + \tau(x^k - x^*)]}{\alpha_i(x^*)} d\tau \\
&\leq \left(1 + \frac{H}{\min_i \{\alpha_i(x^*) \|a_i\|^2\}} \|x^k - x^*\|\right) - \left(1 - \frac{H}{2\alpha_i(x^*) \|a_i\|^2} \|x^k - x^*\|\right) \\
-H\|x^k - x^*\| \left(\frac{1}{\min_i \{\alpha_i(x^*) \|a_i\|^2\}} + \frac{1}{2\alpha_i(x^*) \|a_i\|^2} \right) &\leq \beta(x^k) - \int_0^1 \frac{\alpha_i[x^* + \tau(x^k - x^*)]}{\alpha_i(x^*)} d\tau \\
&\leq H\|x^k - x^*\| \left(\frac{1}{\min_i \{\alpha_i(x^*) \|a_i\|^2\}} + \frac{1}{2\alpha_i(x^*) \|a_i\|^2} \right)
\end{aligned}$$

That's why we have that

$$\left| \beta(x^k) - \int_0^1 \frac{\alpha_i[x^* + \tau(x^k - x^*)]}{\alpha_i(x^*)} d\tau \right| \leq H\|x^k - x^*\| \left(\frac{1}{\min_i \{\alpha_i(x^*) \|a_i\|^2\}} + \frac{1}{2\alpha_i(x^*) \|a_i\|^2} \right). \quad (28)$$

0.2 New notes (version 2)

We need to assume the following properties of $B(x)$ and $\nabla^2 f(x)$

1. f has Lipshitz Hessian:

$$\begin{aligned} \nabla^2 f(x^*) - H\tau\|x^k - x^*\| &\leq \nabla^2 f\left[x^* + \tau(x^k - x^*)\right] \leq \nabla^2 f(x^*) + H\tau\|x^k - x^*\| \\ \nabla^2 f(x^*) - \frac{H}{2}\|x^k - x^*\| &\leq \int_0^1 \nabla^2 f\left[x^* + \tau(x^k - x^*)\right] d\tau \leq \nabla^2 f(x^*) + \frac{H}{2}\|x^k - x^*\|. \end{aligned} \quad (29)$$

2. $B(x^*) = \nabla^2 f(x^*)$ and $B(x^k)$ is also M-Lipshitz

$$\begin{aligned} B(x^*) - M\|x^k - x^*\| &\leq B(x^k) \leq B(x^*) + M\|x^k - x^*\| \\ \nabla^2 f(x^*) - M\|x^k - x^*\| &\leq B(x^k) \leq \nabla^2 f(x^*) + M\|x^k - x^*\|. \end{aligned} \quad (30)$$

Combining the above we get

$$-\left(M + \frac{H}{2}\right)\|x^k - x^*\| \leq B(x^k) - \int_0^1 \nabla^2 f\left[x^* + \tau(x^k - x^*)\right] d\tau \leq \left(M + \frac{H}{2}\right)\|x^k - x^*\|. \quad (31)$$

Now we use it in the following inequality

$$\|x^{k+1} - x^*\| \leq \frac{1}{\lambda}\|x^k - x^*\| \left\| B(x^k) - \int_0^1 \nabla^2 f\left[x^* + \tau(x^k - x^*)\right] d\tau \right\| \quad (32)$$

we get

$$\|x^{k+1} - x^*\| \leq \frac{M + H/2}{\lambda}\|x^k - x^*\|^2. \quad (33)$$