

We have K clients, random fraction C is selected

→ global model sent → perform computation on edge
↓
send update

P_k = set of indexes of data points on client k

$$n_k = |P_k|$$

Let $f(w)$ be our objective function that we minimize

$$f(w) = \sum_{i=1}^N f_i(w) \quad , \text{ here } f_i(w) = l(x_i, y_i; w_i)$$

In our case

$$f(w) = \sum_{k=1}^K \frac{n_k}{n} F_k(w) \quad , \quad F_k(w) = \frac{1}{n_k} \sum_{i \in P_k} f_i(w)$$

Let $C=1$,

→ each client compute $g_k = \nabla F_k(w_t)$

→ then after then for the central server
Update

$$\Rightarrow w_{t+1} \rightarrow w_t - \eta \sum_{k=1}^K \frac{n_k}{n} g_k$$

Algorithm 1 FederatedAveraging. The K clients are indexed by k ; B is the local minibatch size, E is the number of local epochs, and η is the learning rate.

Server executes:

initialize w_0

for each round $t = 1, 2, \dots$ **do**

$m \leftarrow \max(C \cdot K, 1)$

$S_t \leftarrow$ (random set of m clients)

for each client $k \in S_t$ **in parallel do**

$w_{t+1}^k \leftarrow \text{ClientUpdate}(k, w_t)$

$m_t \leftarrow \sum_{k \in S_t} n_k$

$w_{t+1} \leftarrow \sum_{k \in S_t} \frac{n_k}{m_t} w_{t+1}^k$ // Erratum⁴

ClientUpdate(k, w): // Run on client k

$B \leftarrow$ (split \mathcal{P}_k into batches of size B)

for each local epoch i from 1 to E **do**

for batch $b \in B$ **do**

$w \leftarrow w - \eta \nabla \ell(w; b)$

return w to server

global w

select m out of K

update each client

Total No of data points for that epoch w_t^k

Now, there exist another algorithm that can replace SGD here and that is)

Nesterov's Accelerated GD

$f: \mathbb{R}^d \rightarrow \mathbb{R}$

Just have f as the objective function to optimize.

$$f = \frac{1}{2} x^T A x - b^T x, \quad A := \begin{bmatrix} 4 & 1 \\ 1 & 3 \end{bmatrix}, \quad b := \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

Here f is m -strongly convex and M -smooth

★ Define: f is m -strongly convex if, there exists $m > 0$ such that $\forall x, y \in \mathbb{R}^d$ and $\forall \lambda \in [0, 1]$

$$f(\lambda x + (1-\lambda)y) \leq \lambda f(x) + (1-\lambda)f(y) - \frac{m}{2} \lambda(1-\lambda) \|x - y\|^2$$

We will use its equivalent characterizations to prove

Show

$$\|\nabla f(x) - \nabla f(y)\| \geq m \|x - y\|$$

$$\text{Now for } \nabla f(x), \Rightarrow \nabla f(x) = \frac{1}{2} (A + A^T)x - b^T$$

$$\text{As } A \text{ is symmetric, } \Rightarrow \nabla f(x) = Ax - b^T$$

$$\therefore \Rightarrow \nabla f(x) - \nabla f(y) = Ax - b^T - (Ay - b^T) \\ = A(x - y)$$

$$\Rightarrow \|\nabla f(x) - \nabla f(y)\| = \|A(x - y)\|$$

Regarding A , $\begin{bmatrix} 4 & 1 \\ 1 & 3 \end{bmatrix}$, let's consider $\Rightarrow \begin{bmatrix} p & q \end{bmatrix} \begin{bmatrix} 4 & 1 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} p \\ q \end{bmatrix}$
 for p and $q \neq 0$ simultaneously $\Rightarrow \begin{bmatrix} 4p+q & p+3q \end{bmatrix} \begin{bmatrix} p \\ q \end{bmatrix}$
 $\Rightarrow 4p^2 + pq + pq + 3q^2$
 $\Rightarrow 4p^2 + 2pq + 3q^2$
 $\Rightarrow (p+q)^2 + 3p^2 + 2q^2$
 $\underbrace{\hspace{10em}}_{>0} +ve$

Hence A is +ve definite

$$\therefore A \geq \lambda_{\min}(A) I \rightarrow \textcircled{1} \text{ where } \lambda_{\min}(A) \text{ is the smallest eigenvalue}$$

For eigenvalues, $(A - \lambda I)x = 0$

$$|A - \lambda I| = 0$$

$$\Rightarrow \begin{vmatrix} 4-\lambda & 1 \\ 1 & 3-\lambda \end{vmatrix} = 0$$

$$\Rightarrow (4-\lambda)(3-\lambda) - 1 = 0$$

$$\Rightarrow 12 - 7\lambda + \lambda^2 - 1 = 0$$

$$\Rightarrow \lambda^2 - 7\lambda + 11 = 0$$

$$\lambda = \frac{7 \pm \sqrt{49 - 44}}{2} = \frac{7 \pm \sqrt{5}}{2}$$

$$\lambda_{\min} = \frac{7 - \sqrt{5}}{2}$$

Using $\textcircled{1} \Rightarrow \|\nabla f(x) - \nabla f(y)\| \geq \underbrace{\lambda_{\min}}_{\downarrow m} \|x - y\|$

$$m = \frac{7 - \sqrt{5}}{2}$$

Hence f is

m -strongly convex

★ Define: $f: \mathbb{R}^d \rightarrow \mathbb{R}$ satisfies

$$\|\nabla f(x) - \nabla f(y)\|_2 \leq M \|x - y\|_2$$

$M > 0 \Rightarrow M$ -smooth:

From before

$$\Rightarrow \|A(x - y)\|$$

Since $A \leq \lambda_{\max}(A) I$

$$\therefore \|A(x - y)\| \leq \|\lambda_{\max}(A) I\| \|x - y\|$$

$$\leq \lambda_{\max}(A) \|x - y\|$$

Hence proved.

$$M = \lambda_{\max}(A)$$

$$M = \frac{7 + \sqrt{5}}{2}$$