A MATHEMATICAL DEMONSTRATION

We consider the classification problem with C classes in Federated Learning. The function $f: \mathcal{X} \to \mathcal{Z}$ maps data \mathbf{x} to the probability simplex \mathcal{Z} and $\mathcal{Z} = \{\mathbf{z} | \sum_{i=1}^{C} z_i = 1; z_i \geq 0, \forall i \in [C]\}$, where z_i is the probability of class i. The population cross-entropy loss $l(\omega)$ is defined in (9).

$$l(\omega) = \mathbb{E}_{\mathbf{x}, y \sim p} \left[\sum_{i=1}^{C} \mathbb{I}_{y=i} (-\log f_i(\mathbf{x}, \omega)) \right]$$
$$= \sum_{i=1}^{C} p(y=i) \mathbb{E}_{\mathbf{x}|y=i} [-\log f_i(\mathbf{x}, \omega)].$$
(9)

To bound the divergence between the weights obtained by the FedAVG algorithm ω_{mT}^f and the optimal weights ω_{mT}^* on the test dataset, an intermediate variable ω_{mT}^c is introduced in (10) to assist the proof. The ω_{mT}^c physically represents the weights trained over the data from the selected clients in a centralized manner. The m is the round number and T is the number of optimization steps conducted in each round.

$$||\omega_{mT}^{f} - \omega_{mT}^{*}|| \leq ||\omega_{mT}^{f} - \omega_{mT}^{c} + \omega_{mT}^{c} - \omega_{mT}^{*}||$$

$$\leq ||\omega_{mT}^{f} - \omega_{mT}^{c}|| + ||\omega_{mT}^{c} - \omega_{mT}^{*}||.$$
(10)

The optimization step in local SGD is shown in (11), where p_l^k is the local data distribution of client k and η is the learning rate.

$$\omega_t^k = \omega_{t-1}^k - \eta \nabla_{\omega} l(\omega)$$

$$= \omega_{t-1}^k - \eta \sum_{i=1}^C p_l^k(y=i) \nabla_{\omega} \mathbb{E}_{x|y=i} [-\log f_i(x, \omega_{t-1}^k)].$$
(11)

The optimization step in centralized SGD is shown in (12) and $p_o(y=j) = \sum_{k \in \mathcal{S}} p_l^k(y=j)/|\mathcal{S}|$, which is the population data distribution.

$$\omega_{t}^{c} = \omega_{t-1}^{c} - \eta \sum_{i=1}^{C} p_{o}(y=i) \nabla_{\omega} \mathbb{E}_{x|y=i} [-\log f_{i}(x, \omega_{t-1}^{c})].$$
 (12)

We will next derive the boundaries of $||\omega_{mT}^f - \omega_{mT}^c||$ and $||\omega_{mT}^c - \omega_{mT}^*||$ in Section A.1 and Section A.2 separately.

A.1 Boundary of $||\omega_{mT}^f - \omega_{mT}^c||$

$$\begin{split} &\|\omega_{mT}^{f} - \omega_{mT}^{c}\| = \|\frac{1}{K}\sum_{k=1}^{K}\omega_{mT}^{k} - \omega_{mT}^{c}\| \\ &= \|\frac{1}{K}\sum_{k=1}^{K}(\omega_{mT-1}^{k} - \eta\sum_{i=1}^{C}p_{l}^{k}(y=i)\nabla_{\omega}\mathbb{E}_{x|y=i}[-\log f_{i}(x,\omega_{mT-1}^{k})]) \\ &- (\omega_{mT-1}^{c} - \eta\sum_{i=1}^{C}p_{o}(y=i)\nabla_{\omega}\mathbb{E}_{x|y=i}[-\log f_{i}(x,\omega_{mT-1}^{c})])\| \\ &\leq \|\frac{1}{K}\sum_{k=1}^{K}\omega_{mT-1}^{k} - \omega_{mT-1}^{c}\| + \eta\|\frac{1}{K}\sum_{k=1}^{K}\sum_{i=1}^{C}p_{l}^{k}(y=i) \\ &(\nabla_{\omega}\mathbb{E}_{x|y=i}[-\log f_{i}(x,\omega_{mT-1}^{k})] - \nabla_{\omega}\mathbb{E}_{x|y=i}[-\log f_{i}(x,\omega_{mT-1}^{c})])\| \end{split}$$

$$\stackrel{\text{①}}{\leq} \frac{1}{K} \sum_{k=1}^{K} \|\omega_{mT-1}^{k} - \omega_{mT-1}^{c}\| + \frac{\eta \lambda}{K} \sum_{k=1}^{K} \sum_{i=1}^{C} p_{l}^{k}(y=i) \|\omega_{mT-1}^{k} - \omega_{mT-1}^{c}\| \\
= \frac{1}{K} \sum_{k=1}^{K} (1 + \eta \lambda) \|\omega_{mT-1}^{k} - \omega_{mT-1}^{c}\|. \tag{13}$$

The inequality ① holds because we assume $\nabla_{\omega}\mathbb{E}_{x|y=i}[-\log f_i(x,\omega)]$ is λ -Lipschitz for $x,y\sim p$. In that case $\|\nabla_{\omega}\mathbb{E}_{x|y=i}[-\log f_i(x,\omega_1)] - \nabla_{\omega}\mathbb{E}_{x|y=i}[-\log f_i(x,\omega_2)]\| \leq \lambda\|\omega_1-\omega_2\|$. Then we have

$$\begin{split} &\|\omega_{mT-1}^{k} - \omega_{mT-1}^{c}\| \\ &= \|\omega_{mT-2}^{k} - \eta \sum_{i=1}^{C} p_{l}^{k}(y=i) \nabla_{\omega} \mathbb{E}_{x|y=i} [-\log f_{i}(x, \omega_{mT-2}^{k})] \\ &- \omega_{mT-2}^{c} + \eta \sum_{i=1}^{C} p_{o}(y=i) \nabla_{\omega} \mathbb{E}_{x|y=i} [-\log f_{i}(x, \omega_{mT-2}^{c})] \| \\ &\leq \|\omega_{mT-2}^{k} - \omega_{mT-2}^{c}\| + \eta \| \sum_{i=1}^{C} p_{l}^{k}(y=i) (\nabla_{\omega} \mathbb{E}_{x|y=i} [-\log f_{i}(x, \omega_{mT-2}^{k})] \| \\ &- \nabla_{\omega} \mathbb{E}_{x|y=i} [-\log f_{i}(x, \omega_{mT-2}^{c})]) \| \\ &+ \eta \| \sum_{i=1}^{C} (p_{l}^{k}(y=i) - p_{o}(y=i)) \nabla_{\omega} \mathbb{E}_{x|y=i} [-\log f_{i}(x, \omega_{mT-2}^{c})] \| \\ &\leq \|\omega_{mT-2}^{k} - \omega_{mT-2}^{c}\| + \eta \sum_{i=1}^{C} p_{l}^{k}(y=i) \lambda \|\omega_{mT-2}^{k} - \omega_{mT-2}^{c}\| \\ &+ \eta g(\omega_{mT-2}^{c}) \|p_{l}^{k} - p_{o}\|_{1} \\ &= (1 + \eta \lambda) \|\omega_{mT-2}^{k} - \omega_{mT-2}^{c}\| + \eta g(\omega_{mT-2}^{c}) \|p_{l}^{k} - p_{o}\|_{1} \end{split}$$

Note that $\mathbf{g}(\omega) = \max_{i=1}^{C} \|\nabla_{\omega} \mathbb{E}_{x|y=i}[-\log f_i(x,\omega)]\|$. Formula (14) implies that the weight divergence after each step of optimization of client k is restricted by the weight divergence from the last step plus a term which is related to the discrepancy between p_l^k and p_o .

Then, by induction, we have

$$\begin{split} &\|\omega_{mT-1}^{k} - \omega_{mT-1}^{c}\| \\ &\leq (1+\eta\lambda)\|\omega_{mT-2}^{k} - \omega_{mT-2}^{c}\| + \eta g(\omega_{mT-2}^{c})||p_{l}^{k} - p_{o}||_{1} \\ &\leq (1+\eta\lambda)^{2}\|\omega_{mT-3}^{k} - \omega_{mT-3}^{c}\| + (1+\eta\lambda)\eta g(\omega_{mT-3}^{c})||p_{l}^{k} - p_{o}||_{1} \\ &+ \eta g(\omega_{mT-2}^{c})||p_{l}^{k} - p_{o}||_{1} \\ &\leq (1+\eta\lambda)^{T-1}\|\omega_{(m-1)T}^{k} - \omega_{(m-1)T}^{c}\| \\ &+ \eta \sum_{j=2}^{T} g(\omega_{mT-j}^{c})(1+\eta\lambda)^{j-2}||p_{l}^{k} - p_{o}||_{1}. \end{split} \tag{15}$$

Therefore, we have

$$\|\omega_{mT}^{f} - \omega_{mT}^{c}\| \le \frac{1}{K} \sum_{i=1}^{K} (1 + \eta \lambda) \|\omega_{mT-1}^{k} - \omega_{mT-1}^{c}\|$$

$$\le \frac{1}{K} \sum_{i=1}^{K} [(1 + \eta \lambda)^{T} \|\omega_{(m-1)T}^{k} - \omega_{(m-1)T}^{c}\|$$

$$+ \eta || p_l^k - p_o ||_1 \left(\eta \sum_{j=2}^T \mathbf{g}(\omega_{mT-j}^c) (1+\lambda)^{j-1} \right)]. \tag{16}$$

A.2 Boundary of $||\omega_{mT}^c - \omega_{mT}^*||$ The boundary of $||\omega_{mT}^c - \omega_{mT}^*||$ is derived in (17). The derivation process in (17) is similar to the process in (13).

$$\begin{split} &\|\omega_{mT}^{c}-\omega_{mT}^{*}\|\\ =&\|\omega_{mT-1}^{c}-\eta\sum_{i=1}^{C}p_{o}(y\!=\!i)\nabla_{\omega}\mathbb{E}_{x|y=i}[-\log f_{i}(x,\omega_{mT-1}^{c})]\\ &-\omega_{mT-1}^{*}+\eta\sum_{i=1}^{C}p_{u}(y\!=\!i)\nabla_{\omega}\mathbb{E}_{x|y=i}[-\log f_{i}(x,\omega_{mT-1}^{*})]\|\\ \leq&\|\omega_{mT-1}^{c}-\omega_{mT-1}^{*}\|+\eta\|\sum_{i=1}^{C}p_{o}(y\!=\!i)\nabla_{\omega}\mathbb{E}_{x|y=i}[-\log f_{i}(x,\omega_{mT-1}^{c})]\\ &-\sum_{i=1}^{C}p_{u}(y\!=\!i)\nabla_{\omega}\mathbb{E}_{x|y=i}[-\log f_{i}(x,\omega_{mT-1}^{*})]\|\\ \leq&\|\omega_{mT-1}^{c}-\omega_{mT-1}^{*}\|+\eta\|\sum_{i=1}^{C}p_{o}(y\!=\!i)(\nabla_{\omega}\mathbb{E}_{x|y=i}[-\log f_{i}(x,\omega_{mT-1}^{c})]\\ &-\nabla_{\omega}\mathbb{E}_{x|y=i}[-\log f_{i}(x,\omega_{mT-1}^{*})])\|\\ &+\eta\|\sum_{i=1}^{C}(p(y\!=\!i)-p_{u}(y\!=\!i))\nabla_{\omega}\mathbb{E}_{x|y=i}[-\log f_{i}(x,\omega_{mT-1}^{*})]\| \end{split}$$

$$\leq \|\omega_{mT-1}^{c} - \omega_{mT-1}^{*}\| + \eta \sum_{i=1}^{C} p_{o}(y=i)\lambda \|\omega_{mT-1}^{c} - \omega_{mT-1}^{*}\|$$

$$+ \eta g(\omega_{mT-1}^{*}) \|p_{o} - p_{u}\|_{1}$$

$$= (1 + \eta\lambda) \|\omega_{mT-1}^{c} - \omega_{mT-1}^{*}\| + \eta g(\omega_{mT-1}^{*}) \|p_{o} - p_{u}\|_{1}$$

$$\leq (1 + \eta\lambda)^{T} \|\omega_{(m-1)T}^{c} - \omega_{(m-1)T}^{*}\|$$

$$+ \eta \|p_{o} - p_{u}\|_{1} (\sum_{i=1}^{T} (1 + \eta\lambda)^{j-1} g(\omega_{mT-j}^{*})).$$

$$(17)$$

According to (10), (16) and (17), we have the boundary of the weight divergence $||\omega_{mT}^f - \omega_{mT}^*||$ as shown in (18).

$$\begin{split} & ||\omega_{mT}^{f} - \omega_{mT}^{*}|| \\ & \leq \frac{1}{K} \sum_{k=1}^{K} \left[(1 + \eta \lambda)^{T} ||\omega_{(m-1)T}^{k} - \omega_{(m-1)T}^{c}|| \right. \\ & + \eta ||p_{I}^{k} - p_{o}||_{1} \left(\sum_{j=2}^{T} g(\omega_{mT-j}^{c}) (1 + \eta \lambda)^{j-1} \right) \right] \\ & + (1 + \eta \lambda)^{T} ||\omega_{(m-1)T}^{c} - \omega_{(m-1)T}^{*}|| \\ & + \eta ||p_{o} - p_{u}||_{1} \left(\sum_{j=1}^{T} g(\omega_{mT-j}^{*}) (1 + \eta \lambda)^{j-1} \right). \end{split}$$
 (18)

Text in blue is newly added.