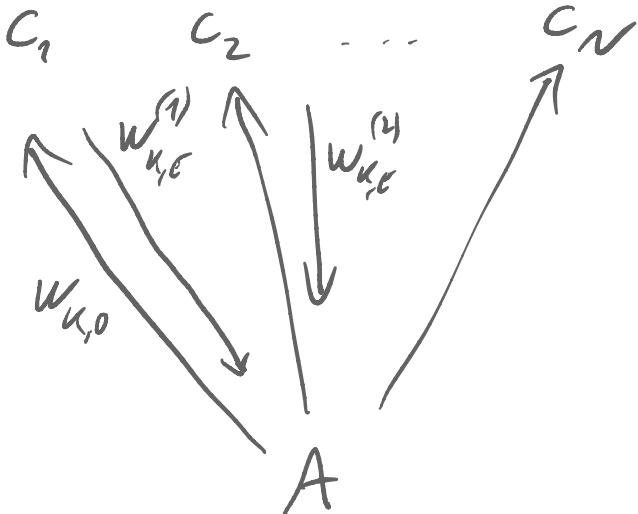


FL & DATA PRIVACY
7/2/23



AGGREGATOR

$$w_{k+1,0} = \sum_{i=1}^N \alpha_i w_{k,E}^{(i)}$$

$$\sum_i \alpha_i = 1$$

EACH CLIENT

$$w_{k,h+1}^{(i)} = w_{k,h}^{(i)} - \gamma_k D.F(w_{k,h}^{(i)})$$

$$\left. \begin{array}{l} \alpha_i = \frac{1}{N} \quad \text{PER-CLIENT FAIRNESS} \\ \alpha_i = \frac{|S_i|}{\sum_j |S_j|} \quad \text{PER-SAMPLE FAIRNESS} \end{array} \right\}$$

CLIENTS ARE NOT ALWAYS AVAILABLE

A_k : SET OF CLIENTS AVAILABLE AT TIME k

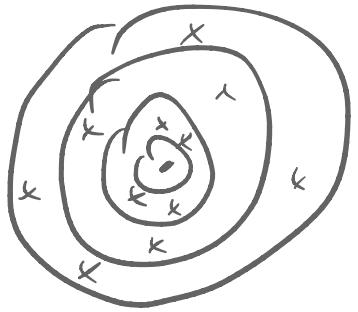
USE ALL CLIENTS IN A_k

$$w_{k+1,0} = \frac{\sum_{i \in A_k} q_i w_{k,0}^{(i)}}{\sum_{i \in A_k} q_i}$$

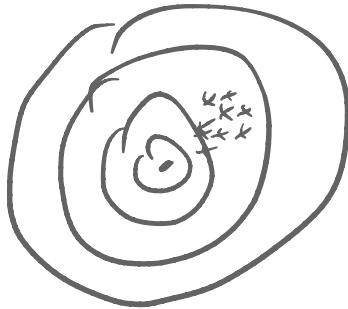
q_i : AGGR. WEIGHTS

WE CONSIDER
THIS
AGGR. SCHEME UNTIL WE SAY SO.

$$w_{k+1,0} = w_{k,0} + \gamma_s \sum_{i \in A_k} q_i (w_{k,0}^{(i)} - w_{k,0}^{(c)})$$
$$\approx -\nabla F^G$$



UNBIASED



BIASED

BUT LESS VARIANCE

α_i : IMPORTANCE WE TARGET FOR
CLIENT i

q_i : AGGREGATION WEIGHT OF CLIENT i

0 1 2 3 4 5 6

C_1 ~~0~~ ~~1~~ - $q_1 = 2$ $\alpha_i = \frac{1}{3}$ $\pi_1 \approx \frac{1}{2}$

C_2 - $q_2 = 6$ $\pi_2 \approx \frac{1}{6}$

C_3 - - - - - $q_3 = 1$ $\pi_3 \approx 1$

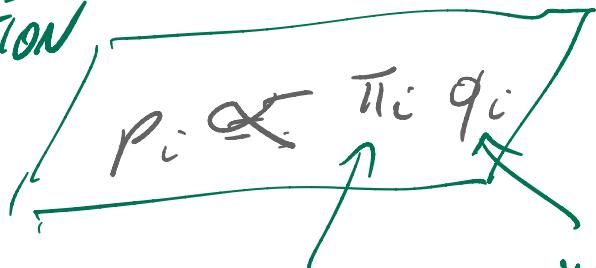
π_i : PROB THAT CLIENT i IS AVAILABLE
AT A GIVEN TIME

$$w_{k+1,0} = w_{k,0} - \eta_s \sum_{i \in \mathcal{K}_k} q_i (w_{k,i}^{(i)} - w_{k,0}^{(i)})$$

↓ WE OPTIMIZE

$$F_B(w) = \sum_i p_i F^{(i)}(w) \neq \sum_i \alpha_i F^{(i)}(w)$$

BIASED
FUNCTION



HOW OFT
THE CLIENT
PARTICIPATES

WHAT INFO
YOU GIVE
TO THE CLIENT
AT THE AGGREGATION

$$p_i = \frac{\pi_i q_i}{\sum_j \pi_j q_j}$$

WHEN ALL CLIENTS ARE AVAILABLE ALL THE TIME

$$\pi_i = 1 \quad \forall i$$

$$p_i \propto \pi_i \cdot q_i = q_i$$

IF WE WANT

$$q_i = ?$$

$$p_i = \alpha_i$$

$$q_i = \alpha_i$$

OR ALL CLIENT ARE EQUALLY AVAILABLE

$$\pi_i = \pi \quad \forall i$$

$$p_i = \frac{\pi q_i}{\sum_j \pi q_j}$$

IF WE TAKE

$$q_i = \alpha_i$$

$$\Rightarrow p_i = \alpha_i$$

$$F_B = F_F$$

\bar{u}_i IN GENERAL DIFFERENT
BUT WE WANT $p_i = \alpha_i$.
(WE WANT TO OPTIMIZE
OUR TARGET OBJECTIVE)
HOW TO PICK q_i ???

$p_i \propto \bar{u}_i \cdot \alpha_i$

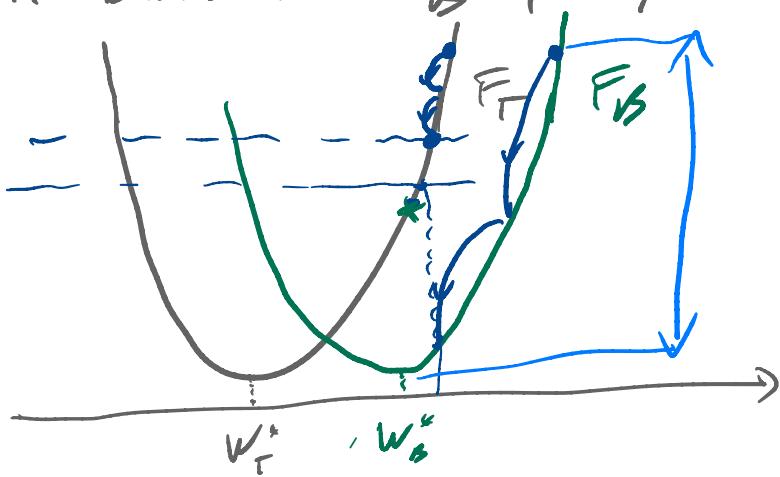
ANSWER

$$q_i = \frac{\alpha_i}{\bar{u}_i} \quad \rightarrow \quad p_i \propto \bar{u}_i \cdot \frac{\alpha_i}{\bar{u}_i} = \alpha_i$$

MAY WE EVER WANT TO OPTIMIZE

A BIASED

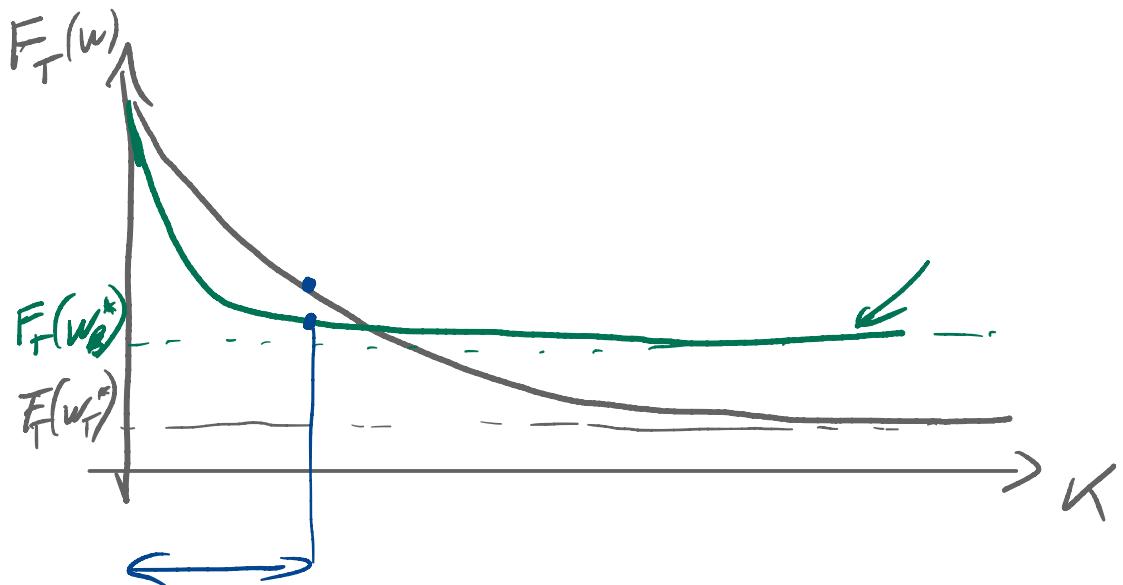
$$F_B \neq F_T$$



$$F_T = \sum_i \alpha_i \cdot F^{(c)}$$

K CORROBORATION
ROUNDS

MAY MAKE SENSE IF OPTIMIZING F_B
IS SOMEWAY EASIER THAN OPTIMIZING F_T
~~AND~~ SO YOU MAY BE BETTER OFF
TARGETTING F_B



IF IT NOT BE EASIER TO MINIMIZE \bar{F}_B

ANNEALED RATIO: FL UNDER HETEROGENEOUS AND CORRELATED CLIENTS

$$F_B(\mathbf{w}) := \sum_{k=1}^N p_k F_k(\mathbf{w}), \text{ with } p_k := \frac{\pi_k q_k}{\sum_{h=1}^N \pi_h q_h}, \quad (4)$$

\vdash

$$F \equiv F_T$$

$\Gamma := \max_{k \in \mathcal{K}} \{F_k(\mathbf{w}^*) - F_k^*\}$.

$F_u = F^{(u)}$

\mathbf{w}^* is optimal

$$F_u^* = \min_{\mathbf{w}} F_u(\mathbf{w})$$

QUANTIFIES THE HETEROGENEITY

Theorem 1 (Decomposing the total error). Under Assumptions 2–4, the optimization error of the target global objective $\epsilon = F(\mathbf{w}) - F^*$ can be bounded as follows:

$$\epsilon \leq \underbrace{2\kappa^2(F_B(\mathbf{w}) - F_B^*)}_{:= \epsilon_{opt}} + \underbrace{2\kappa^4 \chi_{\alpha \parallel p}^2 \Gamma}_{:= \epsilon_{bias}}, \quad (10)$$

where $\kappa := L/\mu$, and $\chi_{\alpha \parallel p}^2 := \sum_{k=1}^N (\alpha_k - p_k)^2/p_k$.

$$q_k = \frac{\alpha_k}{\|\alpha\|_p}$$

$$p_k = \alpha_k, \chi_{\alpha \parallel p}^2 = 0$$

AVAILABILITY

$$F_T(\mathbf{w})$$

$$F_u(\mathbf{w})$$

$F_u = F_u''$ ALL EQUAL

$$F_T = \sum_{\alpha \in \mathcal{K}} F_u = F_K$$

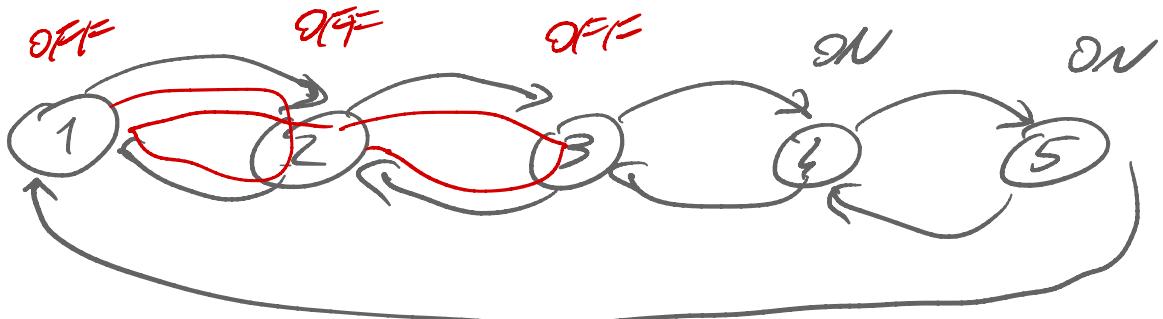
$$\Rightarrow \Gamma = 0$$

Theorem 2 (Convergence of the optimization error ϵ_{opt}). Let Assumptions 1–5 hold and the constants $M, L, D, G, H, \Gamma, \sigma_k, C_P, T_P, \lambda(\mathbf{P})$ be defined as above. Let $Q = \sum_{k \in \mathcal{K}} q_k$. Let the stepsizes satisfy:

$$\sum_t \eta_t = +\infty, \quad \sum_t \ln(t) \cdot \eta_t^2 < +\infty. \quad (11)$$

Let T denote the total communication rounds. For $T \geq T_P$, the expected optimization error can be bounded as follows:

$$\mathbb{E}[F_B(\bar{\mathbf{w}}_{T,0}) - F_B^*] \leq \frac{\frac{1}{2} \mathbf{q}^\top \Sigma \mathbf{q} + v}{(\sum_{t=1}^T \eta_t)} + \psi + \frac{\phi}{\ln(1/\lambda(\mathbf{P}))}, \quad (12)$$



$$\lambda(\mathbf{P}) = \max_u R_2(\mathbf{P}_u)$$

$\frac{1}{\pi^\top \mathbf{q}}$
 minimize $\frac{1}{\pi^\top \mathbf{q}}$
 subject to
 OFF mode available clients

P IS TRANS. MATRIX

$$|\lambda_1(P)| \geq |\lambda_2(P)| \geq \dots \geq |\lambda_n(P)|$$

$$\lambda_1(P) = 1$$



AFFECTS TIME CORRELATION

$|\lambda_2(P)| \approx 1$ HIGH CORRELATED

$|\lambda_2(P)| \rightarrow 0$ NO CORRELATION

WHAT WE LEARN FROM THEOREM 2

IN ORDER TO SPEED UP TIME TRAINING

WE MAY WANT

- 1) TO IGNORE CLIENTS WITH
LOW AVAILABILITY
- 2) " " CLIENTS HIGHLY
CORRELATED
(OVER TIME)

$$w_{k+1,0} = w_{k,0} + \gamma_s \sum_{i \in \mathcal{C}_k} q_i (w_{k,i}^{(i)} - w_{k,0}^{(i)})$$

VARIABLE # OF CLIENTS

QUITE NOISY

$\Rightarrow \gamma_s$ SMALL

1st PROBLEM

2nd : YOU MAY NOT BE ABLE
TO MANAGE ALL
AVAILABLE CLIENTS

GOOGLE ALGO (FedAvg)

FOR $k = 1$ TO K

A_k : CLIENTS AVAILABLE

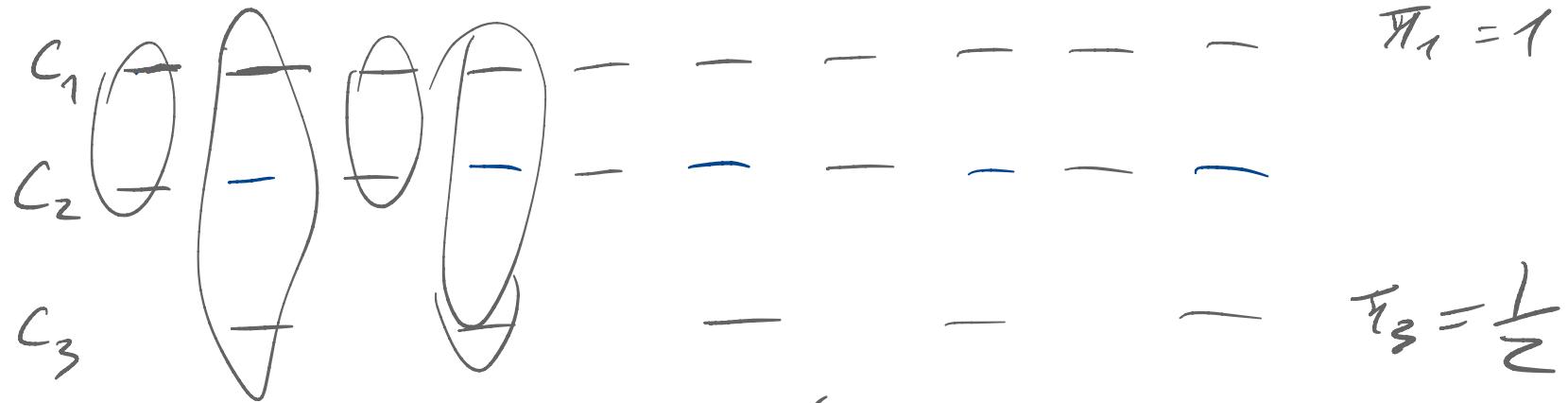
PICK N CLIENTS UNIFORMLY FROM A_k
(SAMPLING) $B_k \subset A_k$

SEND $w_{k,0}$ TO THE N CLIENTS

$$w_{k+1,0} = \frac{1}{N} \sum_{i \in B_k} q_i w_{k,i}^{(c)} \quad \checkmark$$

$$\left(= w_{k,0} + \gamma_s \frac{1}{N} \sum_{i \in B_k} q_i (w_{k,i}^{(c)} - w_{k,0}^{(c)}) \right)$$

STILL PRONE TO THE BIKS
DUE TO HETEROGEN. CLIENT AVAILAB.



IF $N=2$ $\text{Prds} (1 \in B_d) = \bar{H}_1 = 1$

$$\begin{aligned}\text{Prds} (1 \in B_d) &= \frac{1}{2} \cdot 1 + \frac{1}{2} \cdot \frac{2}{3} = \\ &= \frac{5}{6} < 1\end{aligned}$$

MUCH MORE DIFFICULT TO STUDY
HOW TO ACCOUNT THE EFFECT
OF HERO G. AVAILAB. ?

IN CONVERG. OF FedAvg UNDER NOVEL
M CLIENTS & STRAT.

- 1) , EVERY CLIENT HAS
PROB. $\frac{N}{M}$ TO BE SELECTED AT ANY SLOT
AND $q_{ik} = \alpha_i$
- 2) EVERY CLIENT IS SELECTED W. P. α_i
AND THE $q_i = \frac{1}{M}$

RECENT PAPER FROM

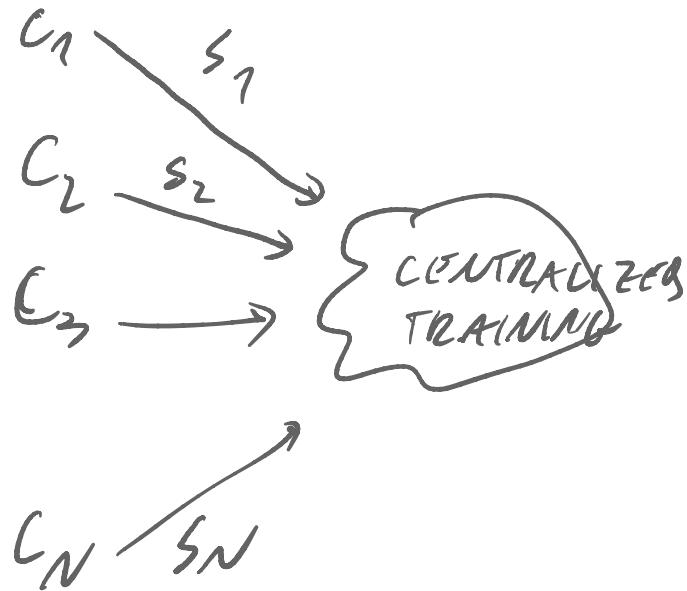
NIBERO, DE VECIANA, ?.

PRACTICAL NUMBERS

- $N = 100 - 500$
- TRAINING DURATION: WEEKS
- 1 CORR. ROUND : 1h

$$20 \times 7 \times 2 \times 100 \approx 30K \text{ CHECKS INVOLVED}$$

COMMUNICATION COST REDUCED?



COMM. COST

$$\text{IS } |S_1| + |S_2| + \dots + |S_N|$$



COMM. COST

$$\sum_{k=1}^K |\mathcal{B}_k| \cdot \text{SIZE model}$$
$$= K N \cdot \text{SIZE model}$$

REDUCE COMMUNICATION COST

COMPRESS

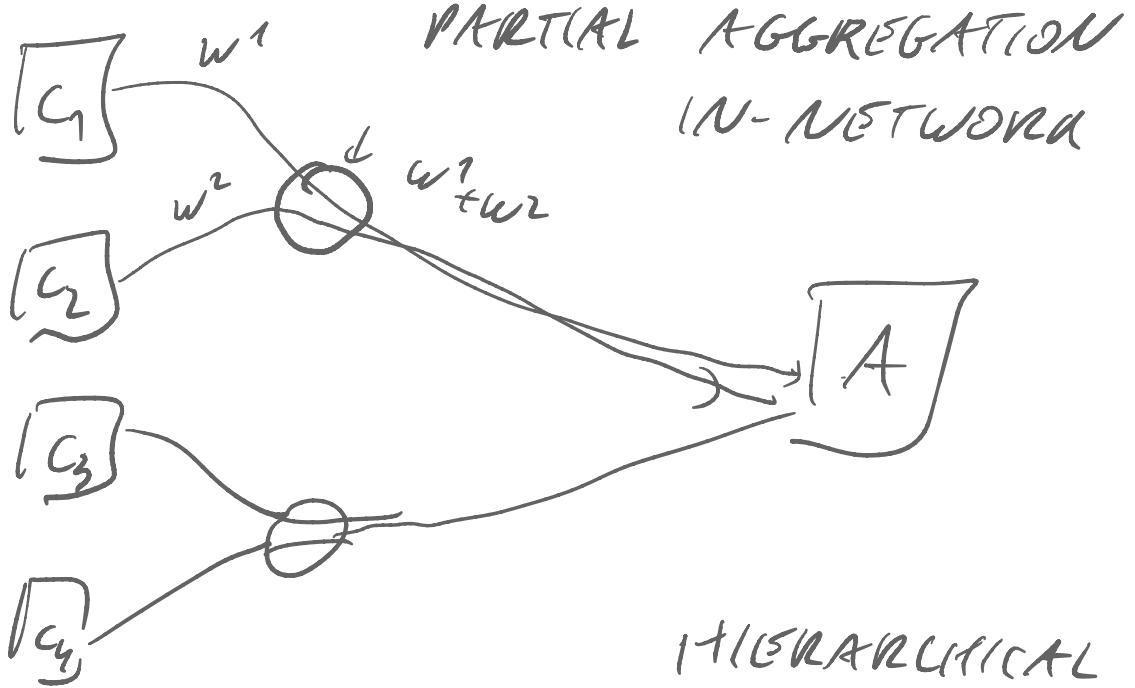
→ ENCODE NUMBERS WITH
LESS STATES

TOP - K

NOT NECESSARILY TO $w_{k,\epsilon}^{(c)}$

BUT

$w_{k,\epsilon}^{(c)} - w_{k,0}$



HIERARCHICAL FL
LARGE SENSOR
NETWORK

The diagram shows a hierarchical structure for Federated Learning (FL) in a large sensor network. It consists of two main components:

- Bottom Level:** Two small sensor nodes are shown. One node has a single line connecting to a central node labeled "FL". The other node has multiple lines connecting to the same central node.
- Central Node:** The central node is labeled "FL" with a plus sign (+), indicating it performs aggregation or combines data from multiple sources.