

# Federated Multi-Task Learning

## General Multi-Task Learning Setup

- General formulation :

$$\min_{W, \Omega} \left[ \sum_{k=1}^K \sum_{i \in \mathcal{P}_k} f_k(w_k^\top x_k^i, y_k^i) + \mathcal{R}(W, \Omega) \right]$$

$\Omega \in \mathbb{R}^{K \times K}$  models *relationships among tasks*.  $f_k$  is a convex loss function.

- MTL problems differ based on their assumptions on  $\mathcal{R}$

$$\mathcal{R}(W, \Omega) = \lambda_1 \operatorname{tr}(W \Omega W^\top) + \lambda_2 \|W\|_F^2$$

# Federated Multi-Task Learning

## MOCHA : A Framework for Federated Multi-Task Learning

$$\min_{W, \Omega} \left[ \sum_{k=1}^K \sum_{i \in \mathcal{P}_k} f_k(w_k^\top x_k^i, y_k^i) + \mathcal{R}(W, \Omega) \right]$$
$$\mathcal{R}(W, \Omega) = \lambda_1 \text{tr}(W\Omega W^\top) + \lambda_2 \|W\|_F^2$$

- ✓ Not jointly convex in  $W$  and  $\Omega$
- ✓ When fixing  $\Omega$ , updating  $W$  depends on the data  $X$ .
- ✓ When fixing  $W$ , optimizing for  $\Omega$  only depends on  $W$ , not on  $X$ .