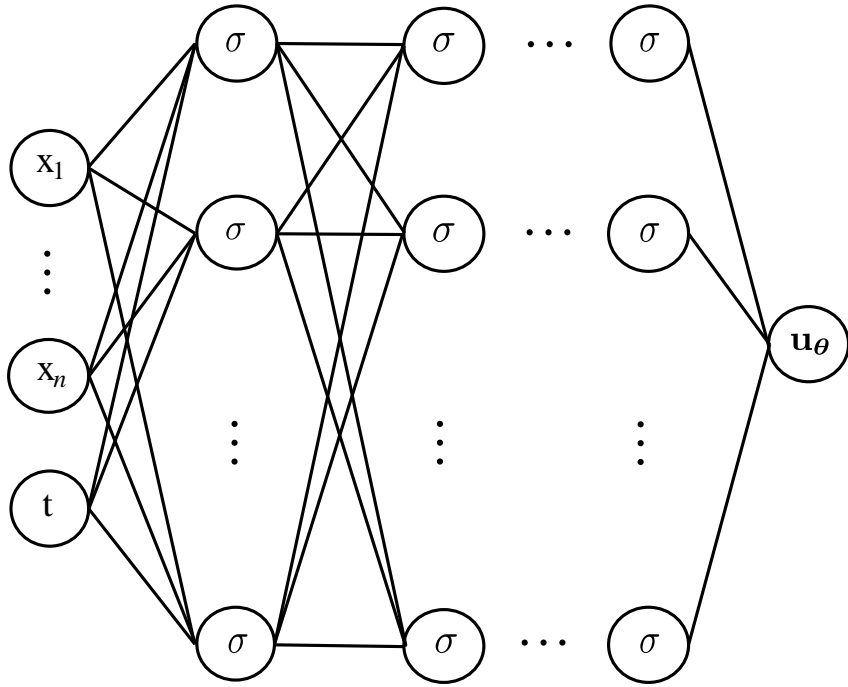


Pre-training phase



Data loss:

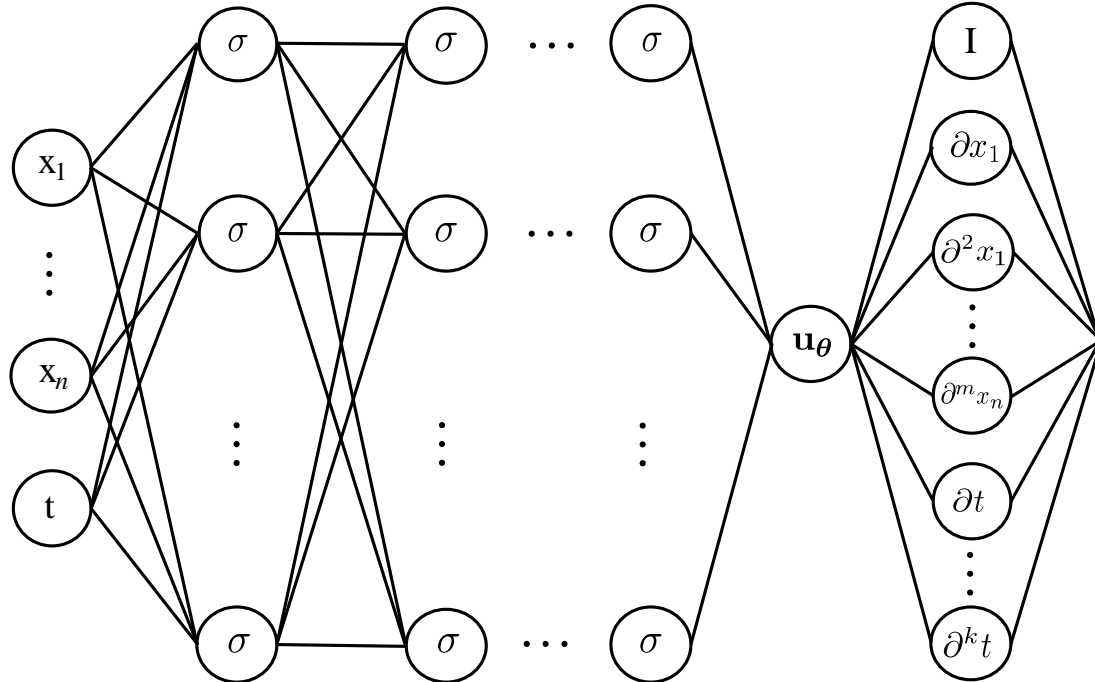
$$\mathcal{L}_d(\theta) = \frac{1}{N_d} \sum_{k=1}^{N_d} (u_{\theta}(\mathbf{x}_d^k, t_d^k) - u(\mathbf{x}_d^k, t_d^k))^2$$

Minimize:

$$\theta_d = \arg \min_{\theta} \mathcal{L}_d(\theta)$$



Fine-tuning phase



PDE residual loss:

$$\mathcal{L}_r(\Theta) = \frac{1}{N_r} \sum_{k=1}^{N_r} (\mathcal{F}[u_{\theta}(\mathbf{x}_r^k, t_r^k); \gamma] - s(\mathbf{x}_r^k, t_r^k))^2$$

Initial condition loss:

$$\mathcal{L}_i(\theta) = \frac{1}{N_i} \sum_{k=1}^{N_i} (\mathcal{I}[u_{\theta}(\mathbf{x}_i^k, t_i^k)] - f(\mathbf{x}_i^k))^2$$

Boundary condition loss:

$$\mathcal{L}_b(\theta) = \frac{1}{N_b} \sum_{k=1}^{N_b} (\mathcal{B}[u_{\theta}(\mathbf{x}_b^k, t_b^k)] - q(\mathbf{x}_b^k, t_b^k))^2$$

Data loss:

$$\mathcal{L}_d(\theta) = \frac{1}{N_d} \sum_{k=1}^{N_d} (u_{\theta}(\mathbf{x}_d^k, t_d^k) - u(\mathbf{x}_d^k, t_d^k))^2$$

Minimize:

$$\Theta^* = \arg \min_{\Theta} [\lambda_r \mathcal{L}_r(\Theta) + \lambda_i \mathcal{L}_i(\theta) + \lambda_b \mathcal{L}_b(\theta) + \lambda_d \mathcal{L}_d(\theta)]$$

subject to $\theta_0 = \theta_d$