

Ncd without forgetting

Formulation:

labeled data $\mathcal{D}_{lab} = \{(x_i, y_i) \sim \mathcal{P}(x, y_{lab})\}$

model ψ

unlabeled $\mathcal{D}_{unlab} = \{(x_i) \sim \mathcal{P}(x | \mathcal{Y}_u)\}$

$$\mathcal{Y}_l \cap \mathcal{Y}_u = \emptyset$$

feature extractor: ϕ_{FE}

labeled head ϕ_{LAB}

unlabeled head ϕ_{unlab}

Constraint: Labeled data can't be accessed
during VCD time

utilize feature Distillation loss L_{FD}

$$L_{FD} = \|\phi_{FE}(u) - \phi_{FE}^{lab}(x)\|_2^2$$

$\underbrace{\hspace{10em}}$
 trained on lab (frozen
 layer of model)
 model regularizer.

① retain performance in L:

- Generate pseudo-latents.

$\left(\text{cross} / \text{for labeled training data} \right)$
 that maximally activate c class.

step 1 \leftarrow sample z_i

Latent inversion: frozen.

$$\Rightarrow p_i = \phi_{\text{Lab}}(z_i)$$

$$\Rightarrow z_{i+1} = z_i + \nabla p_i[c]$$

corresponding class mean of labeled

training instances z_μ^c

Not Sure Why

Return Dictionary for pseudo-latent for the
labeled classes D_{pseudo}

① Enhancing VCD

→ Sinkhorn-Knopp clustering:

→ for $x \in D_{\text{unlab}}$

$$l = \phi_L(\phi_{f_E}(x))$$

$$u = \phi_U(\phi_{f_E}(x))$$

$$I(l; u) = H(l) - H(l|u)$$

$$= -\mathbb{E}_l[\log p(l)] + \mathbb{E}_{l,u}[\log p(l|u)]$$

$$\geq -E_x[\log p(l)] + E_{x,u} \log q(l|u) \\ + E_u \left[K_L [p(l|u) \| q(l|u)] \right]$$

$$J(x;u) \geq -E_x[\log p(l)] + E_{x,u}[\log q(l|u)]$$

$$= -E_x[\log p(l)] + E_{x,u} \left[\sum_{i=1}^n \log \sigma_w^i \right. \\ \left. - L_{ni} \left[\frac{(x_i - \mu_\theta(u))^2}{2(\sigma_w^i)^2} \right] \right]$$

[gaussian simplification]