

# **Paper: The details matter: preventing class collapse in supcon**

Notations:

labeled input data:  $\mathcal{D} = \{(x_i, y_i)\}_{i=1}^N$

$$(x, y) \in \mathcal{D}$$

$$x \in \mathcal{X} \in \mathbb{R}^{\text{data dim.}}$$

$$y \in \mathcal{Y} = \{1, 2, \dots, K\}$$

$$\text{data label } h(x) \in \mathcal{Y} \quad // \quad p(y|x)$$

$$p(y=i) = 1/K$$

Target: to learn a model  $\hat{p}(y|x)$

Data point belongs to categories beyond labels

STRATA

strata as latent labels  $z \in \mathcal{Z} = \{1, 2, \dots, C\}$

$\mathcal{Z}$  divided into disjoint subset:  $S_1, S_2, \dots, S_K$

$$z \in S_k, \quad y = k$$

$S(c)$  denotes deterministic label  $c$ .

1st state is sampled  $\mathbf{z}$   
 $\mathbf{x}$  is sampled  $p_z = p(\cdot | \mathbf{z})$

label is  $y = S(\mathbf{z})$

## SupCon and collapse embedding

similarity

$$\sigma(x, x') = \frac{f^T(x) f(x')}{c}$$

$B \rightarrow$  set of batches of labeled dataset on  $\mathcal{D}$

positive

$$P(i, B) = \{p \in B \setminus i : h(p) = h(i)\}$$

SupCon loss:

$$\hat{L}_{sc}(f, x_i, B) = \frac{1}{|P(i, B)|} \sum_{p \in P(i, B)} \log \frac{\exp(\sigma(x_i, x_p))}{\sum_{a \in B \setminus i} \exp(\sigma(x_i, x_a))}$$

positive pair.      negative pair.

class collapse: simplex embedding scenario.

if  $h(x) = i$  then  $f(x) = v_i \quad \forall x \in B$

$\{v_i\}_{i=1}^K \rightarrow$  forms regular simplex.

with

properties  
1. set

i)  $\sum_{i=1}^K v_i = 0$

ii)  $\|v_i\|_2 = 1$

or  
in collapse

$$(iii) \exists c_k \in \mathbb{R} \text{ s.t. } v_i^T y = c_k \text{ for } i \neq j$$

End model

Linear Classifier:  $w \in \mathbb{R}^{k \times d}$

$$\|w_y\|_2 \leq 1 \quad ; \quad y \in \mathcal{Y}$$

Empirical loss:

$$\hat{\mathcal{L}}(w, \mathcal{D}) = \sum_{x_i \in \mathcal{D}} -\log \frac{\exp(f(x_i)^T w_{h(x)})}{\sum_{j=1}^k \exp(f(x_i)^T w_j)}$$

Prediction:

$$\hat{p}(y|x) = \hat{p}(y|f(x))$$

generalized error:

$$\mathcal{L}(x, y, f) = \mathbb{E}_{x, y} \left[ -\log \hat{p}(y|f(x)) \right]$$

**Methodologies**

1. class collapse minimize  $L(x, z, f) \forall x$

losses stratified:

- i)  $P(y = h(x) | x) = 1$
- ii)  $P(z|x) = \frac{1}{n} ; z \in S_{h(x)}$
- iii)  $P(x|z) = P(x|y)$  // no strata distinction

Modified Contrastive loss

$$L_{spread} = \alpha L_{attnadv} + (1 - \alpha) L_{recpl}.$$

negative examples:  $N(i, \theta) = \{a \in \mathcal{A} : h(a) = h(i)\}$

$$\hat{L}_{att} = (f, x_i, \theta) = \frac{-1}{|P(i, \theta)|} \sum_{p \in P(i, \theta)} \log \frac{\exp(\sigma(x_i, x_p))}{\exp(\sigma(x_i, x_p)) + \sum_{a \in N(i, \theta)} \exp(\sigma(x_i, x_a))}$$

$$\hat{L}_{recp} = -\log \frac{\exp(\sigma(x_i, x_i^{aug}))}{\sum_{p \in P(i, \theta)} \exp(\sigma(x_i, x_p))} \quad \text{[typical solution]}$$

spread the positive around.