

#### **Models under Distribution Shifts**

Recently, a linear trend between the ID and OOD accuracy of models has been observed. (Baek et al, 2022) found that OOD vs. ID agreement also forms a line, and it matches that of the accuracy.



In this paper, we study this phenomenon under a simple theoretical setting.

## **Theoretical Setting**

**Data generation:** We assume that  $\beta \sim \mathcal{N}(0, I_d)$ 

 $x_i \stackrel{\text{i.i.d.}}{\sim} \mathcal{N}(0, \Sigma_s), y_i = \frac{1}{\sqrt{d}} \beta^\top x_i + \epsilon_i, \epsilon_i \stackrel{\text{i.i.d.}}{\sim} \mathcal{N}(0, \sigma_\epsilon^2)$ 

The input distribution *shifts* to the target distribution  $x \sim \mathcal{N}(0, \Sigma_{t})$  at test time.

Random features model: Two-layer neural networks with fixed, randomly generated weights in the first layer  $f_{W,a}(x) = \frac{1}{\sqrt{N}} a^{\top} \sigma \left( W x / \sqrt{d} \right)$ .

**Ridge regression:** Parameters  $a \in \mathbb{R}^N$  are fit via ridge regression with data  $X = (x_1, \ldots, x_n) \in \mathbb{R}^{d \times n}$  and  $Y = (y_1, \ldots, y_n)^\top \in \mathbb{R}^n.$ 

## Conditions

**Proportional limit:** We assume that  $n, d, N \rightarrow \infty$ with  $d/n \to \phi > 0$  and  $d/N \to \psi > 0$ .

**Spectral property:**  $\Sigma_{s} \rightsquigarrow (\lambda_{1}^{s}, v_{1}), \dots, (\lambda_{d}^{s}, v_{d}).$ Define  $\lambda_i^{t} = v_i^{\top} \Sigma_t v_i$  for  $i \in [d]$ . We assume that

$$\frac{1}{d} \sum_{i=1}^{d} \delta_{(\lambda_i^{\mathrm{s}}, \lambda_i^{\mathrm{t}})} \to \mu.$$

# **DEMYSTIFYING DISAGREEMENT-ON-THE-LINE** IN HIGH DIMENSIONS

## Donghwan Lee<sup>\*</sup>, Behrad Moniri<sup>\*</sup>, Xinmeng Huang, Edgar Dobriban, Hamed Hassani University of Pennsylvania



We obtain a simpler expression by taking the ridgeless limit  $\gamma \rightarrow 0$ . The self-consistent equation simplifies to

$$\kappa = \frac{\min(1, \phi/\psi)}{\omega_{s} + \mathcal{I}_{1,1}^{s}(\kappa)}.$$

[See Theorem 3.1 and Corollary 3.2 for details]

#### Disagreement-on-the-Line

Recently, (Tripuraneni et al., 2021) proved that under covariate shift, in the ridgeless, and overparameter**ized** regime  $\phi > \psi$  we have

 $\lim_{\gamma \to 0} \operatorname{Risk}^{t}(\phi, \psi, \gamma) = a \lim_{\gamma \to 0} \operatorname{Risk}^{s}(\phi, \psi, \gamma) + b_{\operatorname{risk}},$ where *a* and  $b_{risk}$  are independent of  $\psi$ .

**Theorem.** In the ridgeless, and overparameterized regime  $\phi > \psi$  and for  $i \in \{I, SS\}$ ,

 $\lim_{\gamma \to 0} \operatorname{Dis}_{i}^{t}(\phi, \psi, \gamma) = a \lim_{\gamma \to 0} \operatorname{Dis}_{i}^{s}(\phi, \psi, \gamma) + b_{i},$ where the slopes and intercept are independent of  $\psi$ .

[See Theorem 4.1 for details]

#### **Approximate Linear Relation**

**Theorem** (Approximate linear relation). Given  $\phi >$  $\psi$ , deviation from the line, for I and SS disagreement, is bounded by

$$\begin{split} |\mathrm{Dis}_{\mathrm{I}}^{\mathrm{t}}(\phi,\psi,\gamma)-a\mathrm{Dis}_{\mathrm{I}}^{\mathrm{s}}(\phi,\psi,\gamma)-b_{\mathrm{I}}| \\ &\leq \frac{C(\gamma+\sqrt{\psi\gamma}+\psi\gamma+\gamma\sqrt{\psi\gamma})}{(1-\psi/\phi+\sqrt{\psi\gamma})^{2}}, \\ |\mathrm{Dis}_{\mathrm{SS}}^{\mathrm{t}}(\phi,\psi,\gamma)-a\mathrm{Dis}_{\mathrm{SS}}^{\mathrm{s}}(\phi,\psi,\gamma)| \\ &\leq \frac{C(\sqrt{\psi\gamma}+\psi\gamma+\gamma\sqrt{\psi\gamma})}{(1-\psi/\phi+\sqrt{\psi\gamma})^{2}} \\ &\text{where } C > 0 \text{ depends on } \phi, \mu, \sigma_{\epsilon}^{2}, \text{ and } \sigma. \end{split}$$

## **Disagreement-on-the-Line in different regimes**

(a) **underparameterized models** 





# **Asymptotics vs. Simulations** I disagr. — SS disagr. **—** SW disagr. $\phi/\psi = N/n$ $[\gamma = 0.01, \sigma_{\epsilon}^2 = 0.25, \text{ and } \mu = 0.5\delta_{(1.5,5)} + 0.5\delta_{(1,1)}, \phi = 0.5]$

## **Real World Experiments**

Similar results hold for real-world datasets where the Gaussianity and linearity are violated.

#### **Experiments of (a) CIFAR and (c) Camelyon17**



(b) ridge regularization

