

Perfectly Balanced: Improving Transfer and Robustness of Supervised Contrastive Learning

Dan Fu*, Mayee Chen*, Avanika Narayan, Michael Zhang, Zhao Song, Kayvon Fatahalian, Christopher Ré.

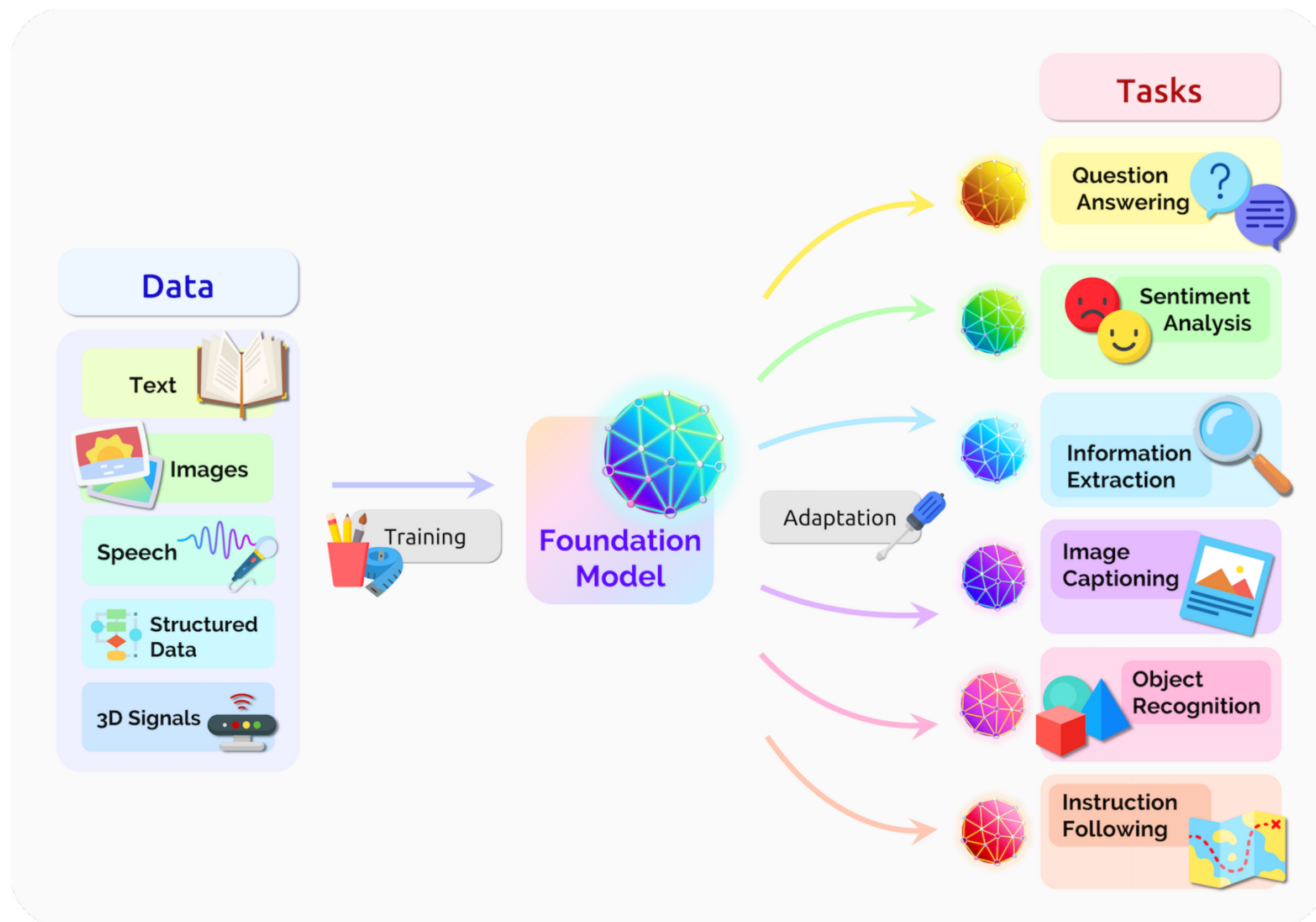
ICML 2022

Mayee F. Chen*, Daniel Y. Fu*, Avanika Narayan, Michael Zhang, Zhao Song, Kayvon Fatahalian, Christopher Ré.
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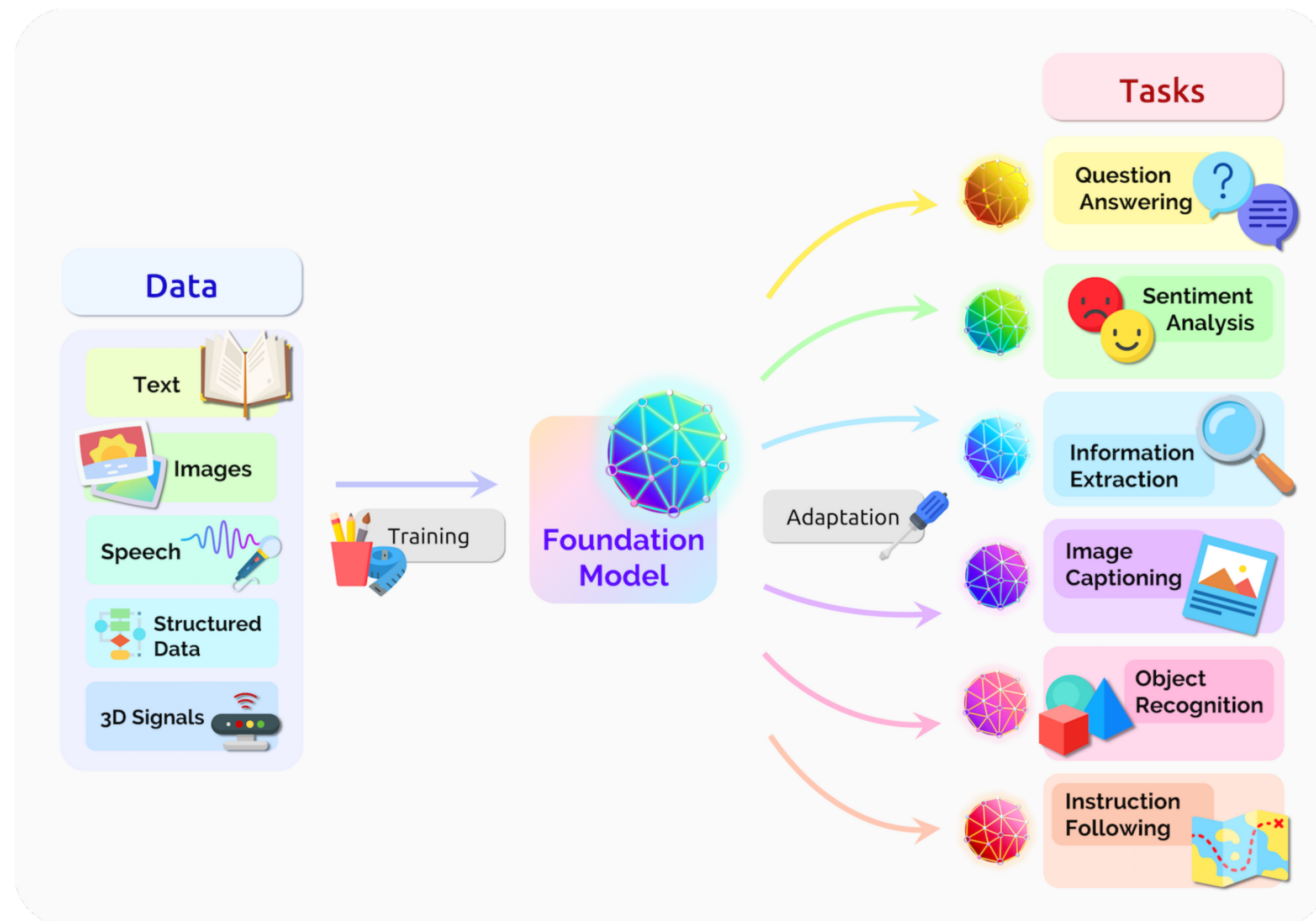
Setting the Stage...

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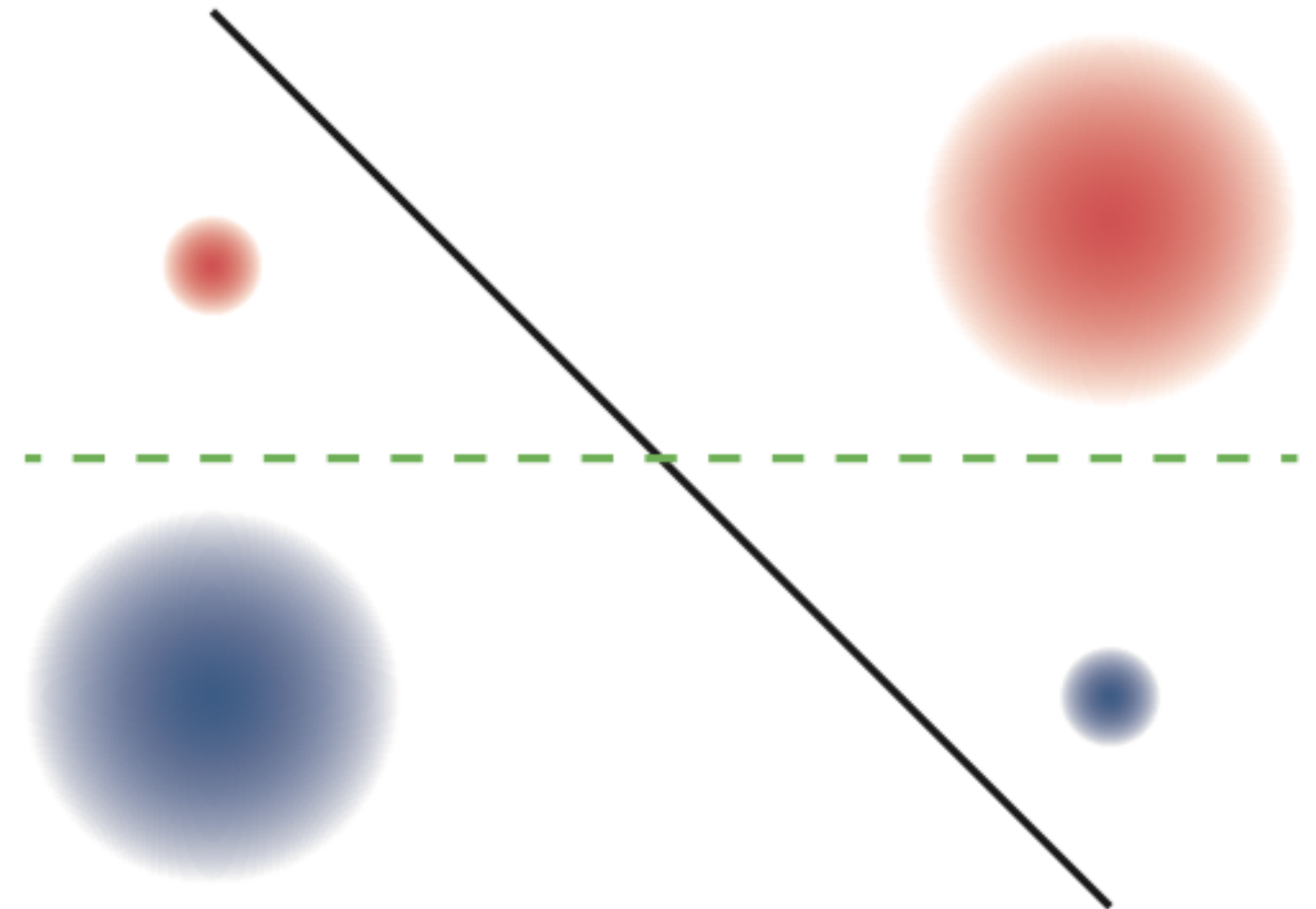


Transfer

Setting the Stage...

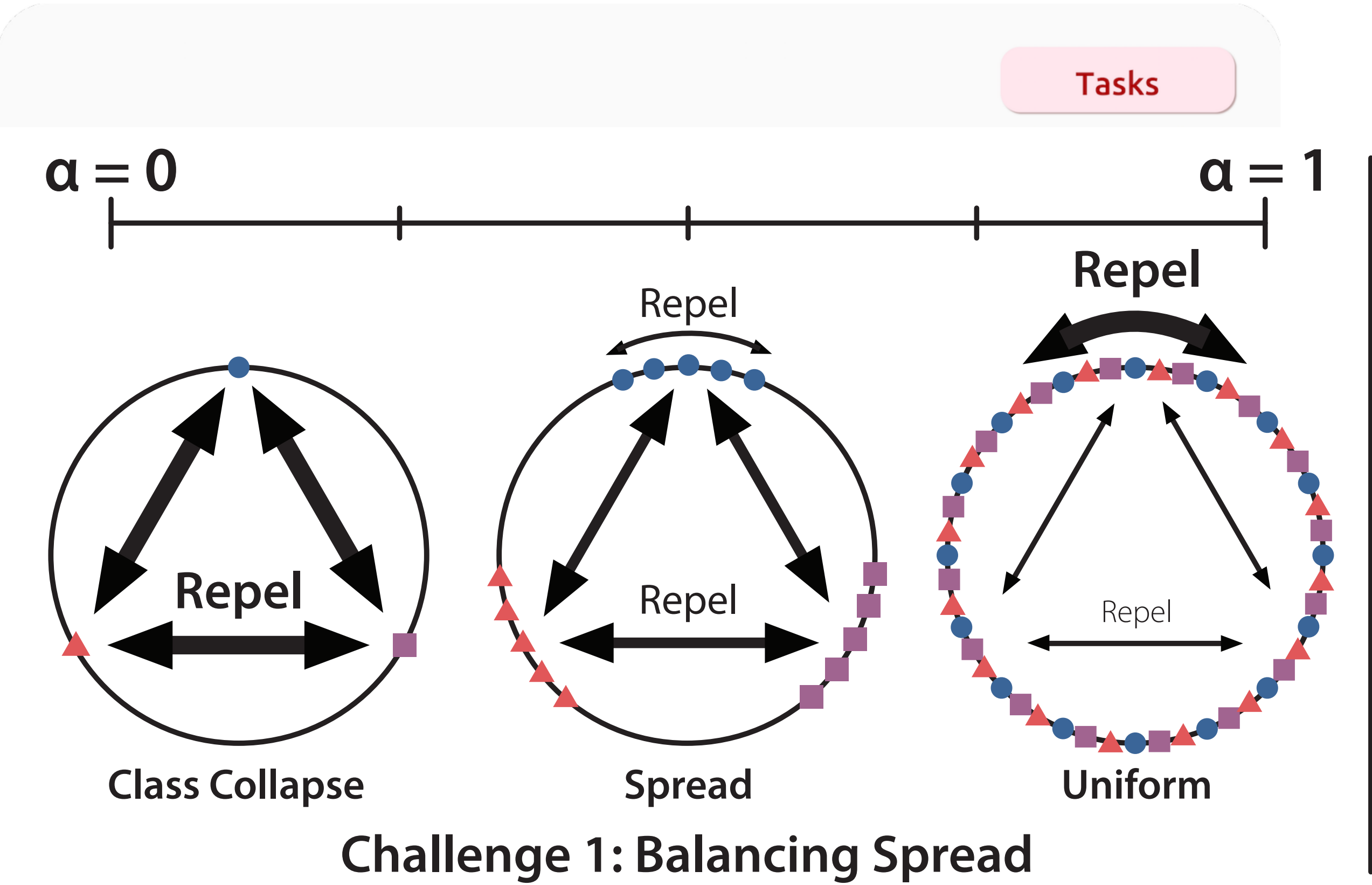


Transfer

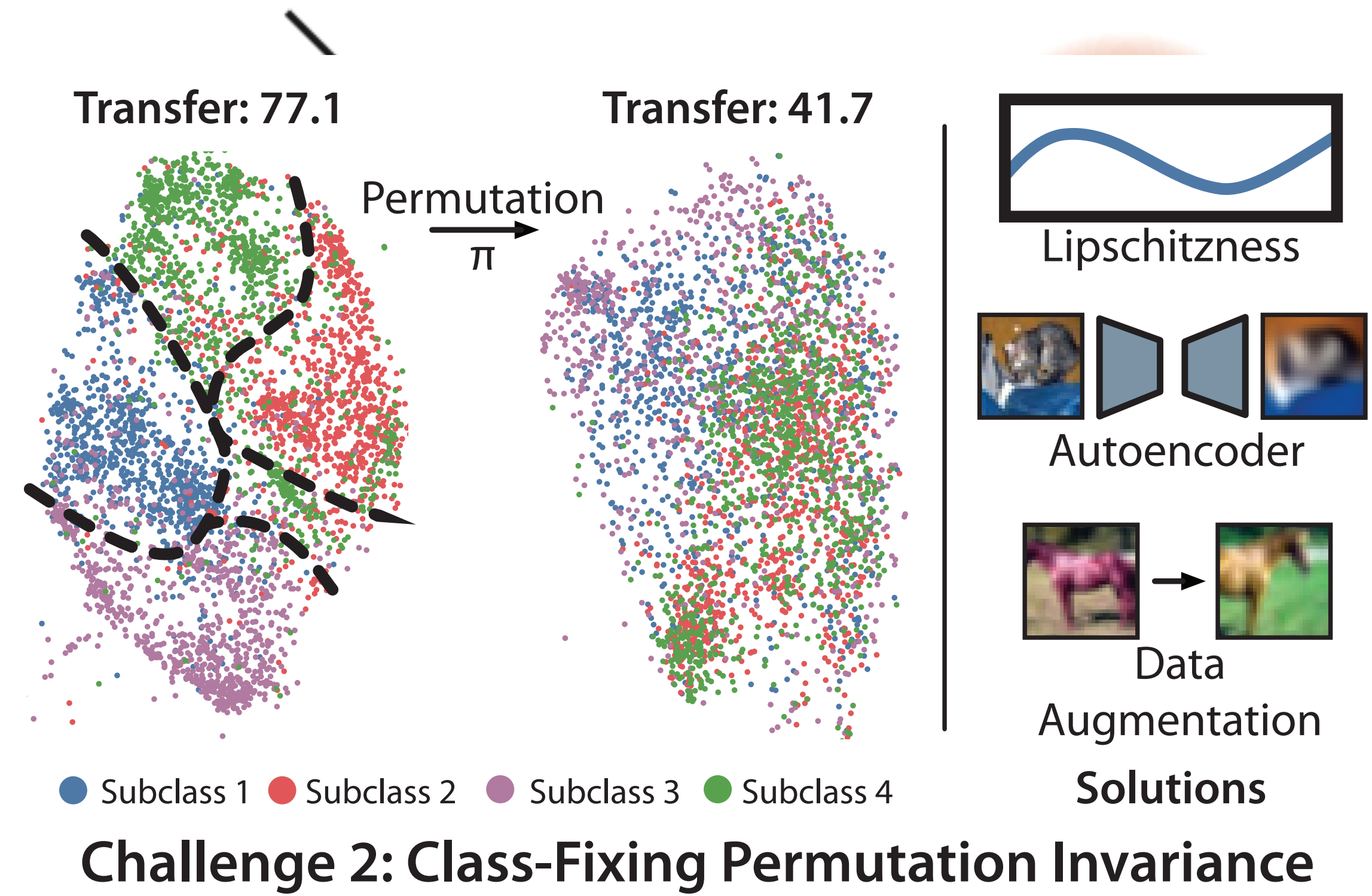


Robustness

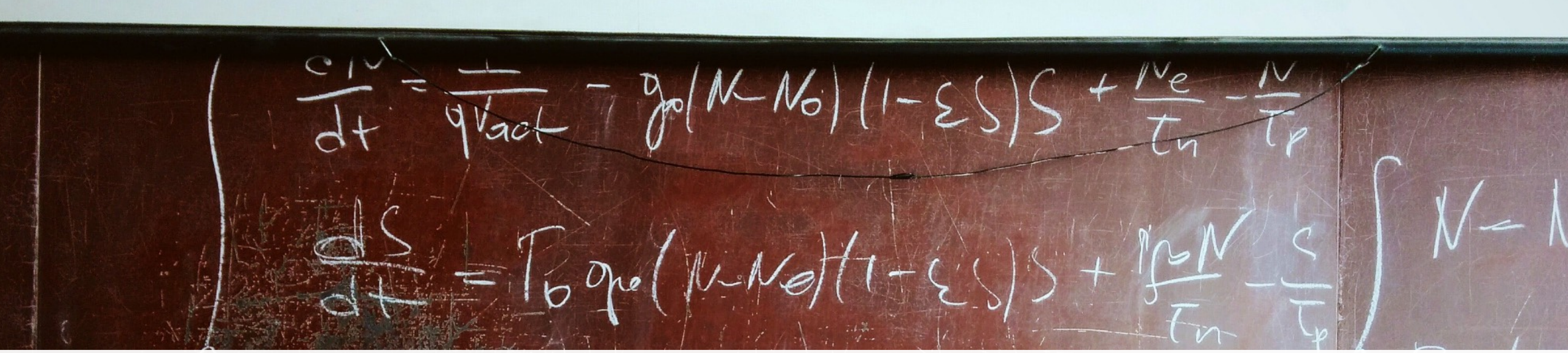
Setting the Stage...



Transfer



Robustness



The chalkboard contains two differential equations. The top equation is $\frac{dN}{dt} = \frac{1}{qV_{act}} - \gamma_0(N-N_0)(1-\epsilon S)S + \frac{N_e}{T_n} - \frac{N}{T_p}$. The bottom equation is $\frac{dS}{dt} = T_0 \gamma_0(N-N_0)(1-\epsilon S)S + \frac{\gamma_0 N}{T_n} - \frac{S}{T_p}$. To the right of these equations, there is a partial equation $N = N_0 + \dots$.

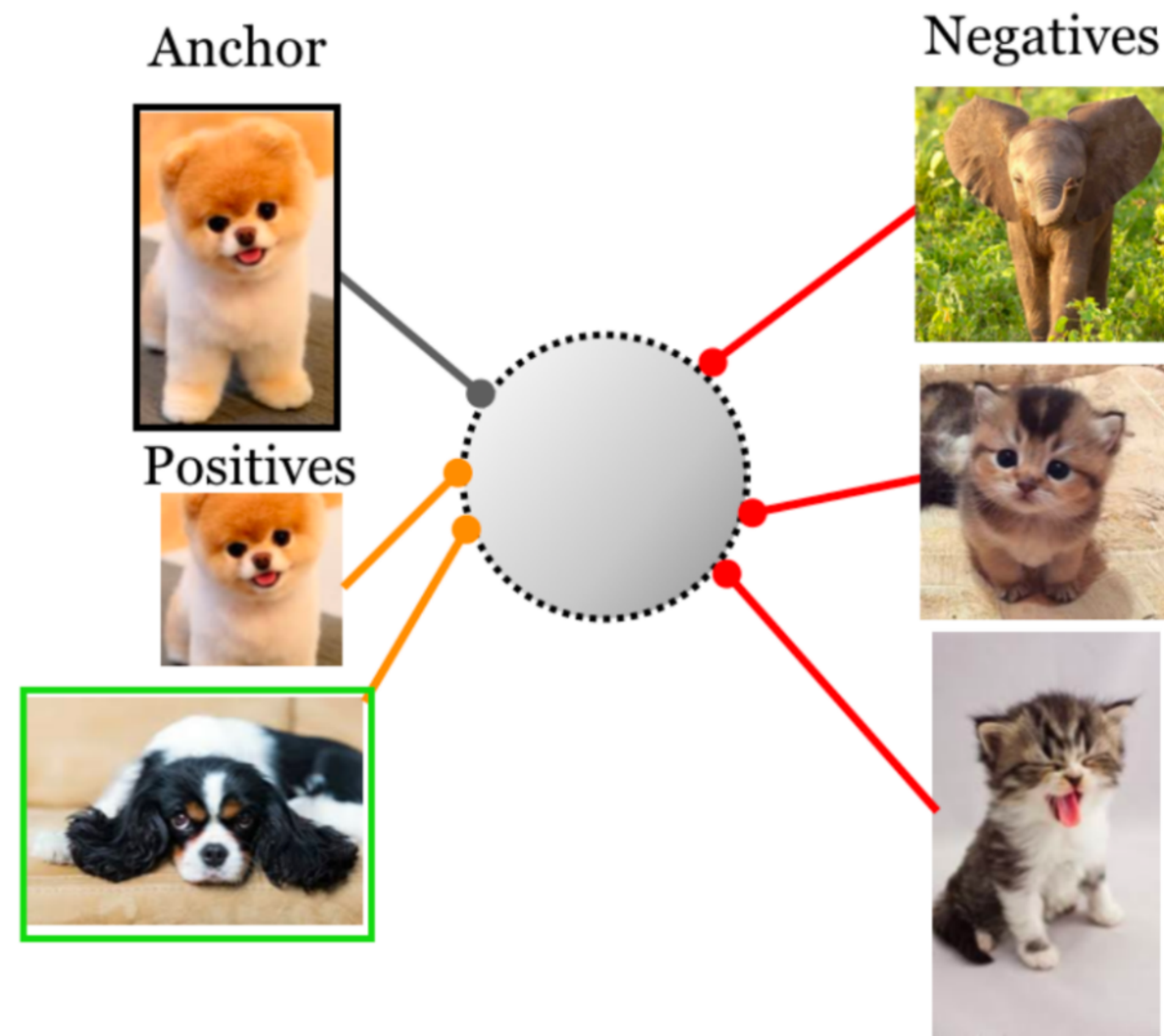
Background

Challenge 1: Balancing Spread

Challenge 2: Subclass Clustering

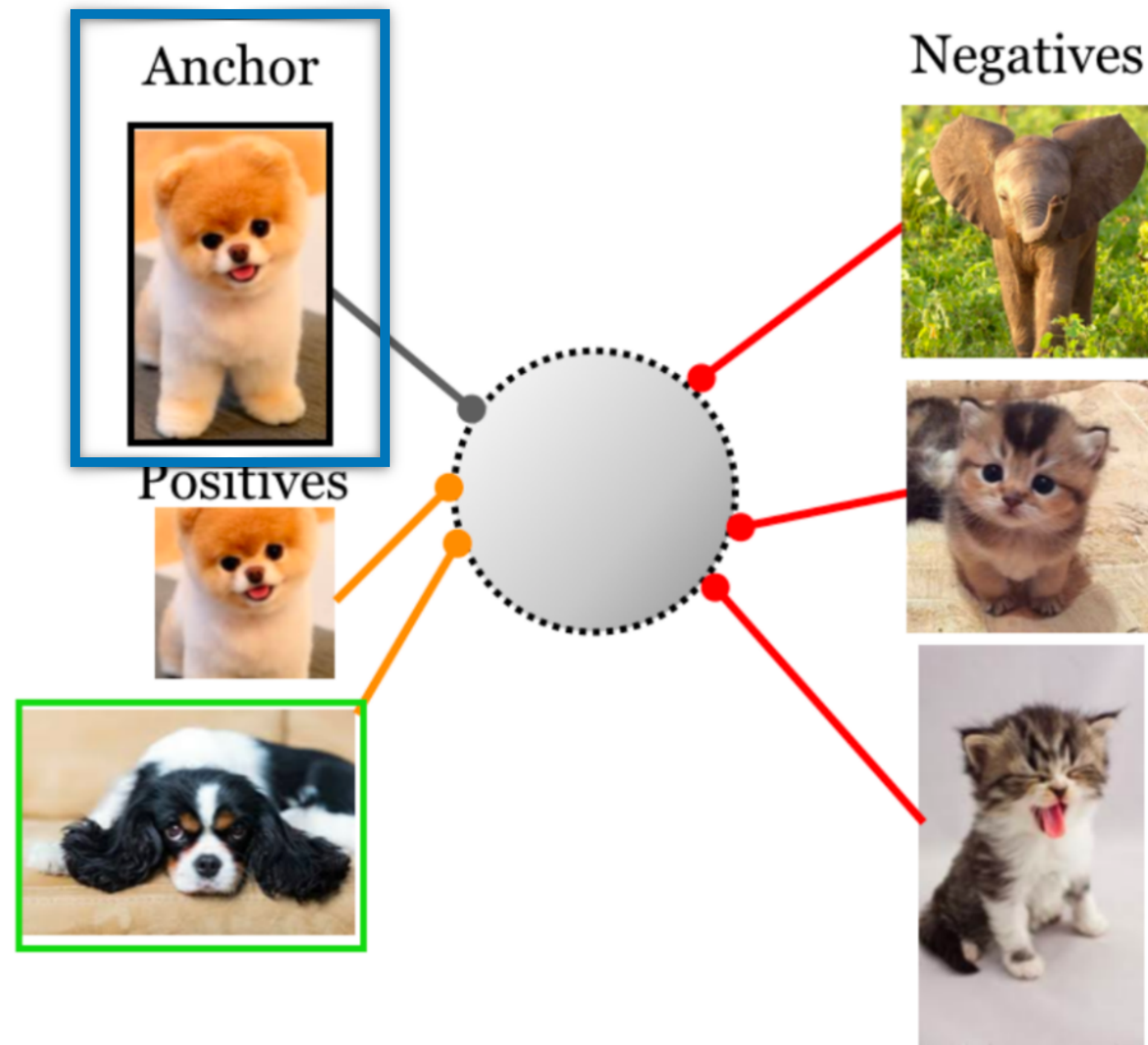
Experiments

Supervised Contrastive Learning



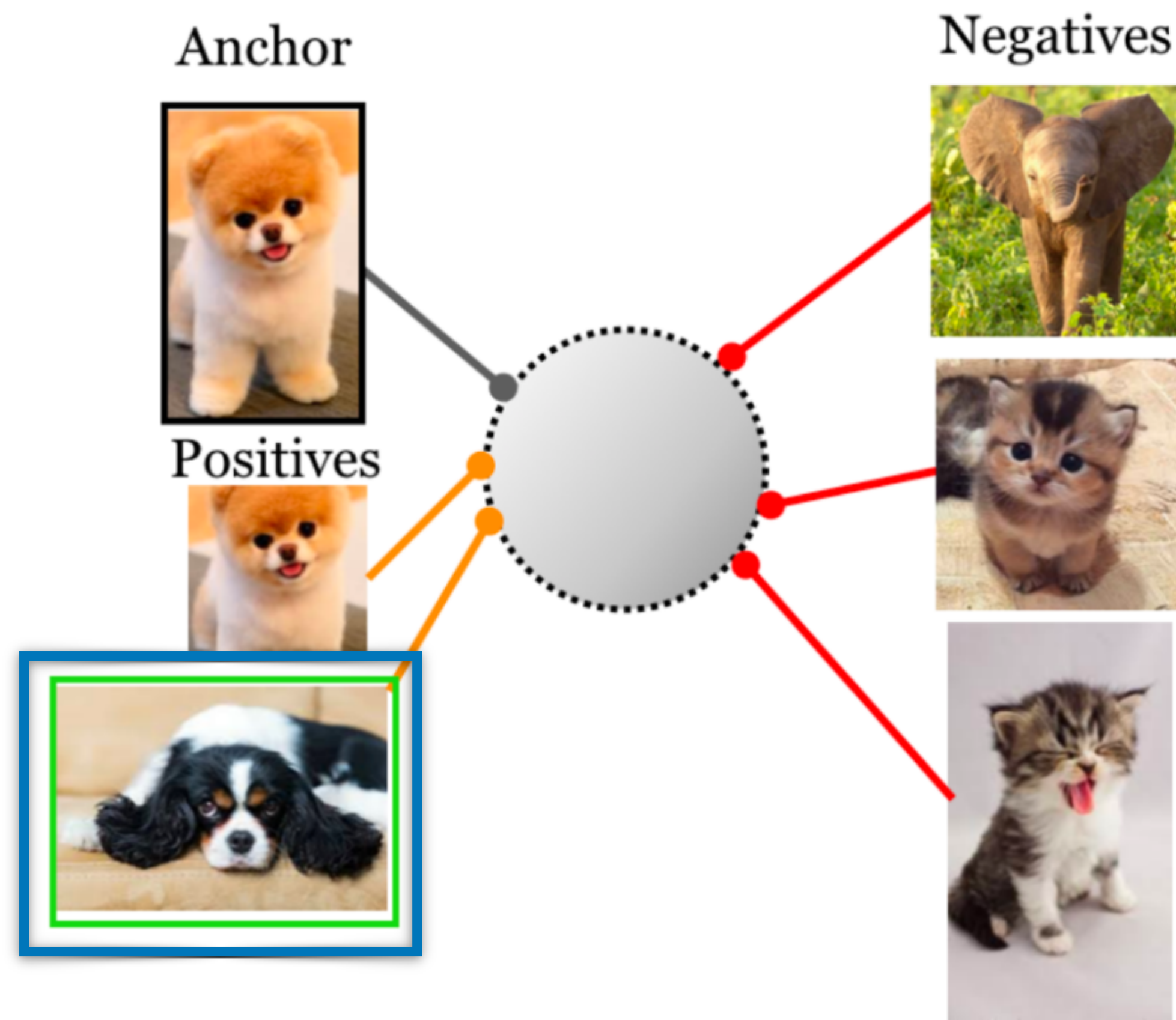
Supervised Contrastive (SupCon)

Supervised Contrastive Learning



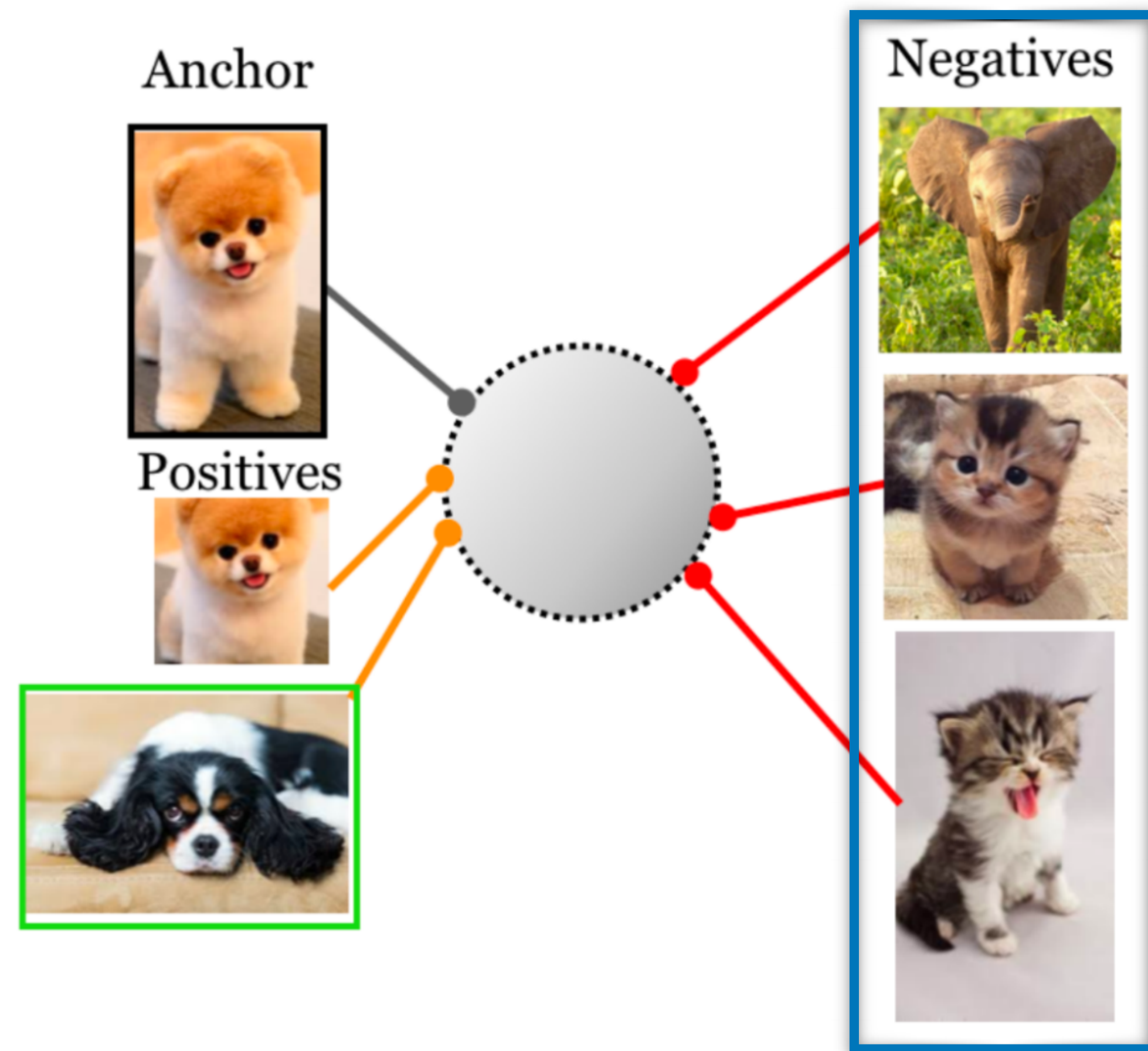
Supervised Contrastive (SupCon)

Supervised Contrastive Learning



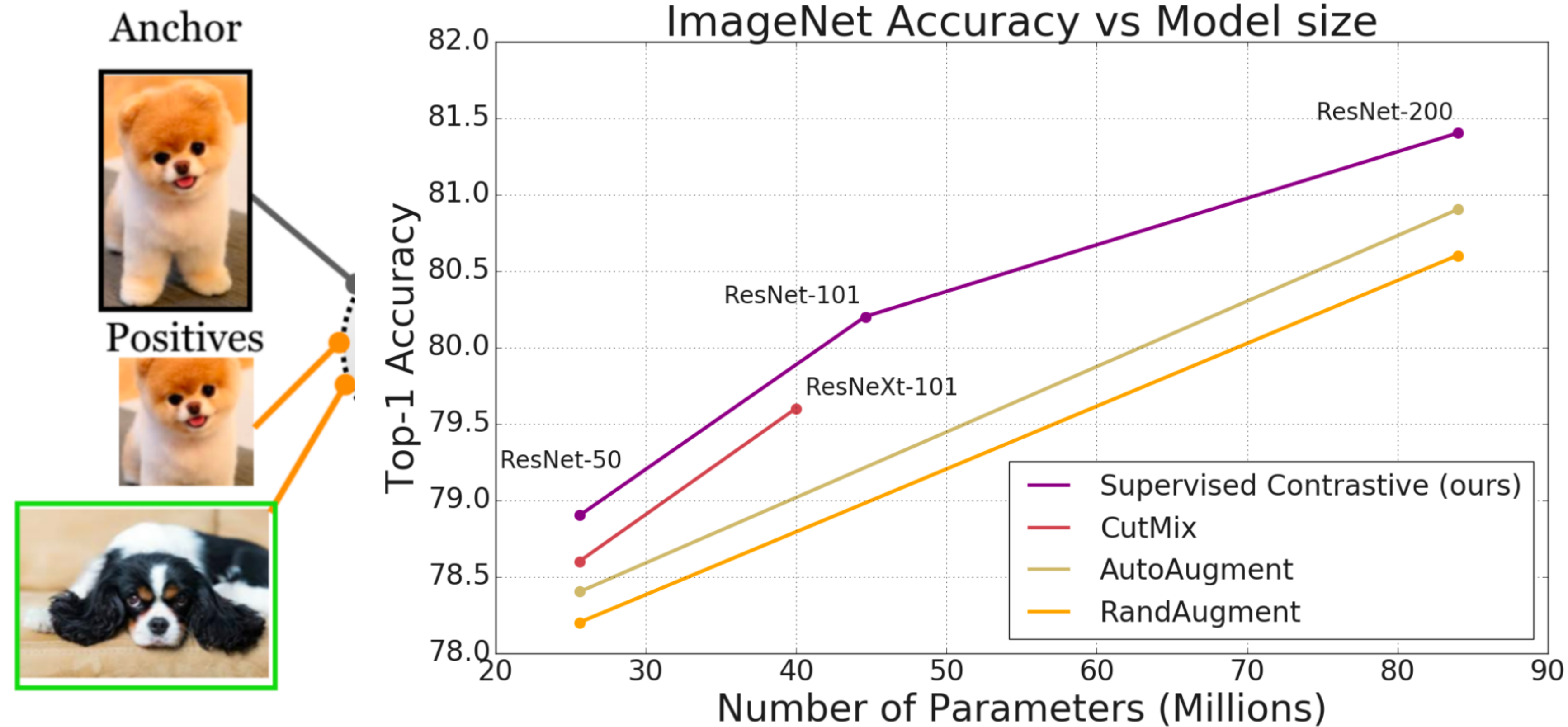
Supervised Contrastive (SupCon)

Supervised Contrastive Learning



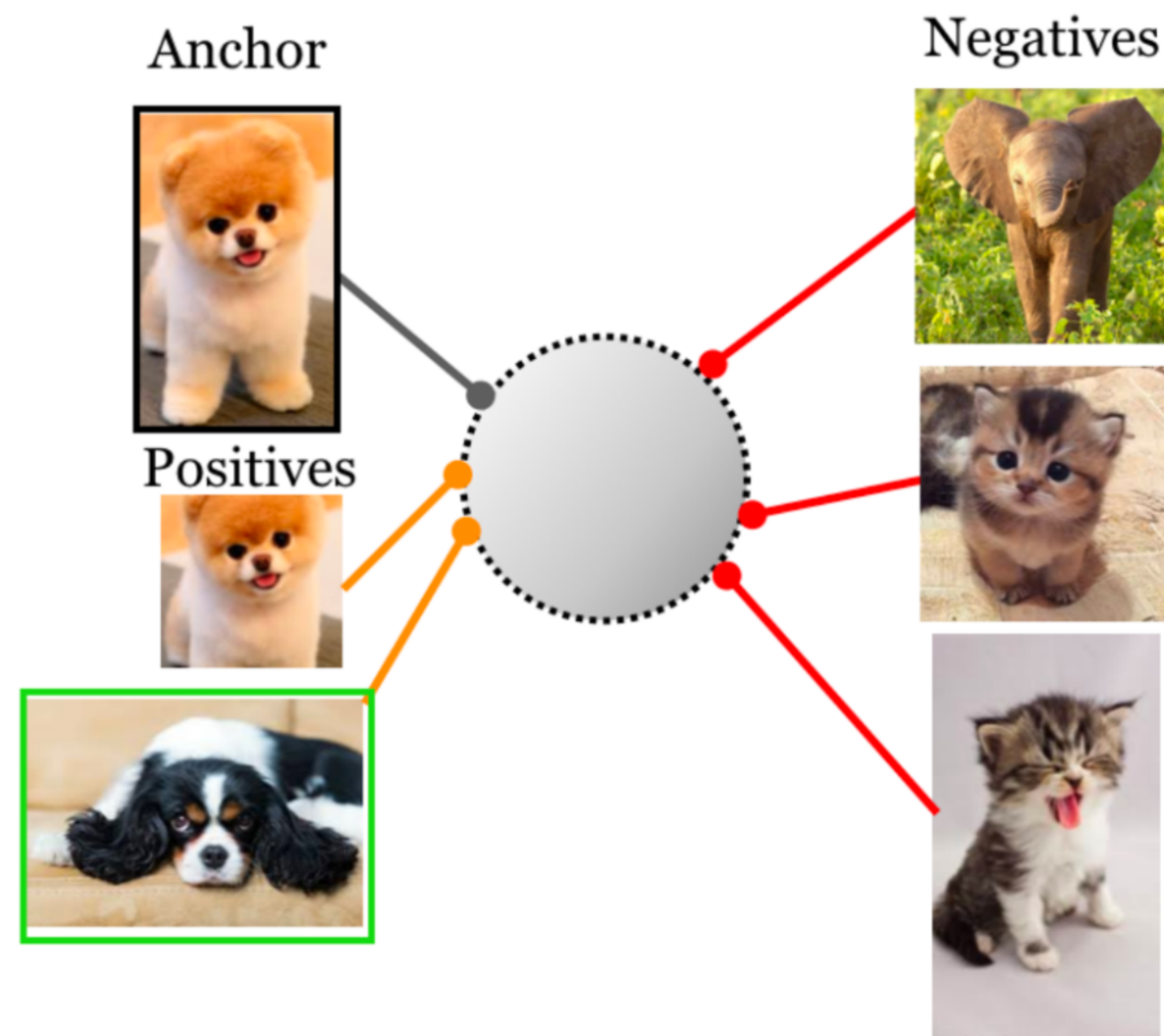
Supervised Contrastive (SupCon)

Supervised Contrastive Learning

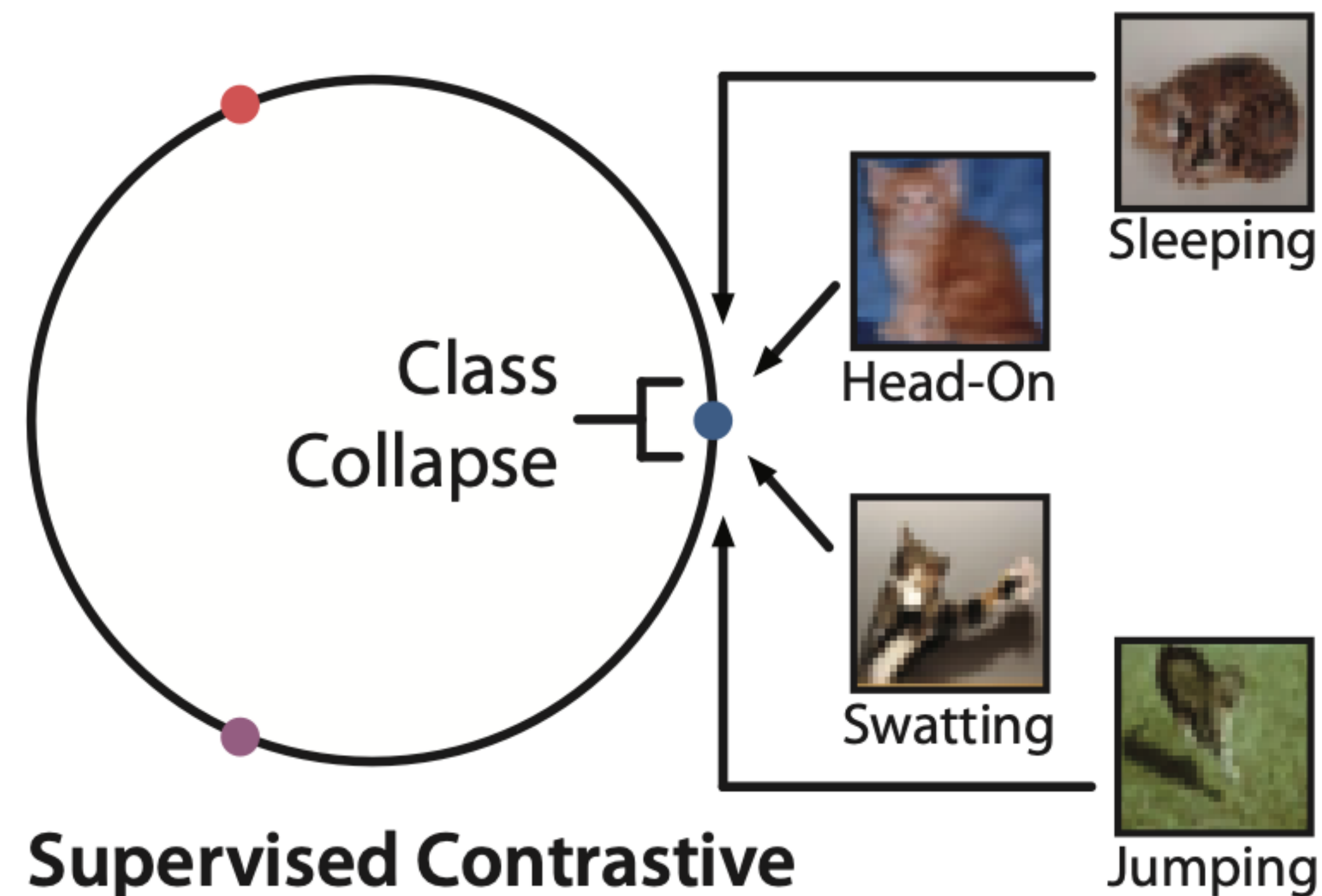


Supervised Contrastive (SupCon)

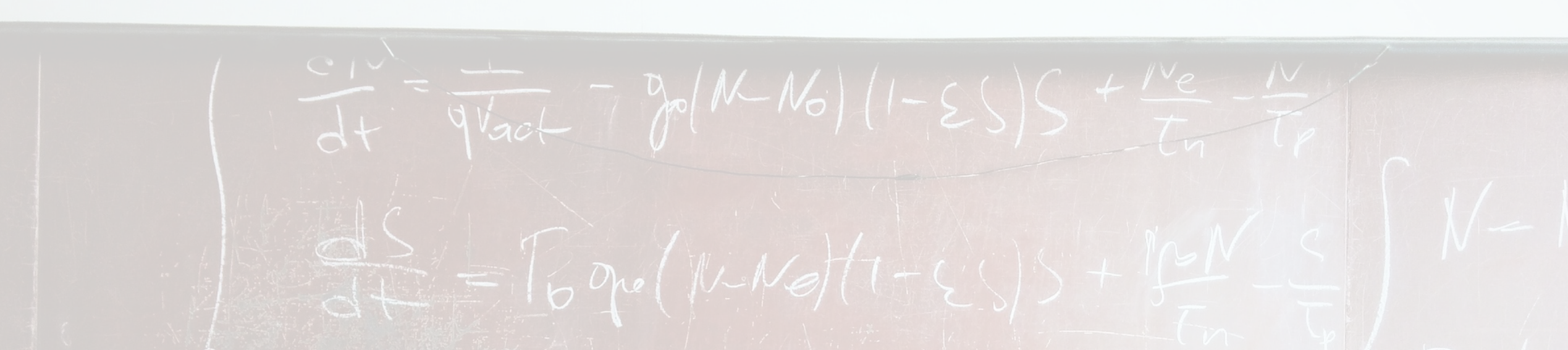
Supervised Contrastive Learning



Supervised Contrastive (SupCon)



...at the cost of class collapse



Handwritten mathematical equations on a chalkboard. The top equation is $\frac{dN}{dt} = \frac{1}{q_{\text{fact}}} - \gamma_0(N-N_0)(1-\epsilon S)S + \frac{N_e}{T_n} - \frac{N}{T_p}$. The bottom equation is $\frac{dS}{dt} = T_0 \gamma_0(N-N_0)(1-\epsilon S)S + \frac{\gamma_0 N}{T_n} - \frac{S}{T_p}$. To the right, there is a partial equation $N =$.

Background



Challenge 1: Balancing Spread



Challenge 2: Subclass Clustering



Experiments

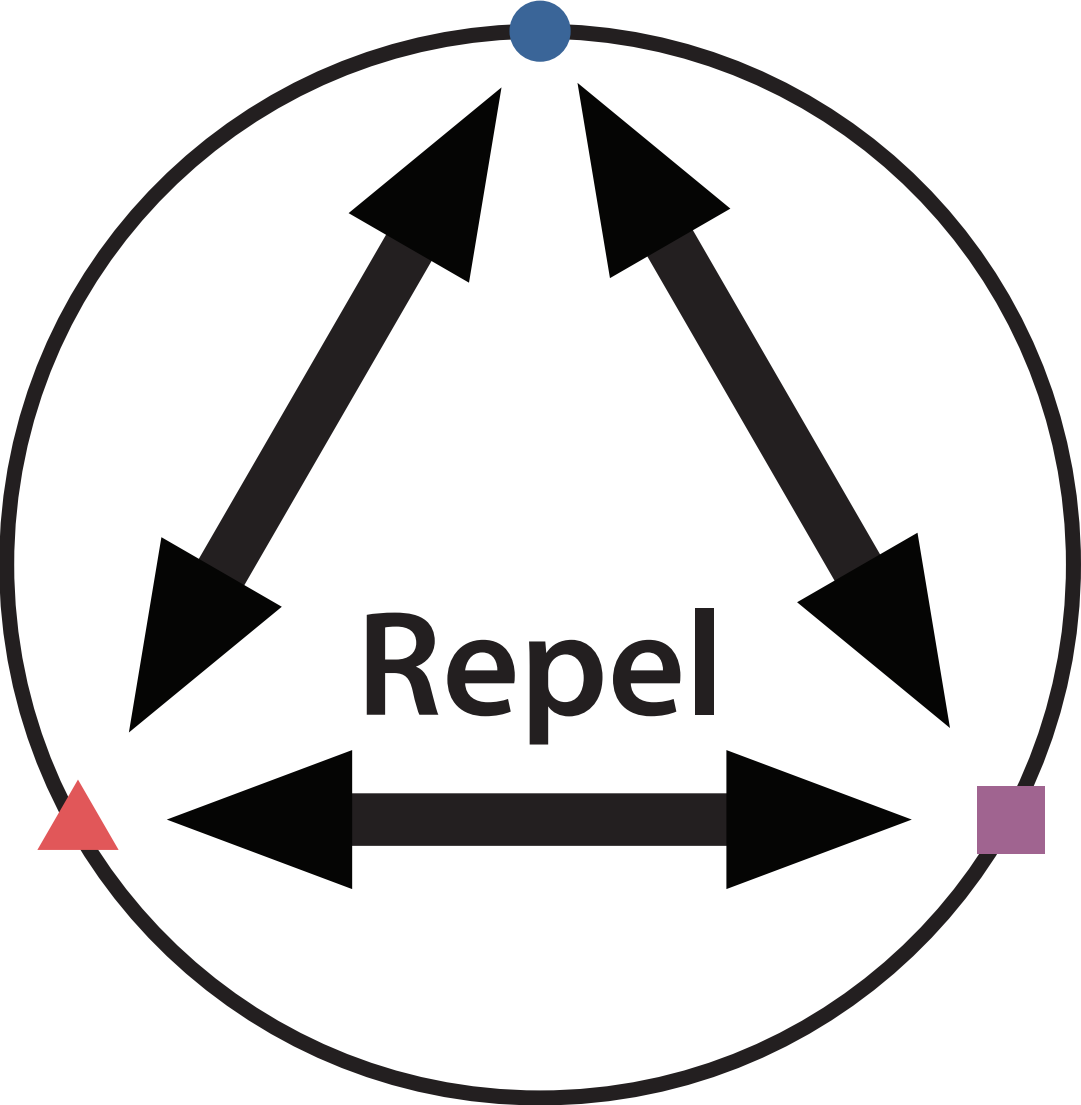


Challenge 1: Balancing Spread

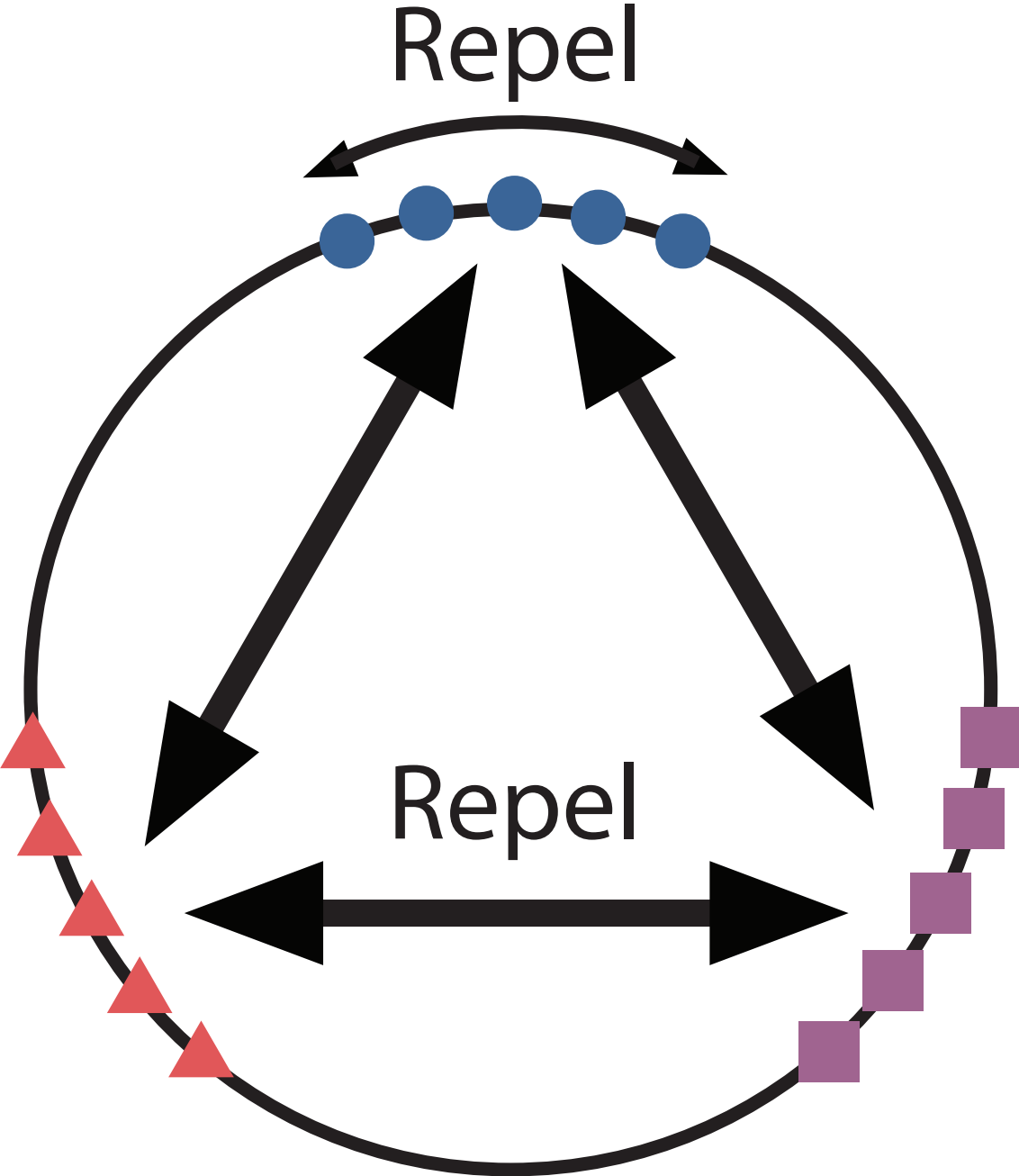
SupCon



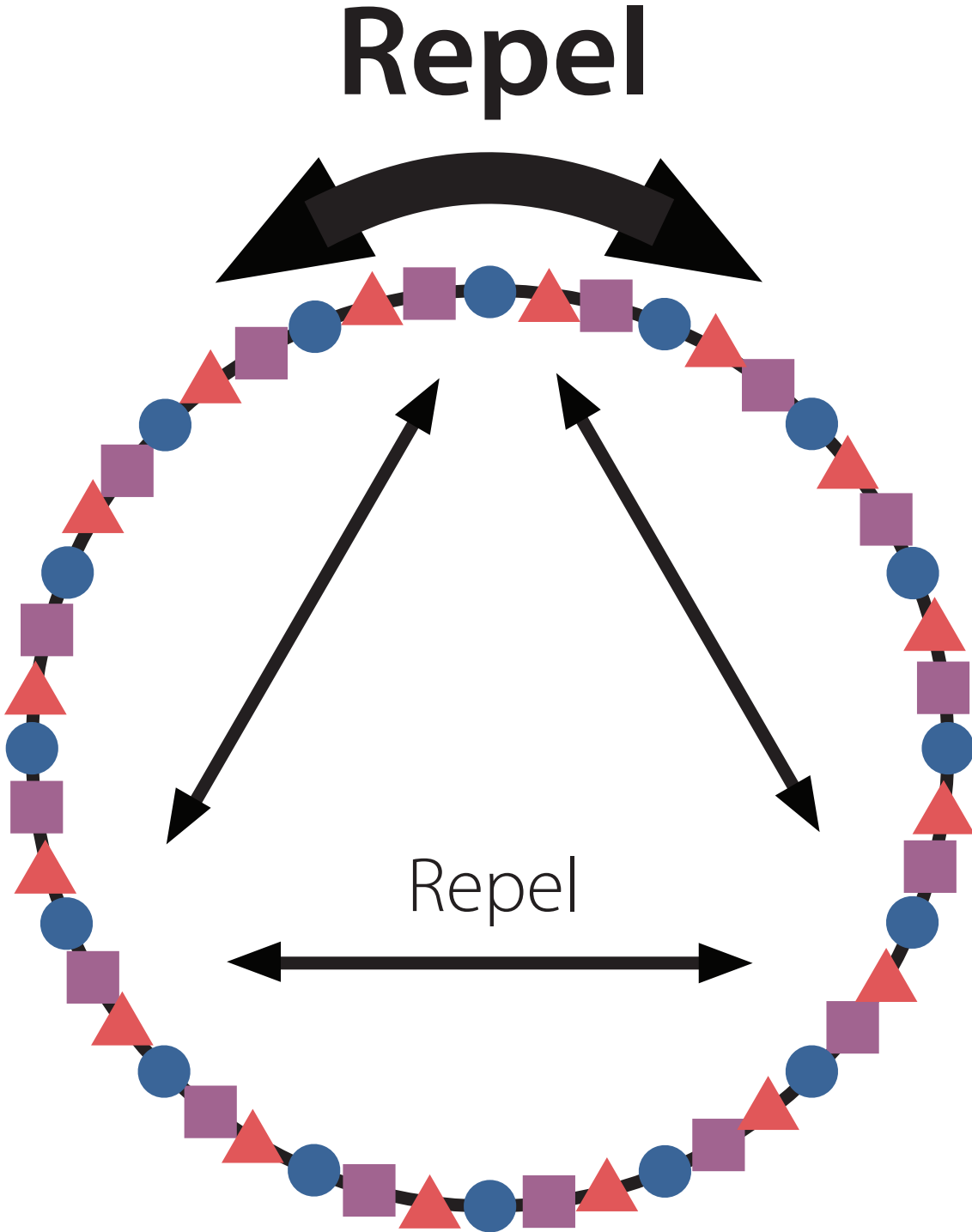
Class-Conditional
InfoNCE



Class Collapse



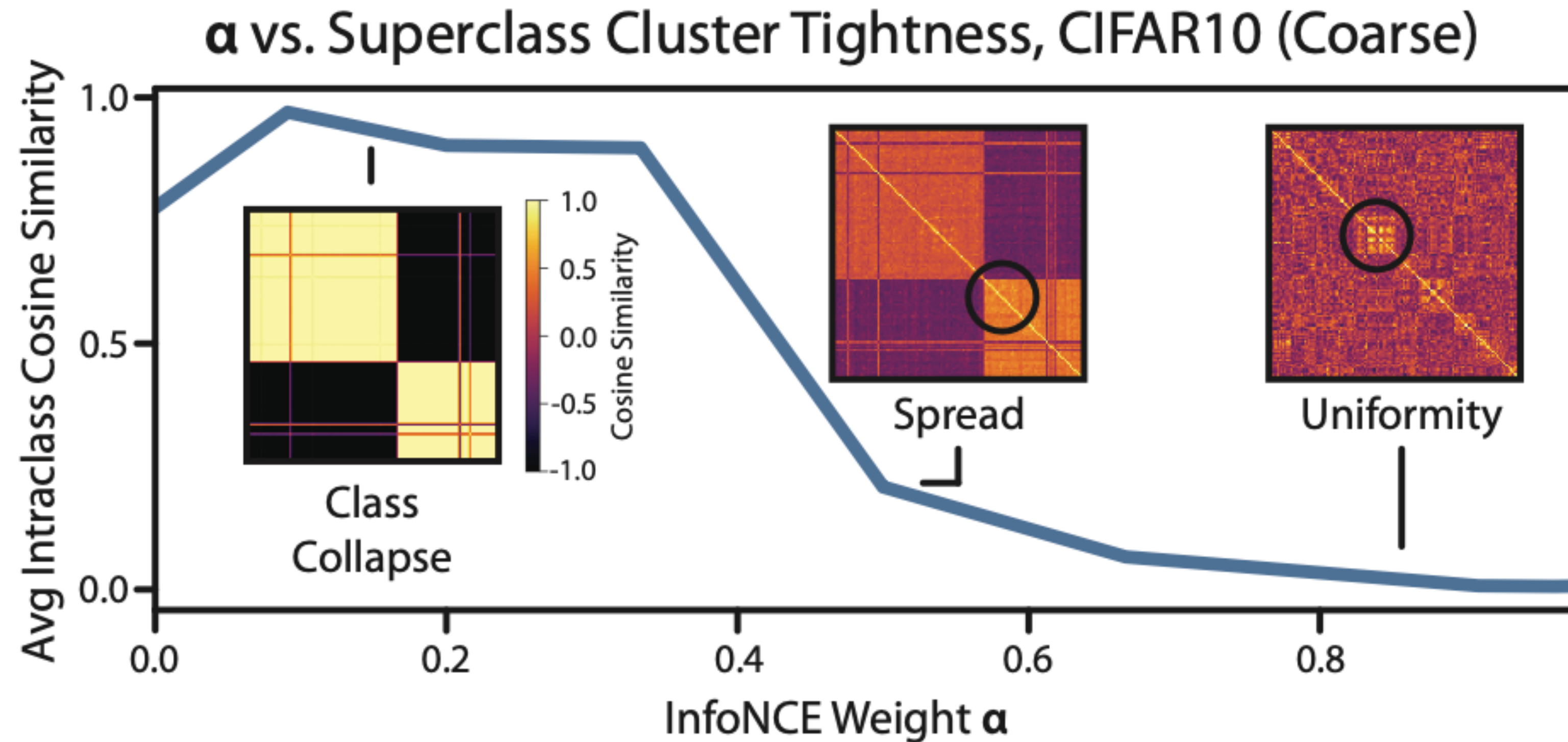
Spread

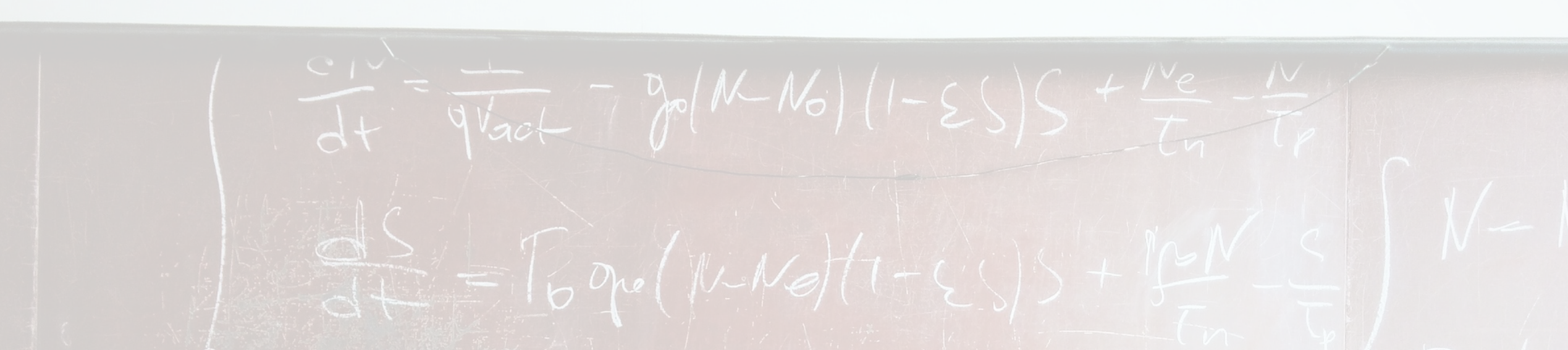


Uniform

Adding a Weighted Class-Conditional InfoNCE Loss

$$\text{Lspread} = \underbrace{(1-\alpha) \text{ Supcon}}_{\text{collapse}} + \underbrace{\alpha \text{ Class-Conditional InfoNCE}}_{\text{uniform per class}}$$





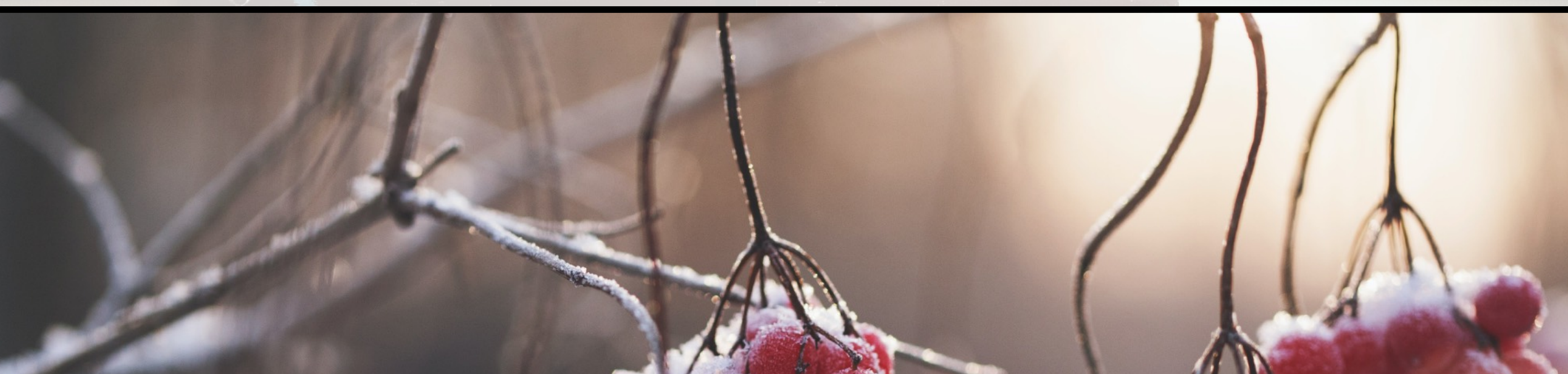
Handwritten equations on a chalkboard:

$$\frac{dN}{dt} = \frac{1}{q_{\text{fact}}} - q_0(N-N_0)(1-\epsilon S)S + \frac{N_e}{T_n} - \frac{N}{T_p}$$
$$\frac{dS}{dt} = T_0 q_0(N-N_0)(1-\epsilon S)S + \frac{q_0 N}{T_n} - \frac{S}{T_p} \quad N \leftarrow N$$

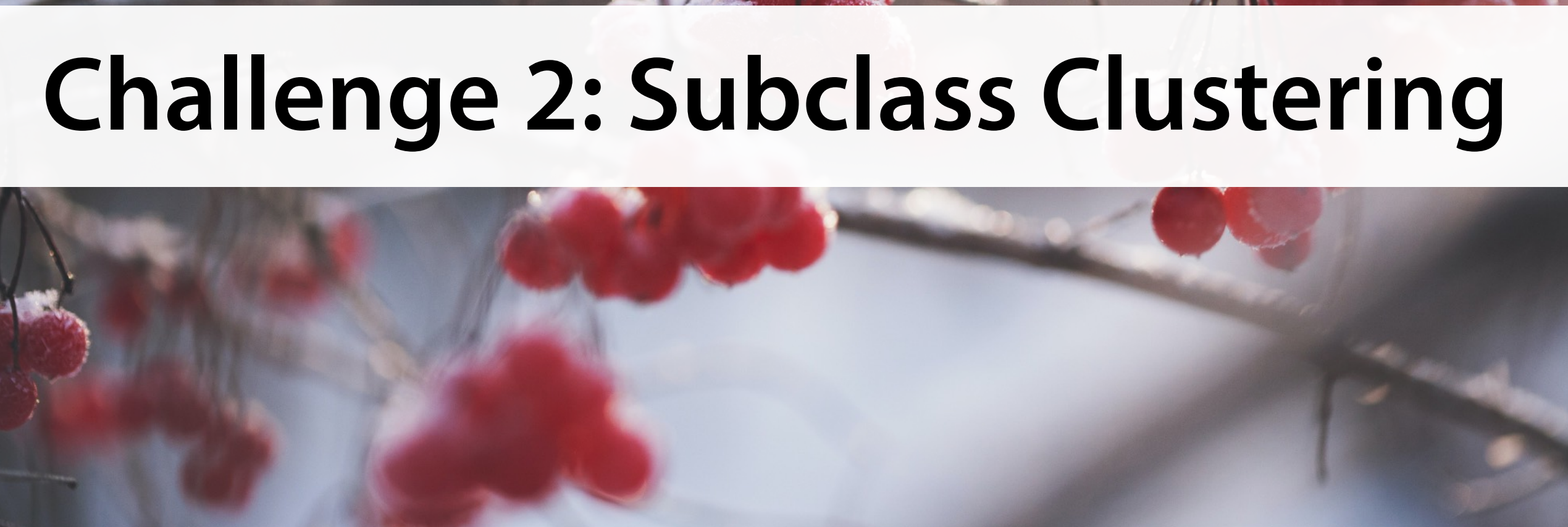
Background



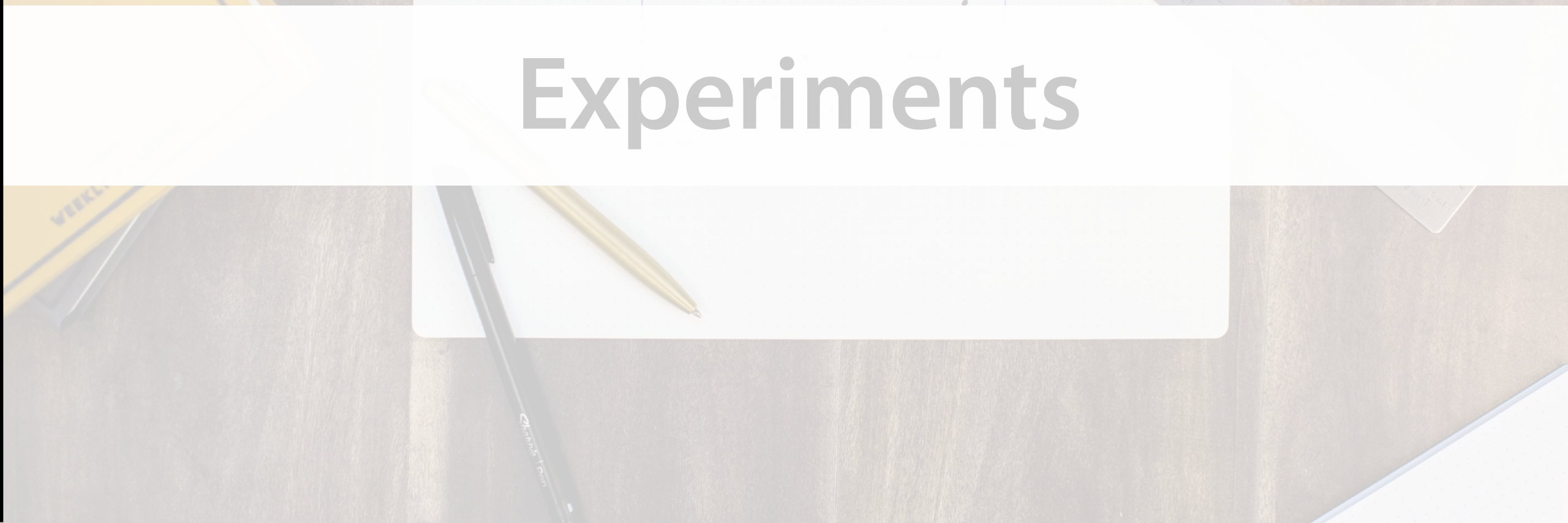
Challenge 1: Balancing Spread



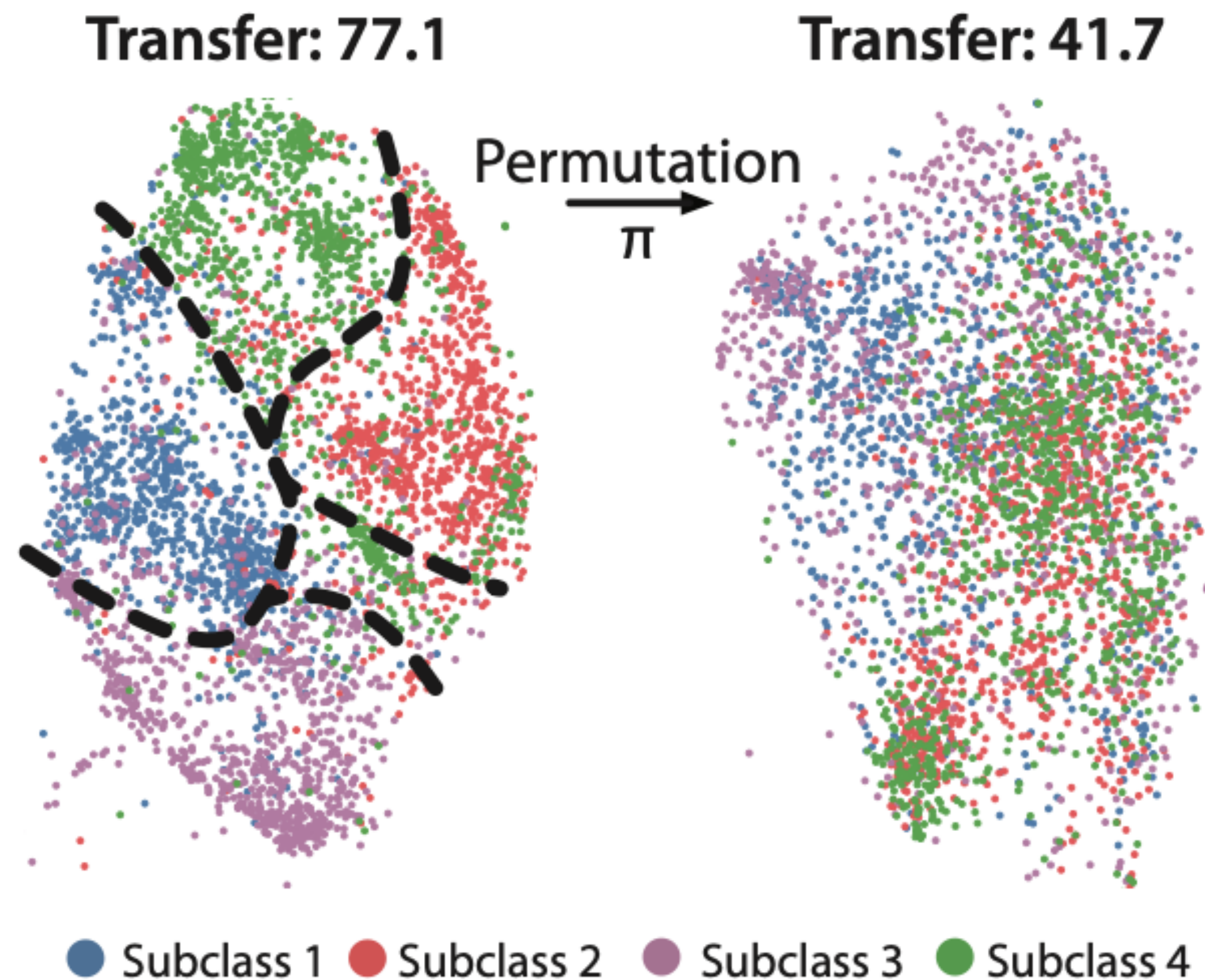
Challenge 2: Subclass Clustering



Experiments



Challenge 2: Subclass Clustering



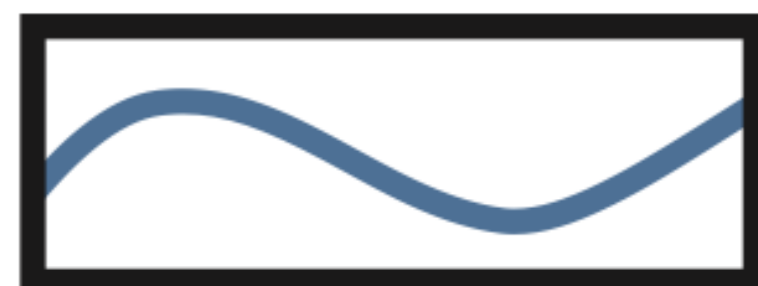
**Superclass structure is not enough,
we also need subclass clustering.**

Breaking Permutation Invariance

How do we break invariance for good downstream transfer?

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Lipschitzness

Unrealistic for modern neural networks

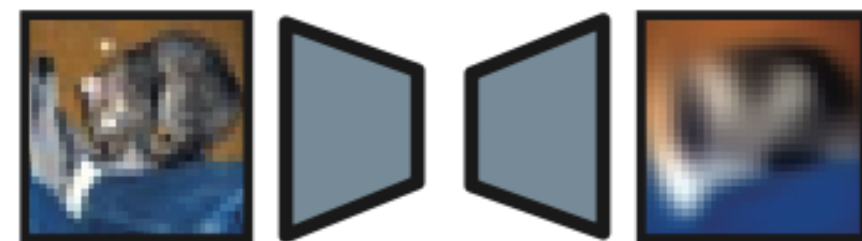
Breaking Permutation Invariance

How do we break invariance for good downstream transfer?



Lipschitzness

Unrealistic for modern neural networks



Autoencoder

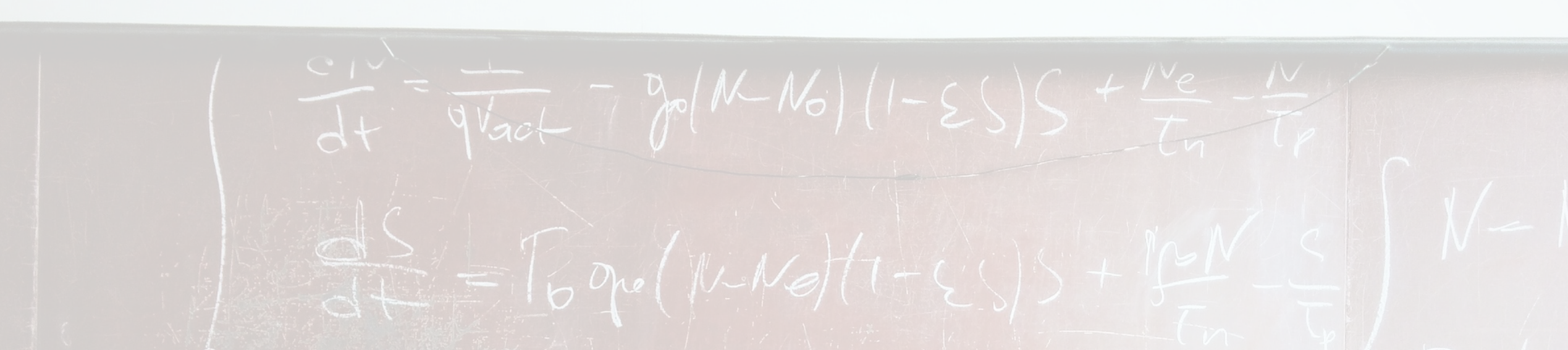
"Reverse Lipschitz" assumption only on the decoder



Data

Augmentation

Lipschitz assumption on augmented data, not on all data



Handwritten equations on a chalkboard:

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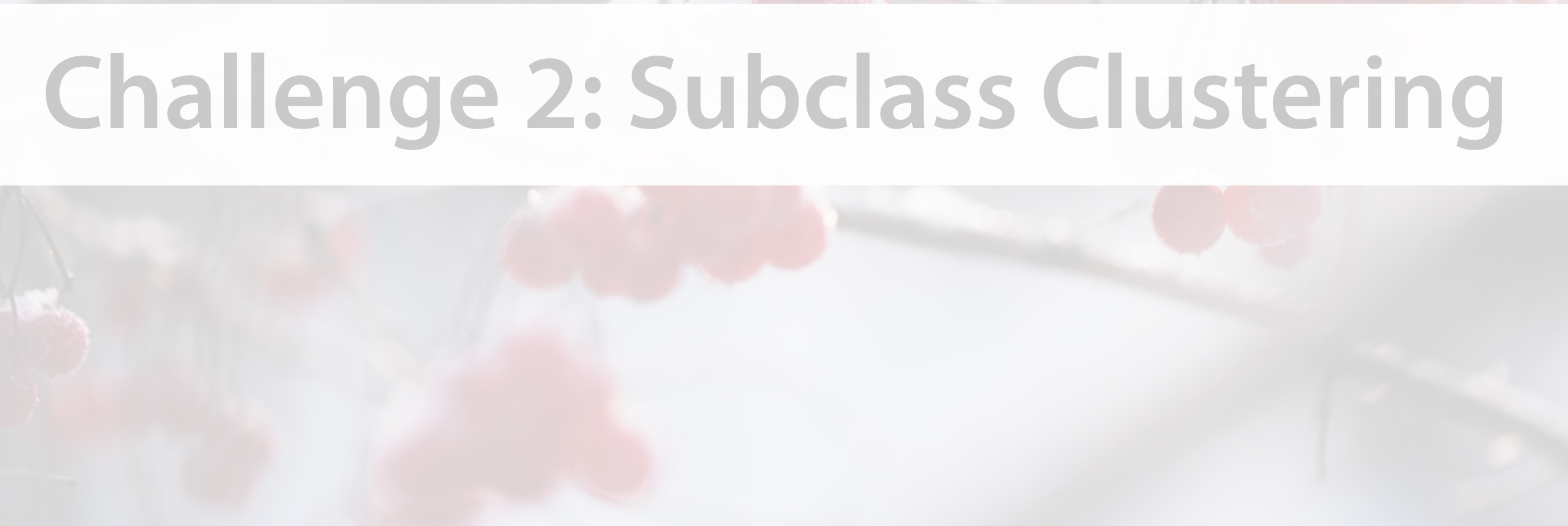
Background



Challenge 1: Balancing Spread

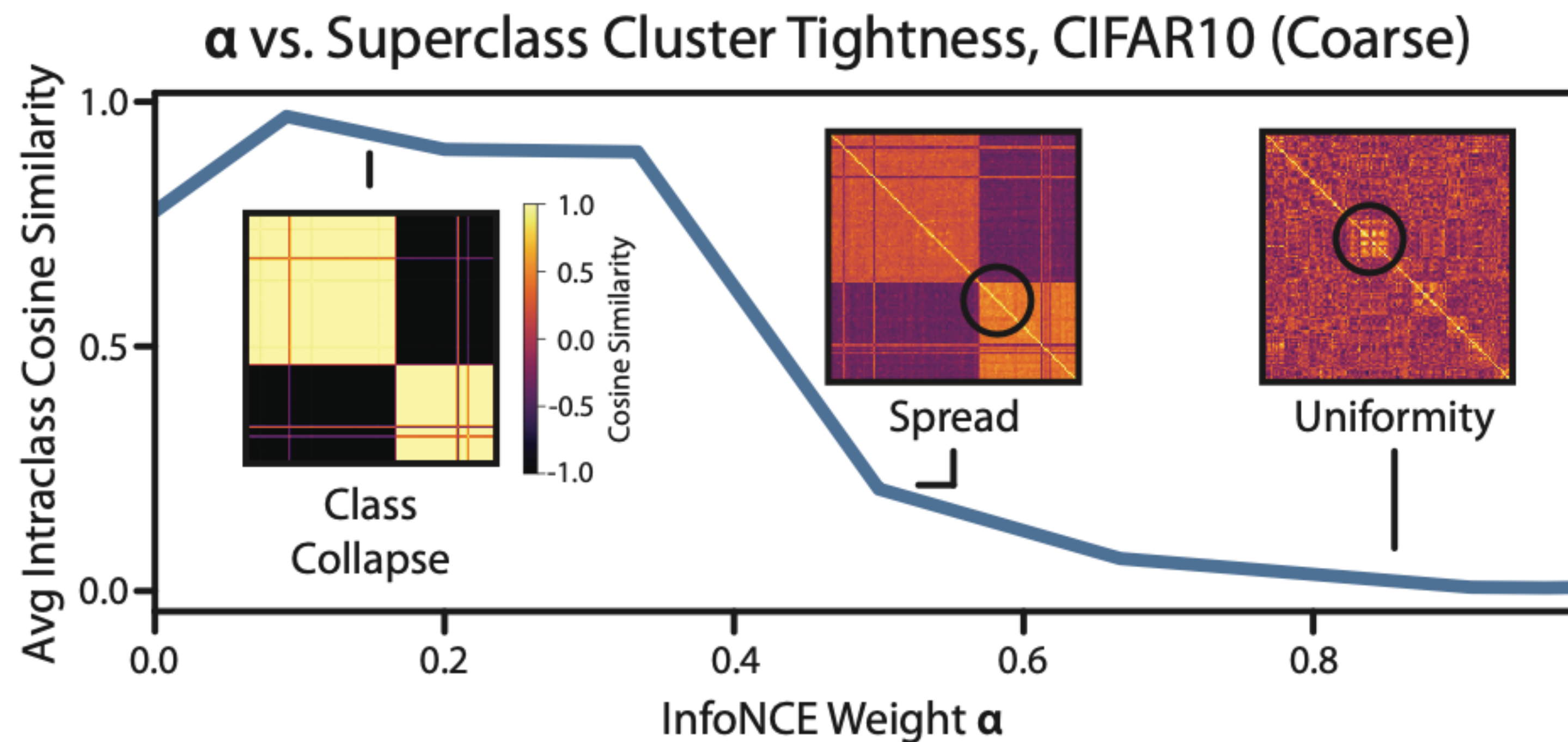


Challenge 2: Subclass Clustering

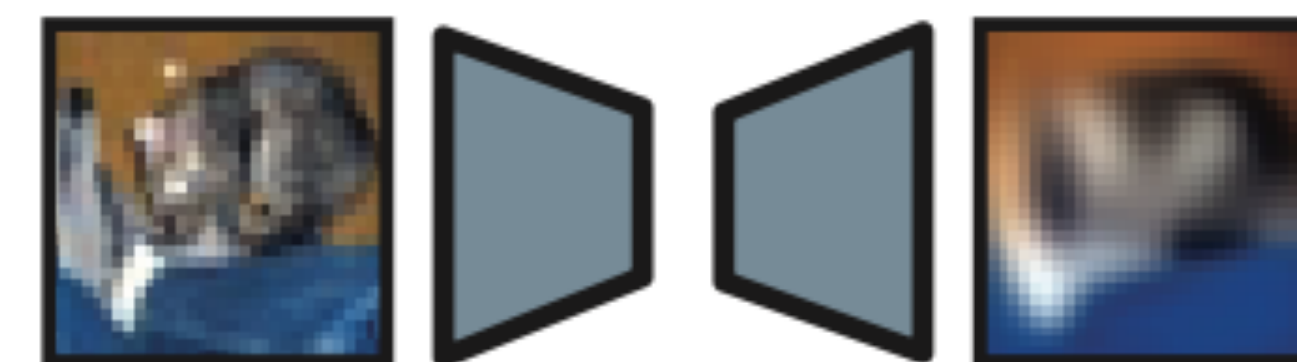


Experiments

Thanos



+



Autoencoder



Data Augmentation

Better Transfer (11.1 points on average)

Table 3. Coarse-to-fine transfer learning performance.

	Method	CIFAR10	CIFAR100	CIFAR100-U	MNIST	TinyImageNet
Baselines	InfoNCE (Chen et al., 2020a)	77.6 ± 0.1	60.5 ± 0.1	56.4 ± 0.3	98.4 ± 0.1	44.9 ± 0.1
	SupCon (Khosla et al., 2020)	51.8 ± 1.2	56.1 ± 0.1	49.8 ± 0.3	95.4 ± 0.1	43.9 ± 0.1
	SupCon + InfoNCE (Islam et al., 2021)	77.6 ± 0.1	55.7 ± 0.1	48.0 ± 0.2	98.6 ± 0.1	46.1 ± 0.1
Ours	cAuto	71.4 ± 0.1	62.9 ± 0.1	58.7 ± 0.5	98.7 ± 0.1	47.1 ± 0.1
	SupCon + cNCE (L_{spread})	77.1 ± 0.1	58.7 ± 0.2	53.5 ± 0.4	98.5 ± 0.1	45.8 ± 0.1
	SupCon + cAuto	71.7 ± 0.1	63.8 ± 0.6	59.8 ± 0.3	98.7 ± 0.1	49.3 ± 0.1
	SupCon + cNCE + cAuto (THANOS)	79.1 ± 0.2	65.0 ± 0.2	59.7 ± 0.3	99.0 ± 0.1	49.6 ± 0.1

Better Transfer (11.1 points on average)

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	SupCon + cNCE + cAuto (THANOS)	79.1 ± 0.2	65.0 ± 0.2	59.7 ± 0.3	99.0 ± 0.1	49.6 ± 0.1

...and Robustness! (4.7 points on average, 11.5 on CelebA)

Table 4. Unsupervised subclass recovery (top, F1), and worst-group performance (AUROC for ISIC, Acc for others).

Method	Group Labels	Waterbirds	ISIC	CelebA
Sub-Group Recovery				
Sohoni et al. (2020)	✗	56.3	74.0	24.2
SupCon	✗	47.1	92.5	19.4
THANOS	✗	59.0	93.8	24.8
Worst-Group Robustness				
Sohoni et al. (2020)	✗	88.4	92.0	55.0
JTT (Liu et al., 2021)	✗	83.8	91.8	77.9
SupCon	✗	86.8	93.3	66.1
THANOS	✗	88.6	92.6	89.4
GroupDRO	✓	90.7	92.3	88.9

Thank You!



Poster Session: Wednesday, July 20, 6:30-8:30 PM

Contact:

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Mayee Chen (mfchen@stanford.edu, @MayeeChen)

arXiv: <https://arxiv.org/abs/2204.07596>

Mayee F. Chen*, Daniel Y. Fu*, Avanika Narayan, Michael Zhang, Zhao Song, Kayvon Fatahalian, Christopher Ré. Perfectly Balanced: Improving Transfer and Robustness of Supervised Contrastive Learning. *ICML 2022*.

AIBSD workshop paper: <https://aibsdworkshop.github.io/2022/index.html>

Daniel Y. Fu*, Mayee F. Chen*, Michael Zhang, Kayvon Fatahalian, Christopher Ré. The Details Matter: Preventing Class Collapse in Supervised Contrastive Learning. *AIBSD @ AAAI 2022*. **Best Paper**.

Blog: <https://hazyresearch.stanford.edu/blog/2022-04-19-contrastive-2>

Code: <https://github.com/HazyResearch/thanos-code>



Dan Fu



Mayee Chen



Avanika
Narayan



Michael
Zhang