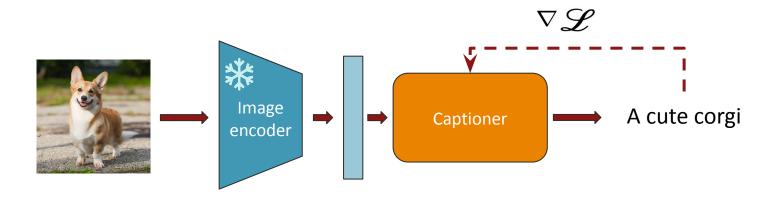
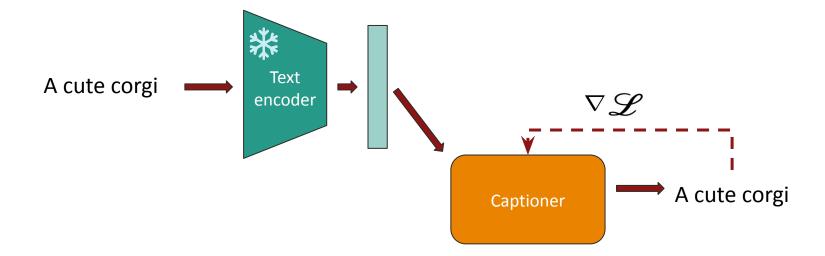
# Connect, Collapse, Corrupt: Learning Cross-Modal Tasks with Uni-Modal Data

Yuhui Zhang\*, Elaine Sui\*, Serena Yeung-Levy ICLR 2024

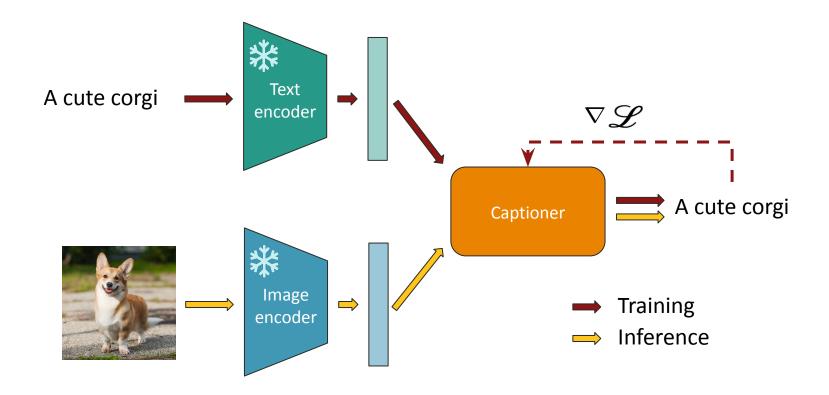
### Traditional Image Captioning Training



### **Text-Only Image Captioning Training**

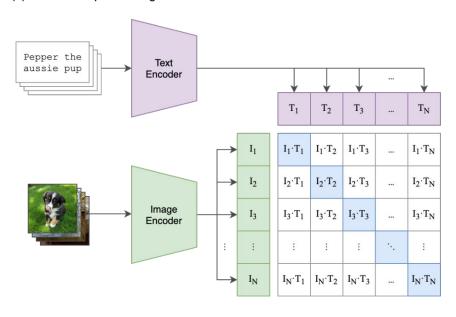


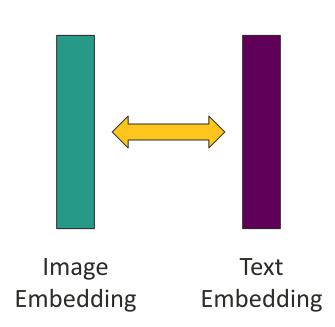
### **Text-Only Image Captioning Training**



### Large-Scale Multi-Modal Contrastive Learning

#### (1) Contrastive pre-training

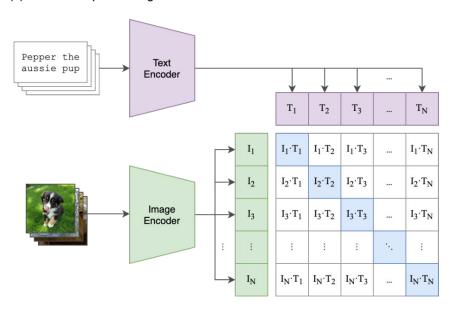


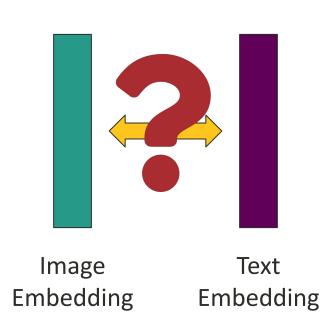


(Radford et al., 2021)

### Large-Scale Multi-Modal Contrastive Learning

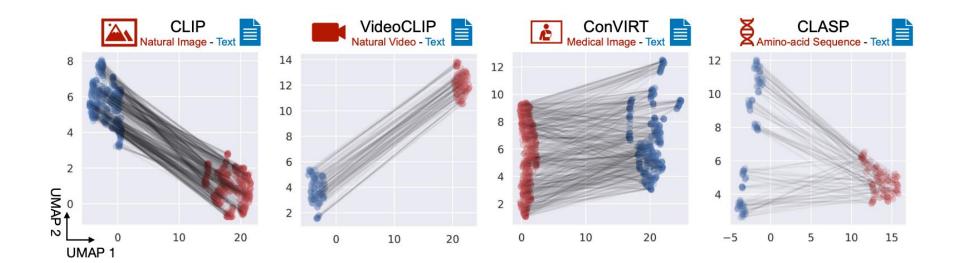
#### (1) Contrastive pre-training





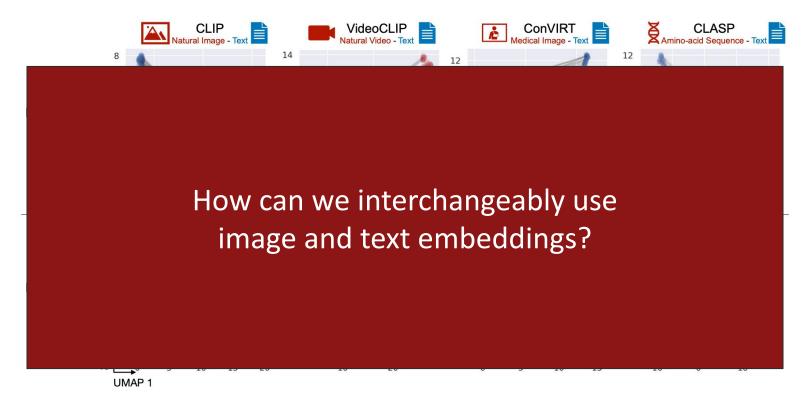
(Radford et al., 2021)

## Modality Gap



(Liang et al., 2022)

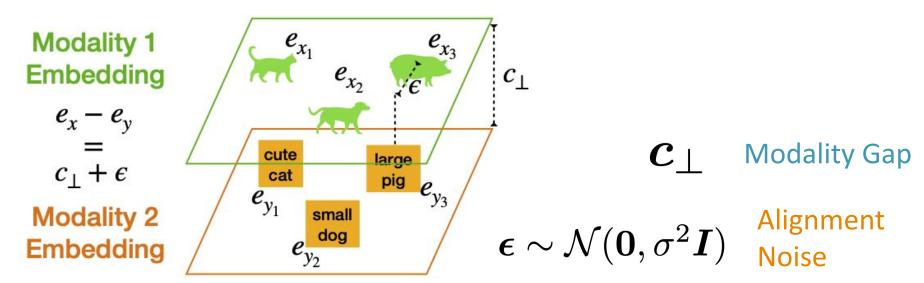
### **Modality Gap**



(Liang et al., 2022)

### Representation Space Geometry

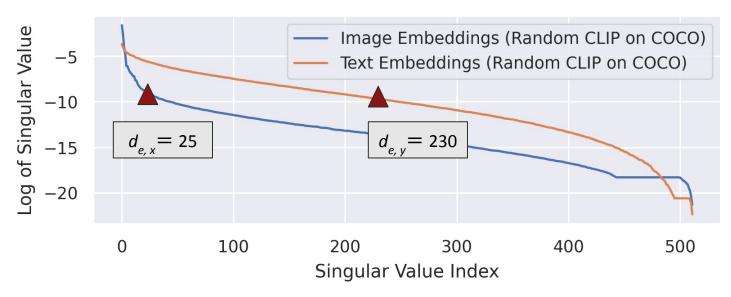
$$oldsymbol{e}_x - oldsymbol{e}_y = oldsymbol{c}_\perp + oldsymbol{\epsilon}$$
 Empirically Verified



### Theoretical Explanation of Geometry

### Modality Gap Emerges at Initialization

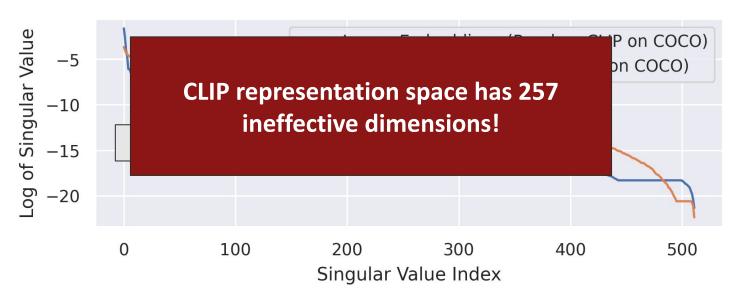
### **Dimensional Collapse of the Representation Space**



Effective dimensions are small given a randomly initialized 512-d CLIP embeddings.

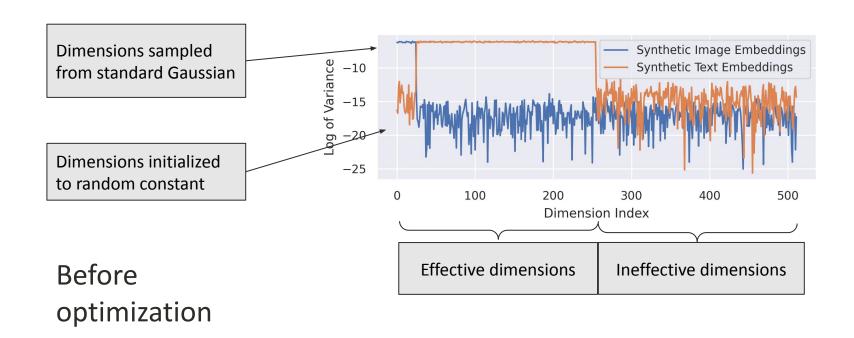
### Modality Gap Emerges at Initialization

#### **Dimensional Collapse of the Representation Space**



Effective dimensions are small given a randomly initialized 512-d CLIP embeddings.

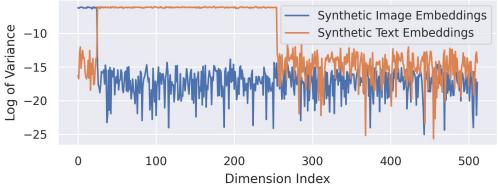
### Modality Gap is Preserved after Optimization

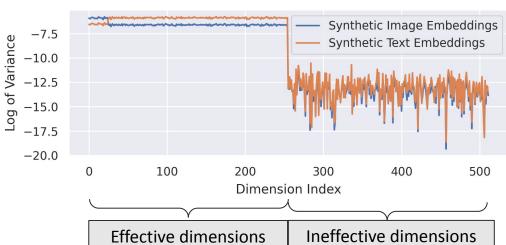


## Modality Gap is Preserved after Optimization

Before optimization

After optimization





are aligned

remain unchanged

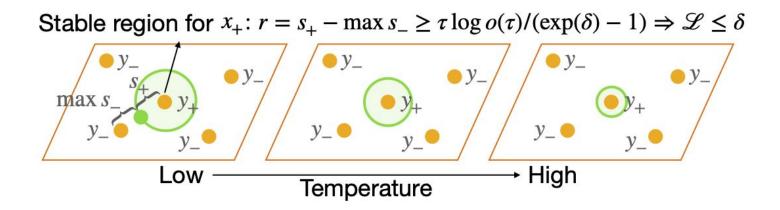
See gradient analysis in paper.

### Alignment Noise

Given the contrastive loss  $\mathcal{L}_i = -\log rac{\exp(m{e}_{x_i} \cdot m{e}_{y_i}/ au)}{\sum_{j=1}^n \exp(m{e}_{x_i} \cdot m{e}_{y_j}/ au)}$ ,

the stable region r is a function of temperature au .

Within the stable region, the optimization halts.



### C<sup>3</sup>: Connect, Collapse, Corrupt

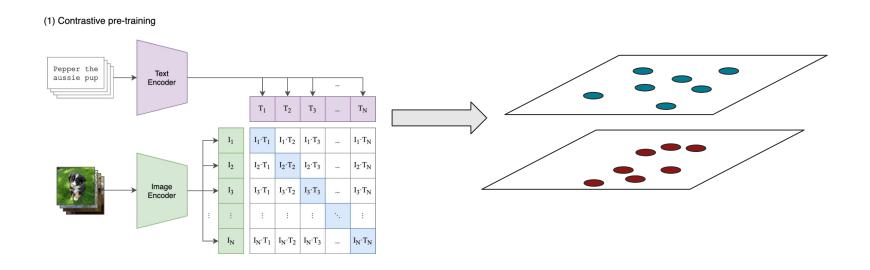
**Connect:** Relate concepts with multi-modal contrastive learning.

**Collapse:** Close the modality gap by subtracting the embedding mean from each modality.

**Corrupt:** Add noise to the embeddings as regularization during training.

### 1. Connect

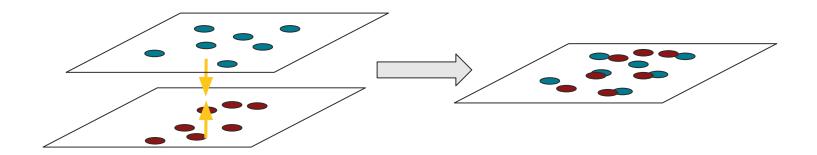
Relate concepts with multi-modal contrastive learning.



(Radford et al., 2021; Liang et al., 2022)

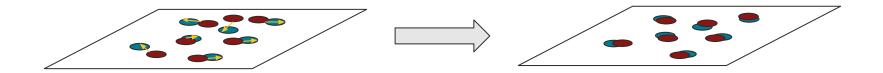
### 2. Collapse

Close the modality gap by subtracting the embedding mean from each modality.



### 3. Corrupt

Add noise to the embeddings as regularization during training.



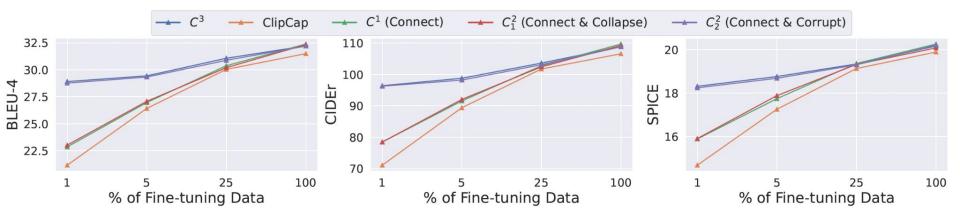
### **Image Captioning on MSCOCO**

Method	Conn.	Coll.	Corr.	BLEU-1↑	BLEU-4↑	<b>METEOR</b> <sub>↑</sub>	<b>ROUGE-L</b> ↑	<b>CIDEr</b> <sub>↑</sub>	<b>SPICE</b> <sub>↑</sub>			
Baselines												
ZeroCap (2022)	X	X	X	49.8	7.0	15.4	31.8	34.5	-			
MAGIC (2022)	X	X	X	56.8	12.9	17.4	39.9	49.3	11.3			
ESPER (2022)	X	×	X	-	21.9	21.9	-	78.2	-			
CLIPRe (2023)	1	X	X	-	4.6	13.3	-	25.6	9.2			
DeCap (2023)	1	X	X	-	8.9	17.5	-	50.6	13.1			
WS-ClipCap (2023)	1	X	X	50.3	9.6	15.2	37.5	33.7	8.6			
WS-ClipCap-Multi (2023)	1	X	1	65.5	22.1	22.2	48.0	74.6	14.9			
CapDec (2022)	✓	X	✓	69.2	26.4	25.1	51.8	91.8	-			
Ours												
$C^1$	1	X	X	28.1	2.4	12.2	25.4	13.0	6.8			
$C_1^2$	1	1	×	44.4	6.1	15.5	33.6	25.2	9.2			
$egin{array}{c} C_2^2 \ C^3 \end{array}$	1	X	1	69.0	25.5	24.3	50.8	87.6	17.6			
$C^{ar{3}}$	1	1	1	<b>71.0</b> ±0.1	<b>27.7</b> ±0.1	$25.0{\scriptstyle \pm 0.0}$	<b>52.0</b> $\pm$ 0.0	<b>93.3</b> ±0.3	$18.3 \pm 0.1$			

Zero-shot image captioning results.

Train: caption-to-caption. Test: image-to-caption.

### Image Captioning in the Low-Data Regime



Semi-supervised image captioning results.

Train: caption-to-caption. Fine-tune: image-to-caption. Test: image-to-caption.

### Image Captioning Qualitative Results



 $C^1$ : Dog sniffing the back of the black and white dog in the cobblestone.

 $C_1^2$ : Dog behind a man's ear in the street.

 $C_2^2$ : A black dog looking at a dog on the sidewalk.

 $C^3$ : A dog looking at a black cat on the street.

Ground truth: A black cat is scared by a large dog.

Connect

Connect + Collapse

**Connect + Corrupt** 

Connect + Collapse + Corrupt



 $C^1$ : A red and white bus is en route to the International Space Station.

 $C_1^2$ : A PERSON IS DRIVING THE SEA WAVE

 $C_2^2$ : A group of people in a boat with a large white boat in the background.

 $C^3$ : A group of people on a boat in the water.

Ground truth: A boat in the water filled with people

Connect

**Connect + Collapse** 

**Connect + Corrupt** 

**Connect + Collapse + Corrupt** 



 $C^1$ : A person and their luggage, resting in the courtyard of the University of Southern California, observing the Eiffel Tower.

 $C_1^2$ : A person on a bench in the midst of the street to take a walk with their library.

 $C_2^2$ : A man sitting on a bench with a book.

 $C^3$ : A man and woman sitting on a park bench.

Ground truth: A man and woman sitting on a park bench under a huge tree

Connect

Connect + Collapse

**Connect + Corrupt** 

**Connect + Collapse + Corrupt** 

### Text-to-Image Generation on MSCOCO

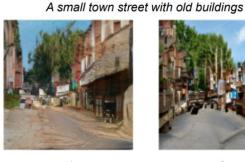
	Method	Conn.	Coll.	Corr.	$ \mathbf{FID}_{\downarrow} $	IS↑				
	Baselines									
	DALL-E (2021)	X	X	X	27.5	17.9				
Language-only	CogView (2021)	X	X	X	27.1	18.2				
	$LAFITE_G$ (2022c)	1	X	✓	20.9	24.9				
	Ours									
	$C^1$	1	X	X	29.8	22.4				
	$C_1^2$	1	✓	X	21.7	24.4				
	$egin{array}{c} C_1^2 \ C_2^2 \ C^3 \end{array}$	1	X	1	19.8	25.5				
	$C^{\overline{3}}$	1	1	✓	19.6	26.0				

Zero-shot text-to-image generation results. Train: image-to-image. Test: text-to-image.

### Text-to-Image Generation Qualitative Results













Ground Truth

 $LAFITE_G$ 

 $C^1$ 

















Ground Truth

 $LAFITE_G$ 

Connect

 $C^1$ 

 $C_1^2$ **Connect** + Collapse

Connect + Corrupt

 $C_2^2$ 

Connect + **Collapse + Corrupt** 

