

Deep Learning for Computer Vision

NTU, Fall 2023, homework3

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● Problem 1: Zero-shot Image Classification with CLIP

1. Methods analysis (3%)

- ◆ *Previous methods (e.g. VGG and ResNet) are good at one task and one task only, and requires significant efforts to adapt to a new task. Please explain why CLIP could achieve competitive zero-shot performance on a great variety of image classification datasets.*

因為 CLIP 在大量的圖像和對應的文字標題上進行訓練，並且 CLIP 的預訓練目標是最大化配對的圖像文本樣本的相似性，同時最小化未配對的樣本。

它的目標是讓配對的圖片和文字之間的相似度最大，而未配對的圖片和文字之間的相似度最小。這種方法與以前的方法有所不同，因為 CLIP 並不直接優化特定的任務，而是在自然語言的指導下進行訓練。

更具體地說，透過 caption 來構建一個線性分類器，而無需任何標記數據，然後它的強大的視覺表示可以以競爭性的性能執行任務。

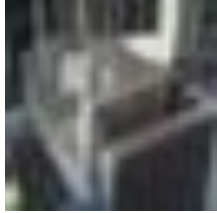




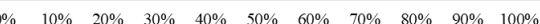
2. Prompt-text analysis (6%)

- ◆ *Please compare and discuss the performances of your model with the following three prompt templates:*
 - “This is a photo of {object}”*
 - “This is not a photo of {object}”*
 - “No {object}, no score.”*

Caption	Accuracy
This is a photo of {object}	67.84%
This is not a photo of {object}	69.52%
No {object}, no score.	45.72%

3. Quantitative analysis (6%)

- ◆ Please sample three images from the validation dataset and then visualize the probability of the top-5 similarity scores.

Picture	Top-5 similarity Scores										
	 <table> <tr> <td>a photo of a chair</td> <td>39.13%</td> </tr> <tr> <td>a photo of a couch</td> <td>31.94%</td> </tr> <tr> <td>a photo of a bed</td> <td>15.80%</td> </tr> <tr> <td>a photo of a television</td> <td>2.50%</td> </tr> <tr> <td>a photo of a willow_tree</td> <td>1.59%</td> </tr> </table>	a photo of a chair	39.13%	a photo of a couch	31.94%	a photo of a bed	15.80%	a photo of a television	2.50%	a photo of a willow_tree	1.59%
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a photo of a willow_tree	1.59%										
	 <table> <tr> <td>a photo of a fox</td> <td>90.38%</td> </tr> <tr> <td>a photo of a wolf</td> <td>3.34%</td> </tr> <tr> <td>a photo of a rabbit</td> <td>1.63%</td> </tr> <tr> <td>a photo of a porcupine</td> <td>1.23%</td> </tr> <tr> <td>a photo of a mouse</td> <td>0.75%</td> </tr> </table>	a photo of a fox	90.38%	a photo of a wolf	3.34%	a photo of a rabbit	1.63%	a photo of a porcupine	1.23%	a photo of a mouse	0.75%
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a photo of a porcupine	1.23%										
a photo of a mouse	0.75%										
	 <table> <tr> <td>a photo of a porcupine</td> <td>60.75%</td> </tr> <tr> <td>a photo of a mushroom</td> <td>7.61%</td> </tr> <tr> <td>a photo of a sunflower</td> <td>7.49%</td> </tr> <tr> <td>a photo of a crab</td> <td>4.40%</td> </tr> <tr> <td>a photo of a bee</td> <td>2.93%</td> </tr> </table>	a photo of a porcupine	60.75%	a photo of a mushroom	7.61%	a photo of a sunflower	7.49%	a photo of a crab	4.40%	a photo of a bee	2.93%
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● Problem 2: PEFT on ViT Model for Image Captioning

1. Evaluation metrics report

- ◆ Report your best setting and its corresponding CIDEr & CLIPScore on the validation data.

Model
vit_large_patch14_clip_224.openai_ft_in12k_in1k
Optimizer
<code>optimizer = torch.optim.Adam(model.parameters(), lr=0.0001)</code>
Scheduler
<code>scheduler = torch.optim.lr_scheduler.CosineAnnealingLR(optimizer, T_max=EPOCHS * len(train_loader) - 1000)</code>
Transform
<code>transform = create_transform(**resolve_data_config({}, model="vit_large_patch14_clip_224.openai_ft_in12k_in1k"))</code>

Adapter 架構	
	<pre> self.adapter_layer1 = nn.Sequential(nn.Linear(cfg.n_embd, 128), nn.GELU(approximate="tanh"), nn.Linear(128, cfg.n_embd),) self.adapter_layer2 = nn.Sequential(nn.Linear(cfg.n_embd, 128), nn.GELU(approximate="tanh"), nn.Linear(128, cfg.n_embd),) </pre>
修改 decoder 的架構	
	<pre> def forward(self, x, encoder_output): x = x + self.attn(self.ln_1(x)) x = x + self.adapter_layer1(x) x = x + self.crossattn(x, encoder_output) x = x + self.mlp(self.ln_2(x)) x = x + self.adapter_layer2(x) return x </pre>
Best (加了 adapter)	
CIDEr	CLIP Score
0.91032	0.71471

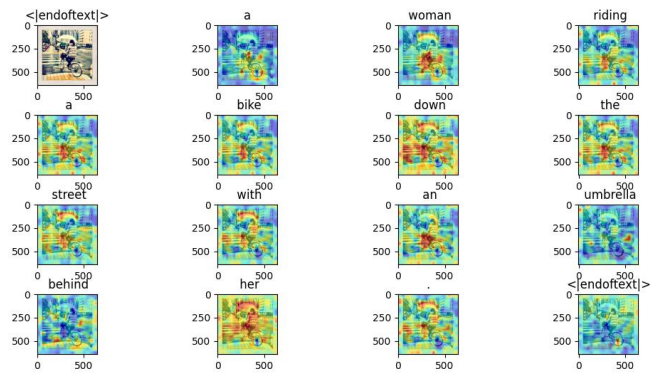
- ◆ Report 3 different attempts of PEFT and their corresponding CIDEr & CLIPScore.

	Adapter	Prefix	Lora
CIDEr	0.921	0.884	0.893
CLIPScore	0.717	0.710	0.721

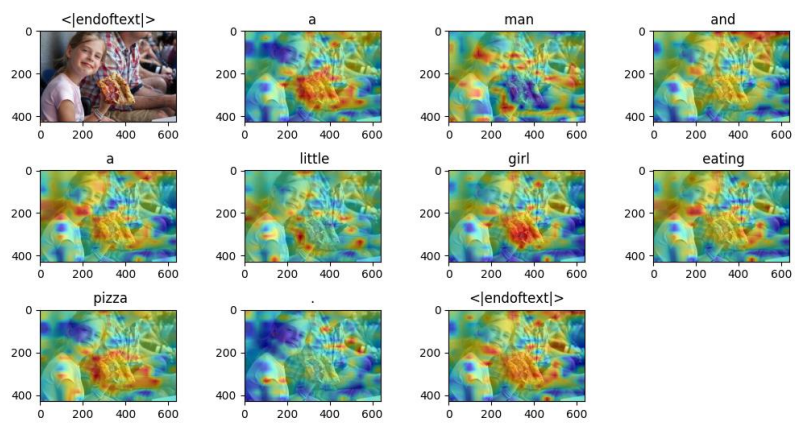
2. Visualization of Attention in Image Captioning

- ◆ please visualize the predicted caption and the corresponding series of attention maps.

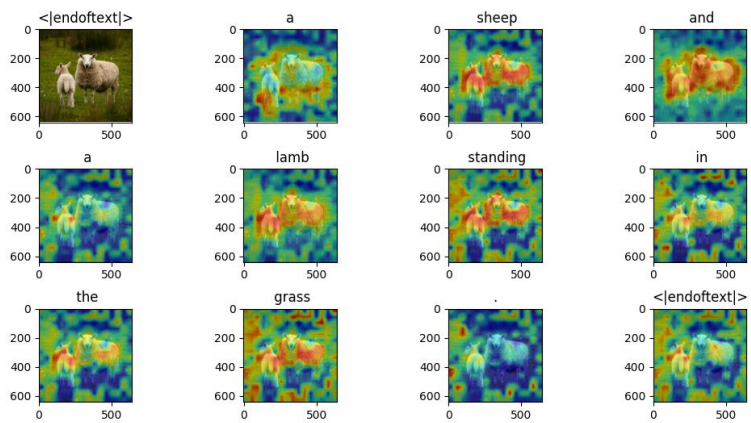
Bike.jpg



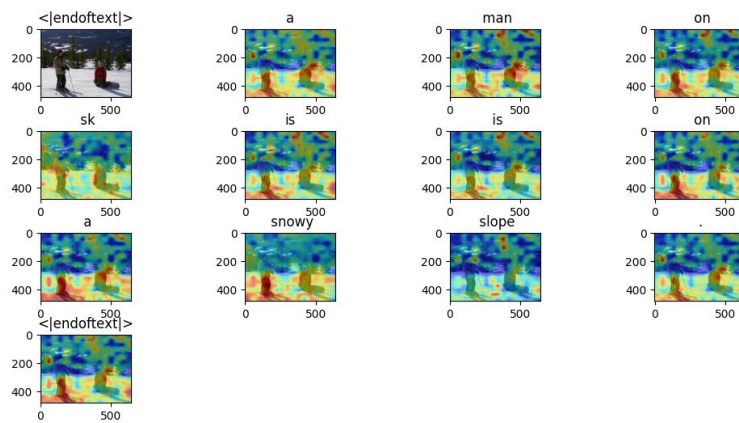
girl.jpg



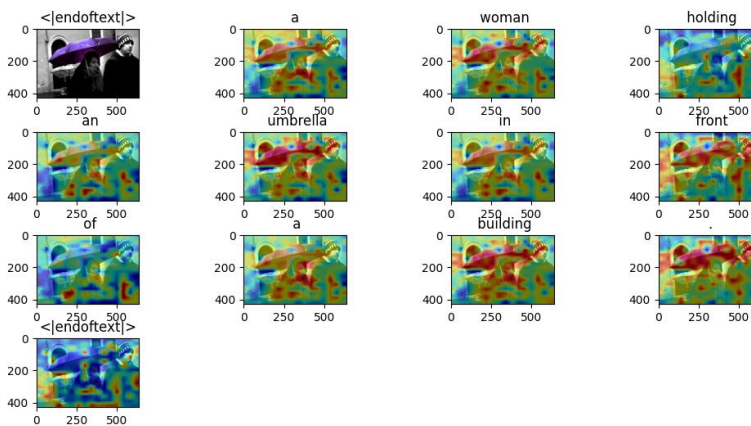
sheep.jpg



ski.jpg



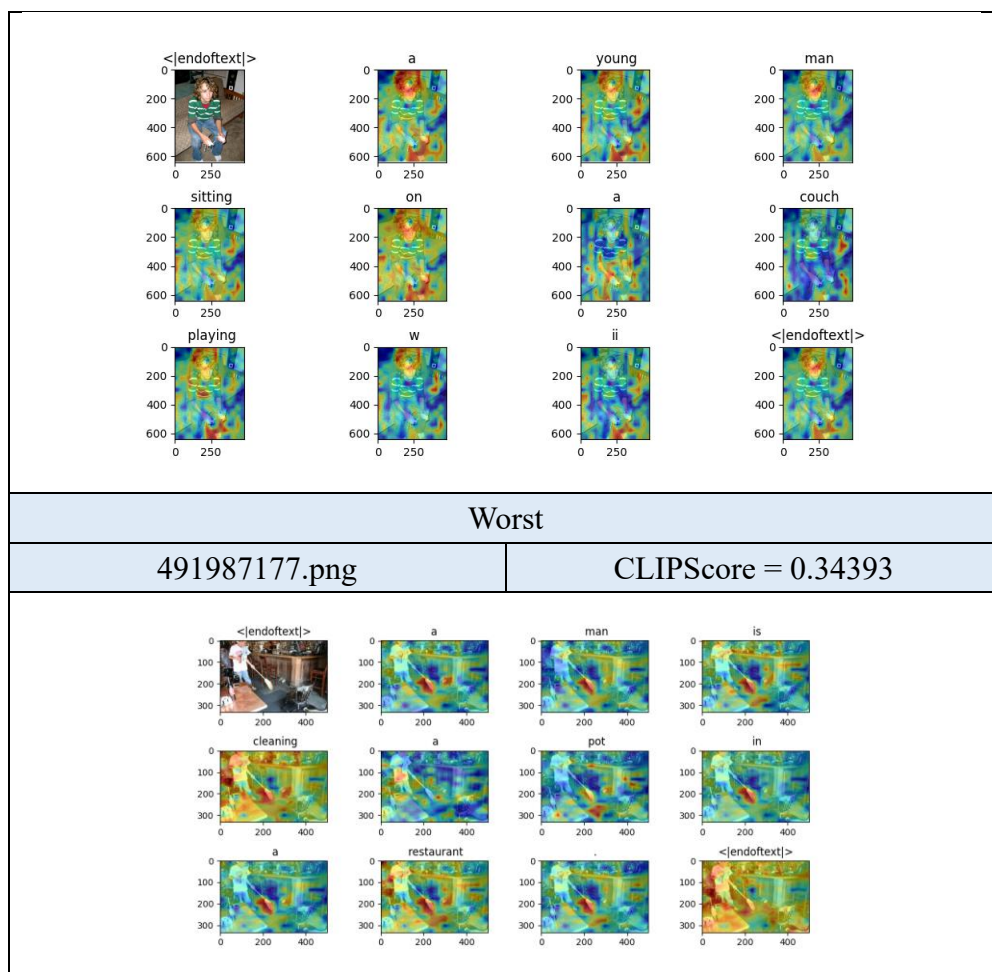
umbrella.jpg



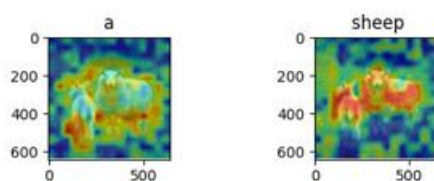
◆ According to CLIPScore, you need to:

- i. visualize top-1 and last-1 image-caption pairs
- ii. report its corresponding CLIPScore in the validation dataset of problem 2.

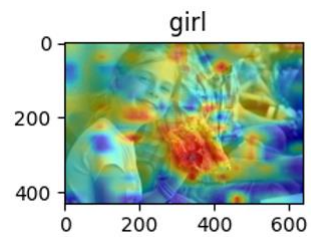
Best	
000000539189.png	CLIPScore = 1.03820



- ◆ Analyze the predicted captions and the attention maps for each word according to the previous question.
 - Is the caption reasonable?
 - Does the attended region reflect the corresponding word in the caption?
1. 大部分看起來都蠻合理的
 2. 字詞都有對應到該注意的地方，而我覺得表現最好的幾個應該是 sheep、000000539189.png、umbrella。以 sheep 為例，只要有對應到 sheep 的名詞動詞形容詞都有很明顯的聚焦(紅色地方)，而遇到 a、the、.之類無關圖片的字詞，注意力就會散開來。



但還是有少部分怪怪的地方，例如 girl.png，他的注意力很常聚焦在 pizza 上，girl 字詞或 a 也都會聚焦在 pizza 上。



● Reference

大神：Chatgpt4

Beam Search：

[pytorch beam search/src/pytorch beam search/autoregressive/search_algorithm.py at master · jarobyte91/pytorch beam search \(github.com\)](https://github.com/jarobyte91/pytorch_beam_search)