

# Read-only Prompt Optimization for Vision-Language Few-shot Learning (ICCV 2023)

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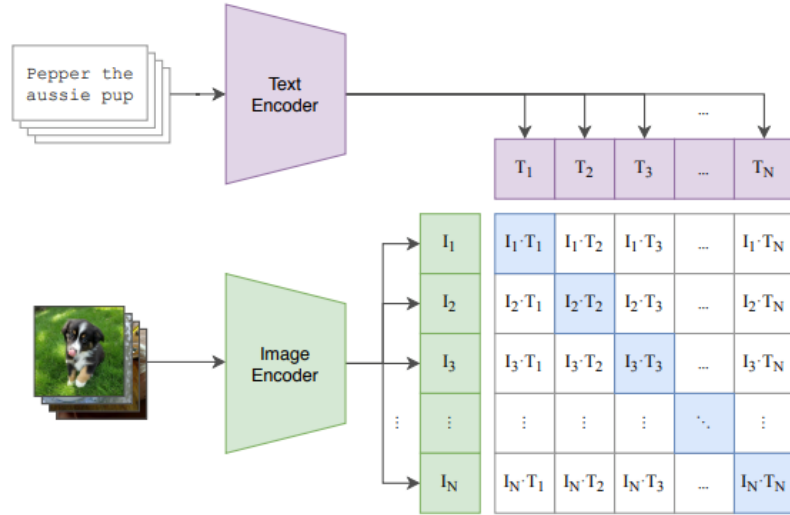
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2024. 2. 29

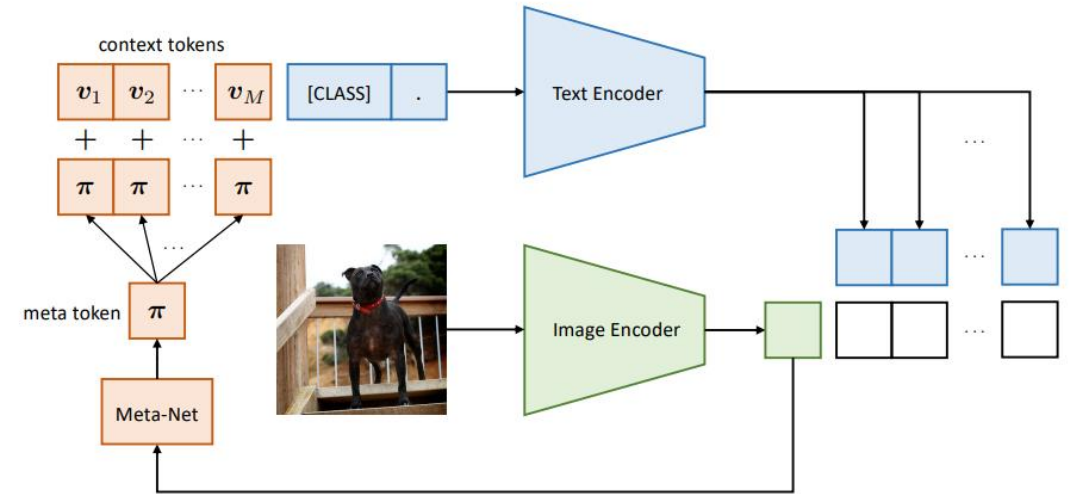


# I . Recap

## (1) Contrastive pre-training

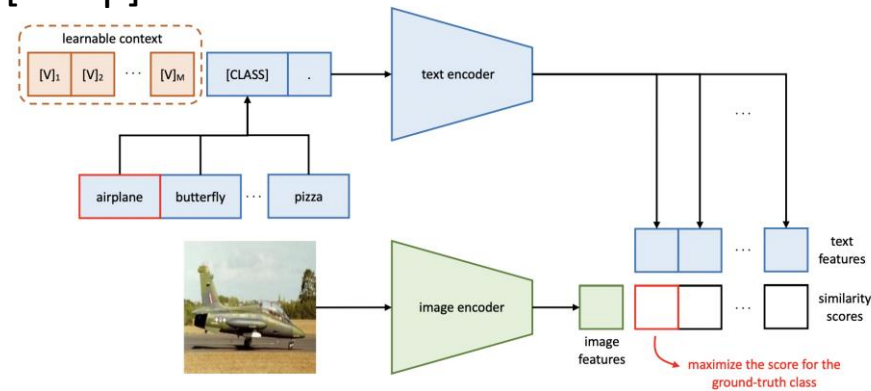


[CLIP]

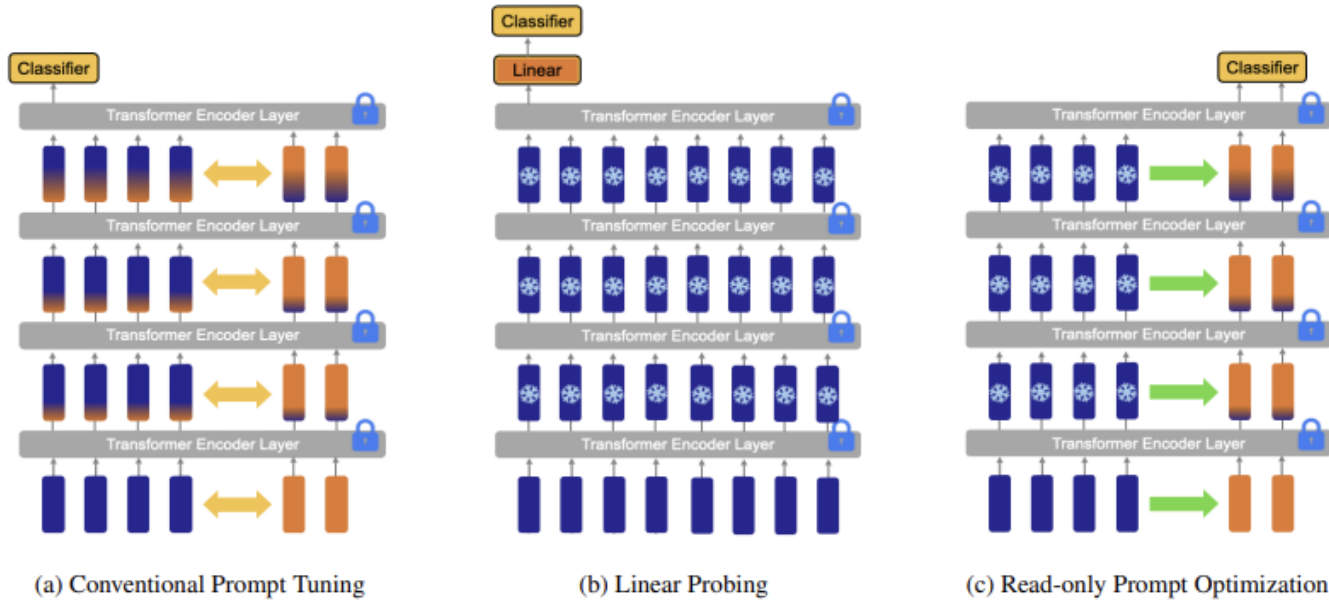


[CoCoOp]

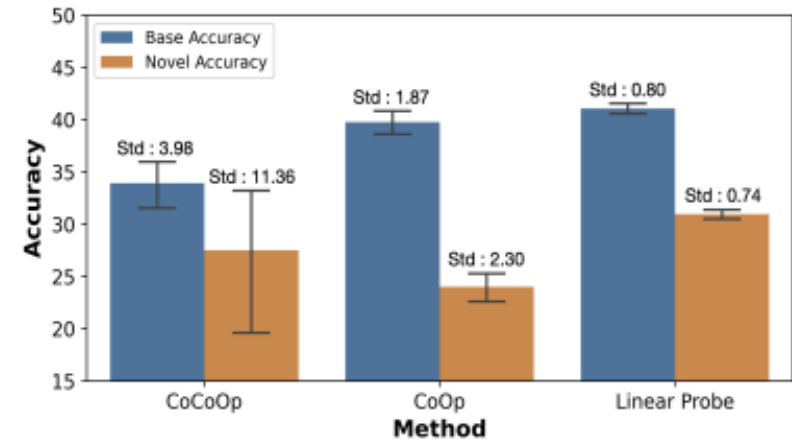
[CoOp]



# I . Recap



“Internal Representation Shift”



**CoOp, CoCoOp:** High Variance

-> may have negatively impact robustness & generalization in data-deficient setting

**Linear probing:** Parameter inefficient(262k), Lack of generalizability in domain-shift task

# I . Recap

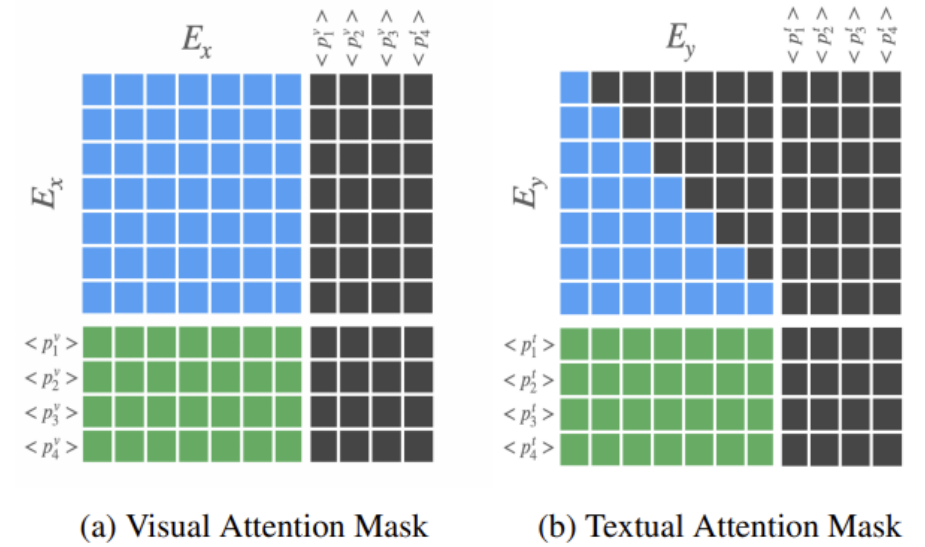
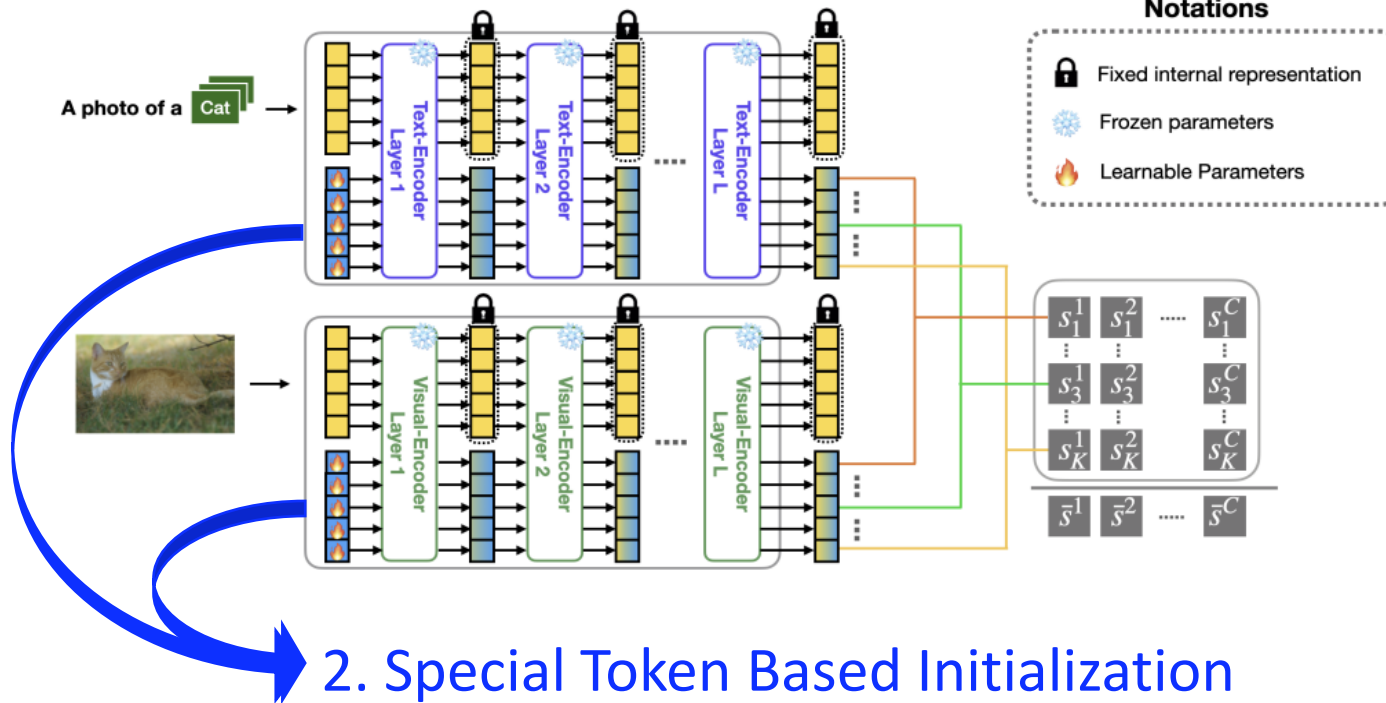


Figure 4: The visualization of attention masks for each encoder.

1. Masked Attention  
-> NO internal representation shift

## II. Experiments

1. K (Num of Prompts) 개수에 따른 성능 변화
2. Epoch에 따른 성능 변화
3. Initial Prompting에 따른 성능 변화
4. Seed에 따른 variance 변화

## II. Experiments

Eurosat[10classes, 27000 images] (sealake, highway, forest..)

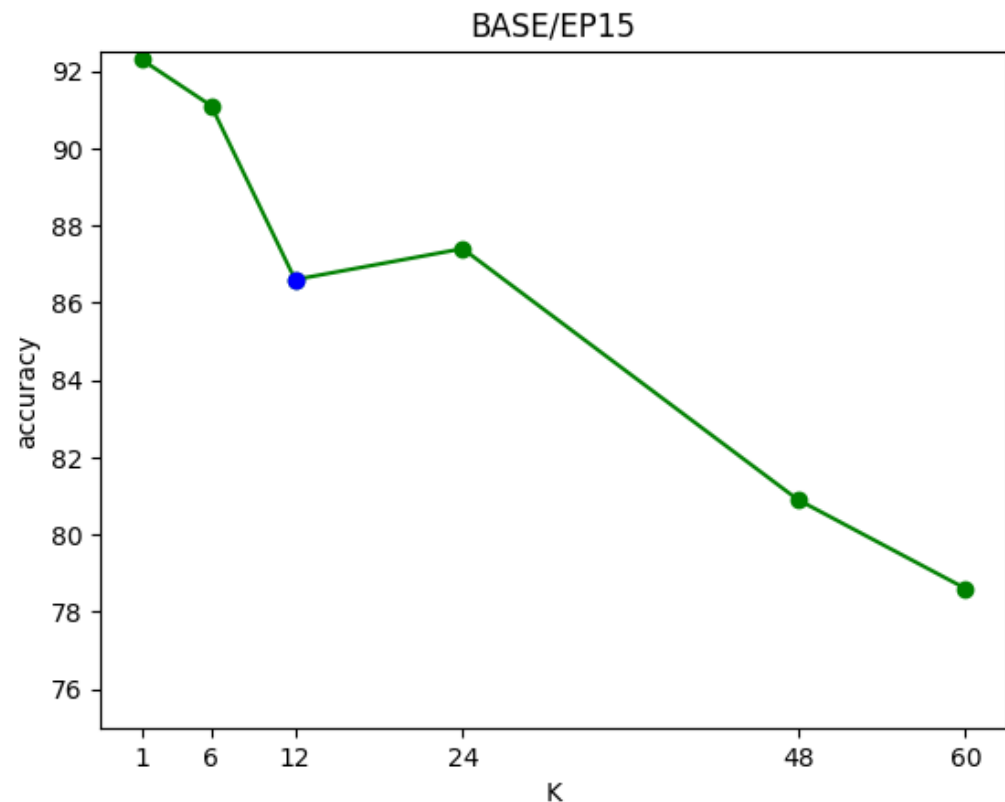
**EuroSAT: Land Use and Land Cover Classification with Sentinel-2**

X16 shots

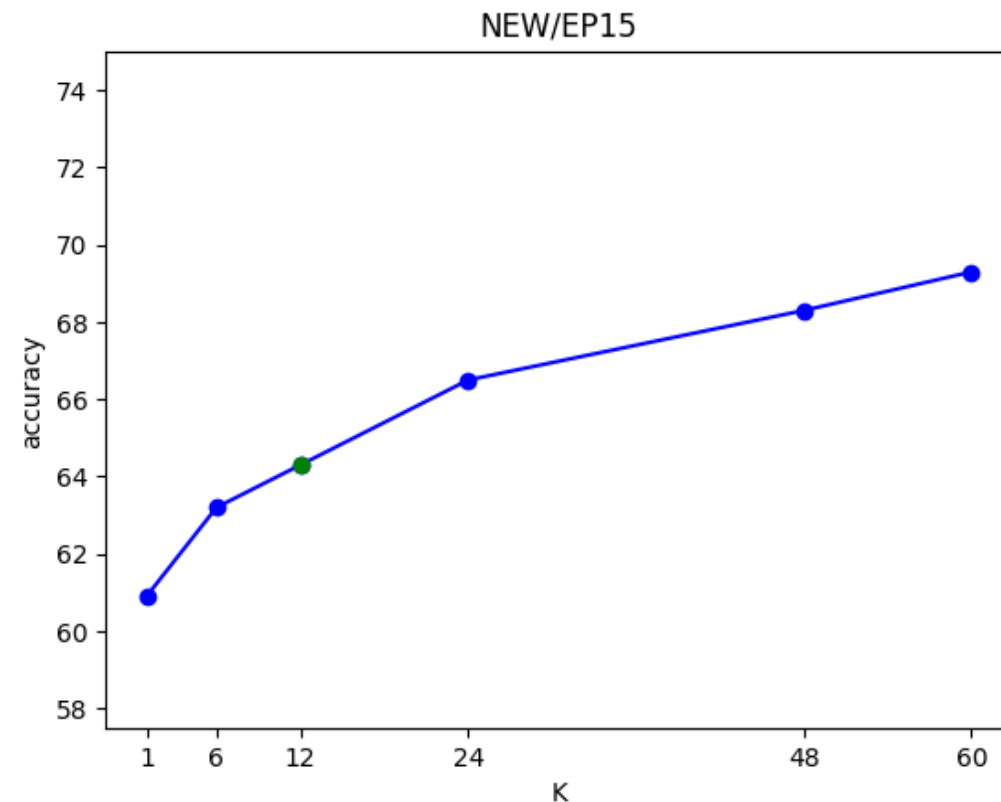
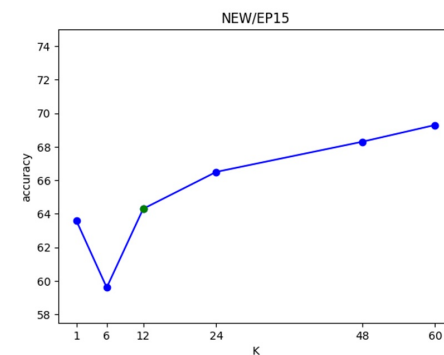
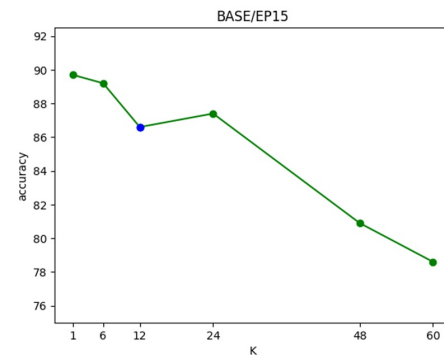


# II. Experiments

## 2.1 K(Num of Prompts) 개수에 따른 성능 변화



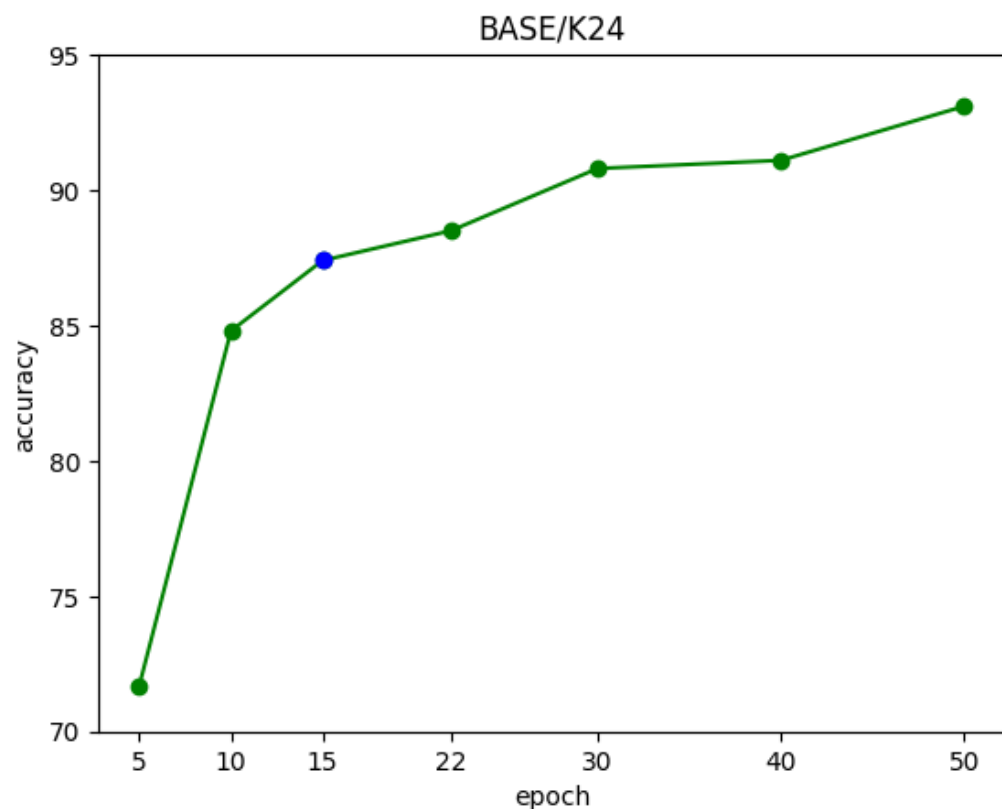
Too many cooks spoil the broth



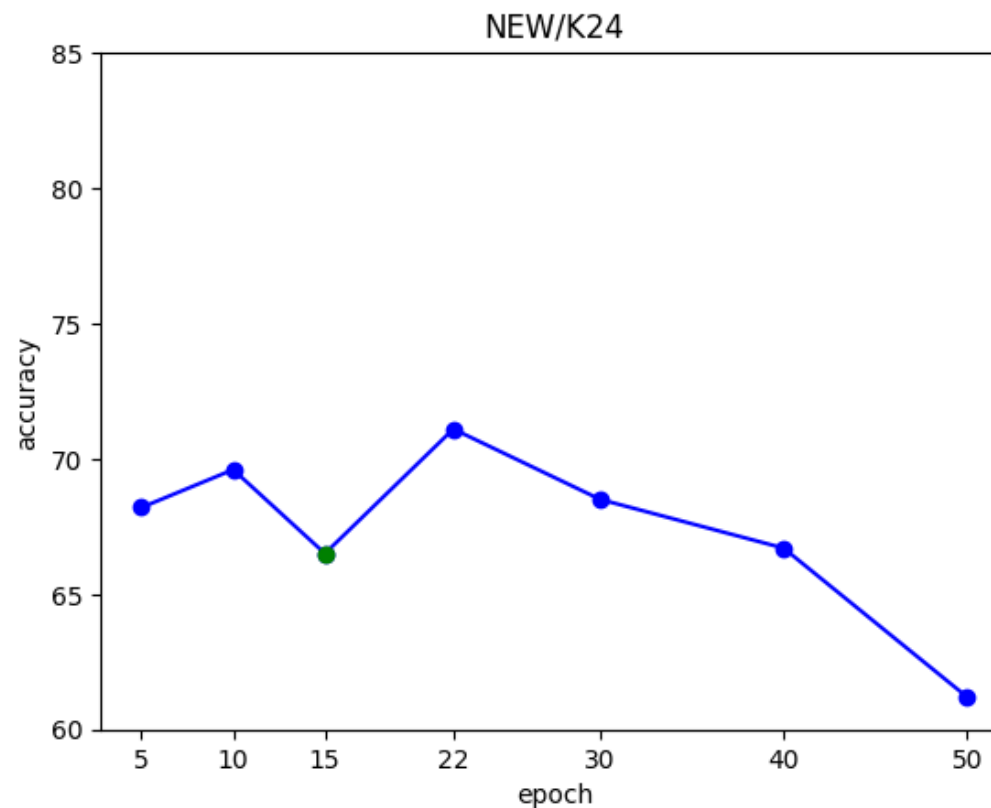
Two heads are better than one

## II. Experiments

### 2.2 Epoch 수에 따른 성능 변화



The more you learn,  
the more intelligent you get




Learning more doesn't always mean  
you get more intelligent



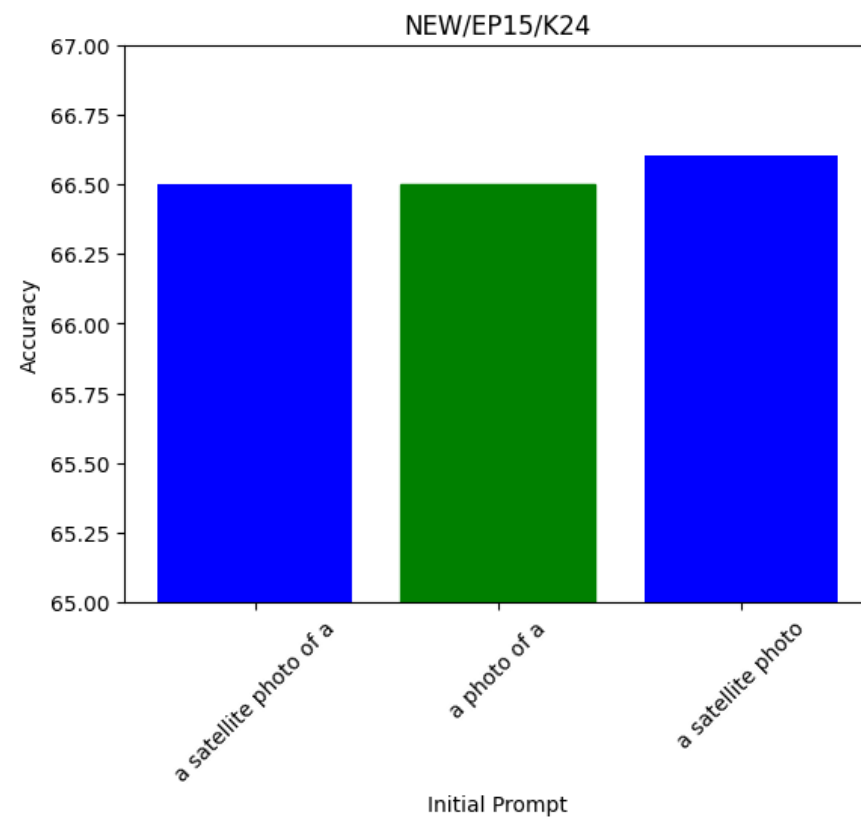
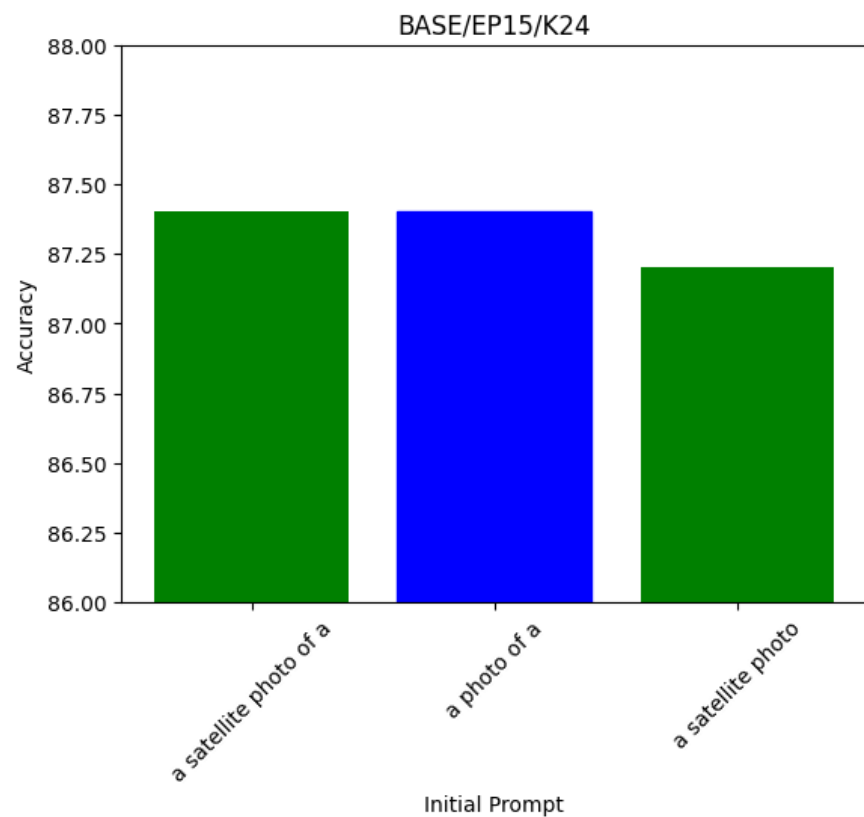
# II. Experiments

## 2.3 Initial Prompt에 따른 성능 변화

EuroSAT	Prompt	Accuracy
	a photo of a [CLASS].	24.17
	a satellite photo of [CLASS].	37.46
	a centered satellite photo of [CLASS].	37.56
	$[V]_1 [V]_2 \dots [V]_M$ [CLASS].	<b>83.53</b>

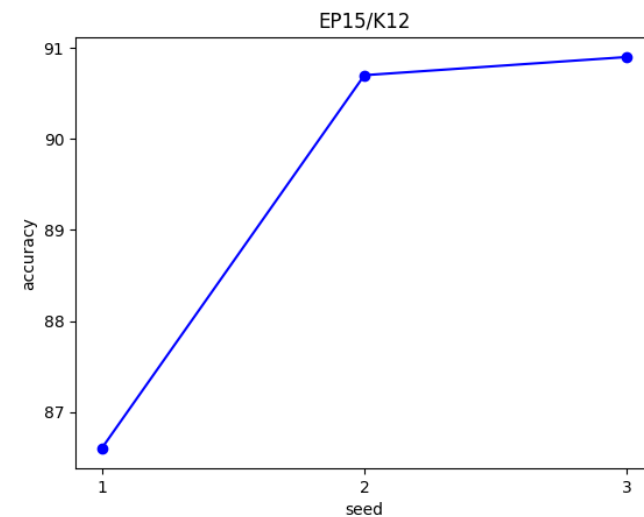
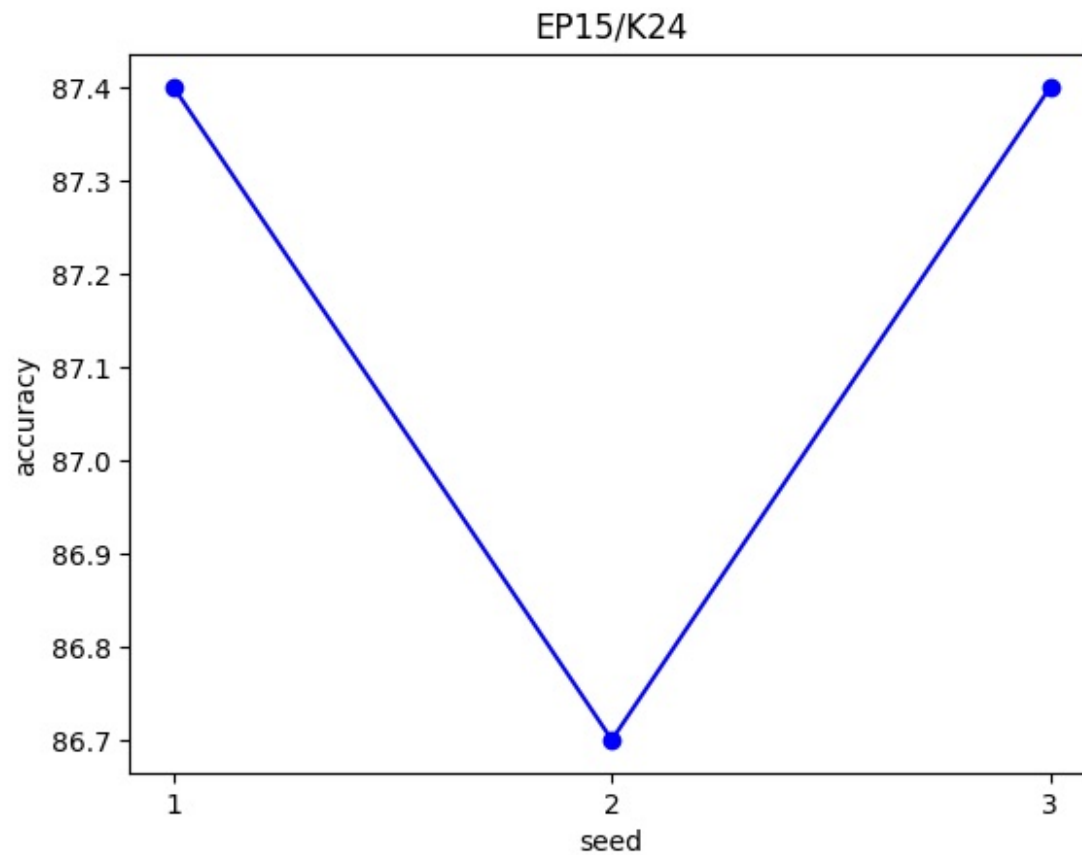
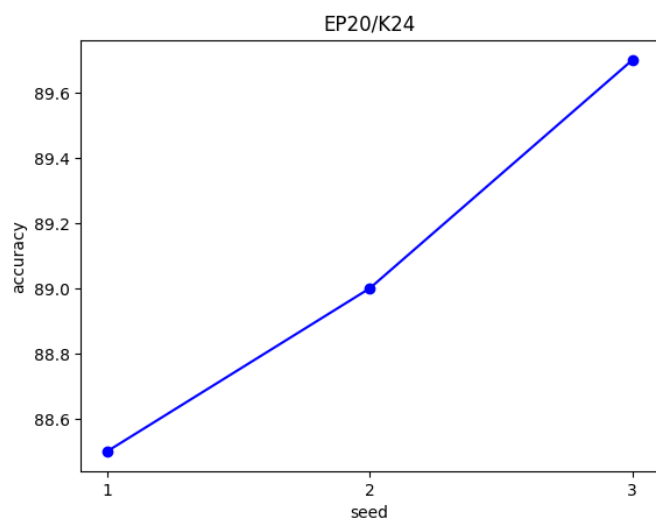
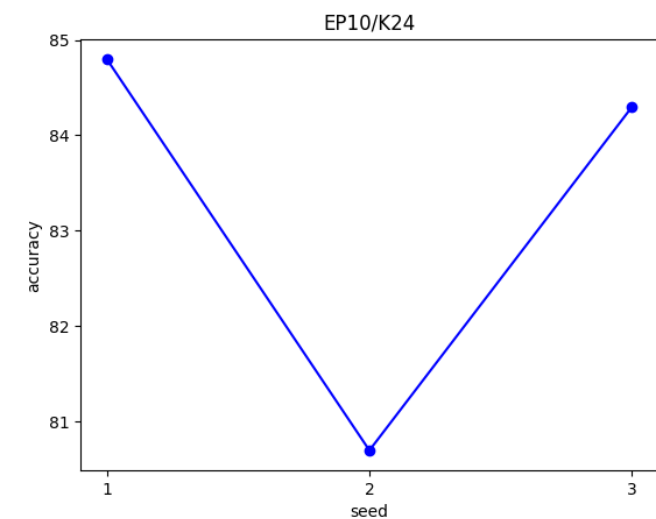
[CoOp]

(d)



# II. Experiments

## 2.4 seed에 따른 variance 변화



# III. Limitations

1. GPU와 Time Limit으로 인해 seed1로 진행  
-> seed 다양화를 통한 variance issue 극복 x
2. Base to New Generalization만 실험  
-> Maple과 같이 Cross Dataset Evaluation을 통한  
'Generalizability'에 관한 성능 평가 다양화의 필요
3. 새로운 Idea의 적용보다는, Analysis에 그침

## IV. Hope Works

1. CoOp과 같이 class별로 Prompt를 달리 하는 specific class와 모든 class에 동일한 prompt를 부여하는 unified class task 진행
2. Image, Text Prompt 다르게 제공 -> 어떤 방식으로 다르게 주어야 할까?  
에 관한 고민 필요

# Thank you!

<https://arxiv.org/pdf/2308.14960v1.pdf>

<https://github.com/mlvlab/RPO>