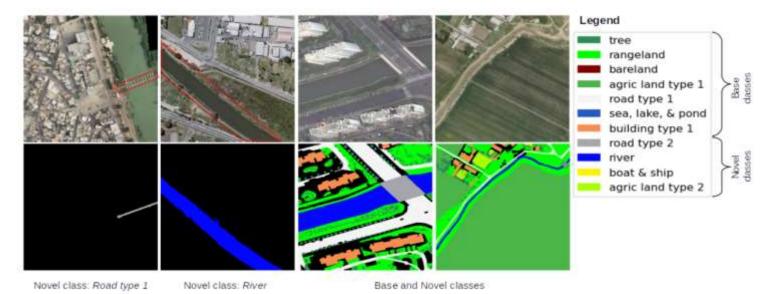
# Dynamic Knowledge Adapter with Probabilistic Calibration for Generalized Few-Shot Semantic Segmentation

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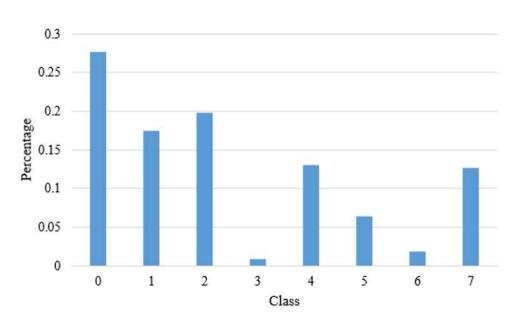
#### Generalized Few-Shot Semantic Segmentation (GFSS)

- Setting
  - Trained on base classes
  - Fed few-shot novel classes
- □ Task
  - Segment for all classes (including base and novel classes)
- □ Key
  - Base-class performance + novel-class generalizability



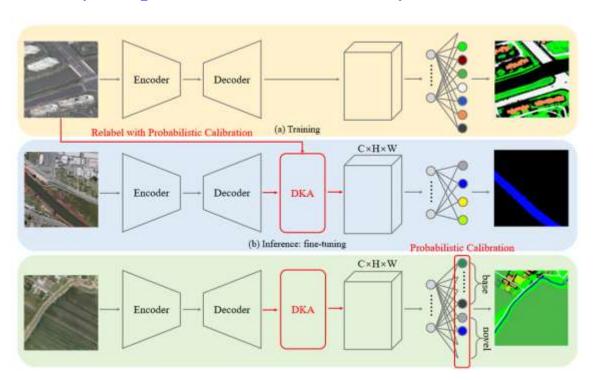
## Dilemma

- ☐ The lack of adaptability to learn novel classes
- ☐ The catastrophic forgetting of base classes
- ☐ The biased prediction of imbalanced classes



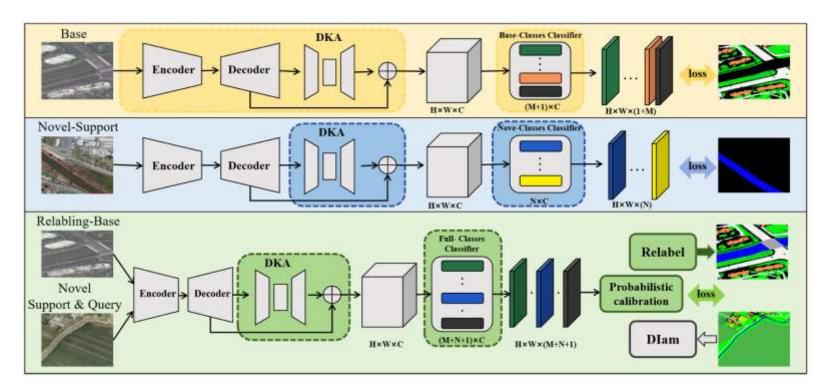
### **Preliminaries**

- Dynamic Knowledge Adapter (DKA)
  - selecting only efficient parameters for finetuning
  - Relabeling: addresses the stability problem
- Probabilistic Calibration Module (PCM)
  - rectify the prediction bias caused by imbalanced data

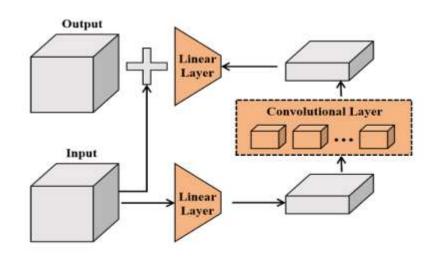


## Method framework

- ☐ Two-stage training paradigm (training-inference)
  - Training: utilize the base class to train encoder, decoder, DKA
  - Finetuning: fix encoder and decoder
    - ☐ Fine-tuning DKA and novel classifier with novel classes
    - Concatenating the novel-class and base-class classifiers initializes a full-classes classifier
    - Randomly sample and relabel base class training samples



## Dynamic Knowledge Adapter

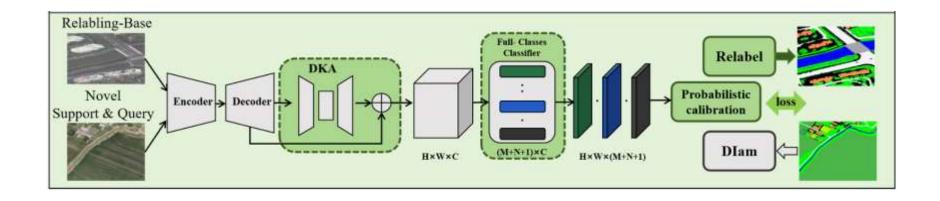


DKA structure

$$W_{\text{in}} \in \mathbb{R}^{C \times r}$$
,  $W_{\text{tran}} \in \mathbb{R}^{r \times r \times 1 \times 1}$   $W_{\text{out}} \in \mathbb{R}^{r \times C}$ 

- Relabelling
  - $\blacksquare$  sample k  $\times$  10 images from the training set
  - k is the total number of categories

## Calibrating Probabilistic



- ☐ Alleviate the probabilistic bias
  - $P = softmax(\tau f_{DKA}(f_{\phi}(X))\theta_n)$

τ : hyper-parameter

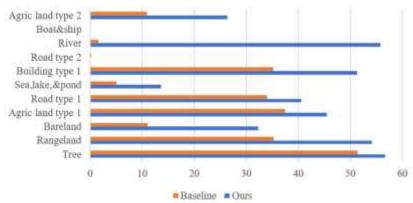
Mitigate the model's bias towards the background

$$P_{new} = (1 - M) \odot P + \gamma M$$

## Experiments

#### ☐ Comparison With Baseline

Method	Base	Novel	Weighted average
Baseline	29.89401	3.15314	13.84949
Ours	42.00045	20.51581	29.10967



#### ☐ Ablation Study

DKA	Relabeling	Probabilistic	boso	novel	Weighted
		Calibration	base		Average
X	X	X	29.89	3.15	13.85
y	X	X	37.20	8.49	19.97
y	У	X	37.93	10.93	21.73
у	У	У	42.00	20.52	29.11

## Thanks!