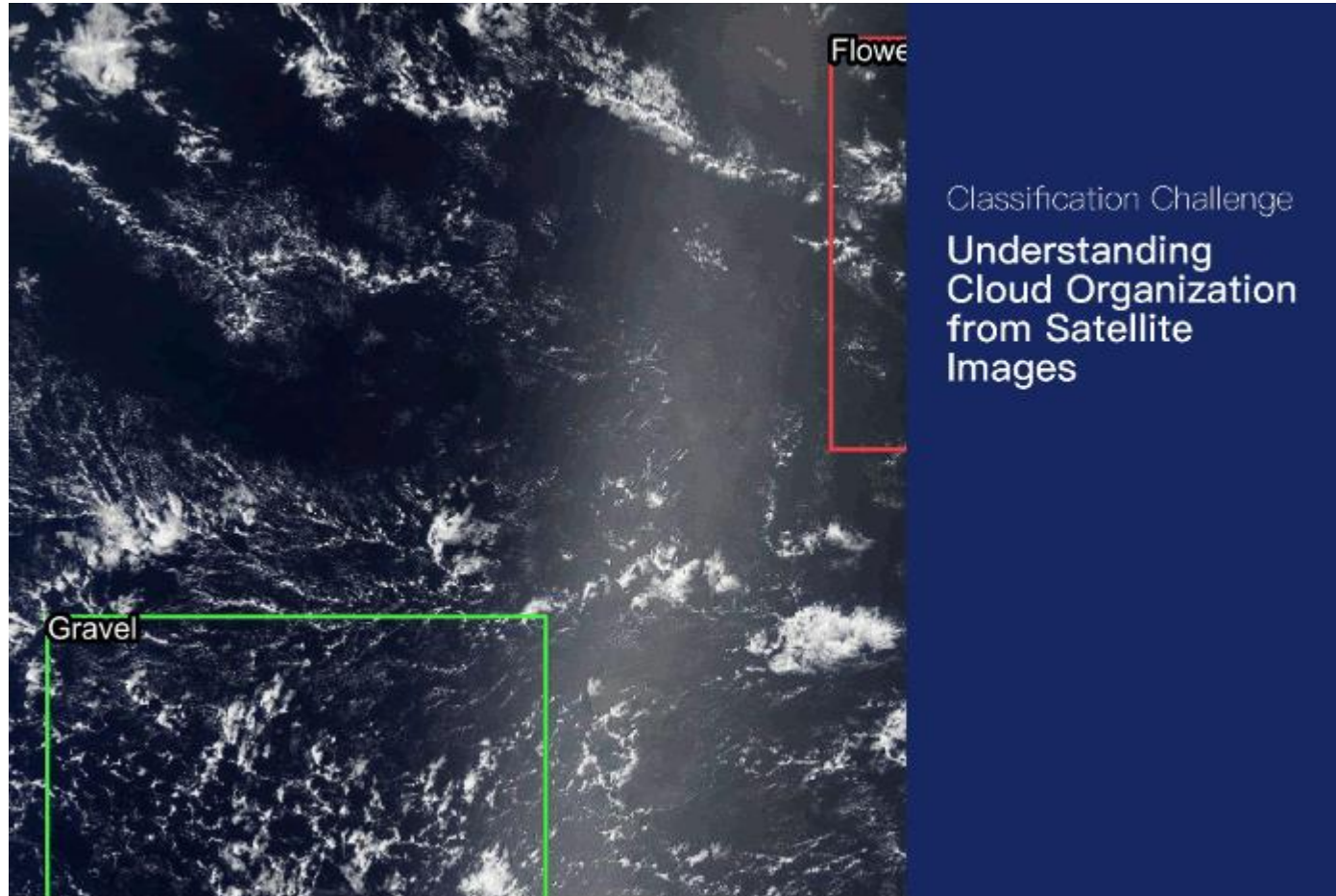


# 1. Overview

(1) 대회 소개

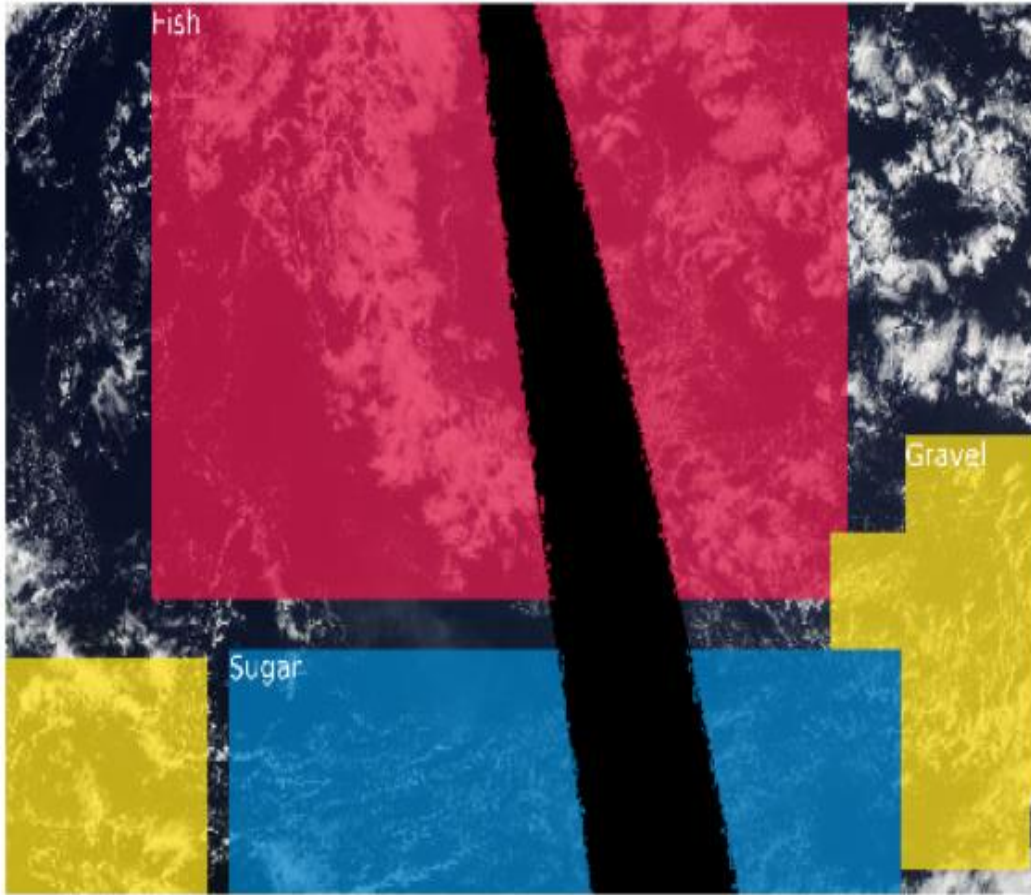


목적 : Help us remove the haze from climate models and bring clarity to cloud identification.

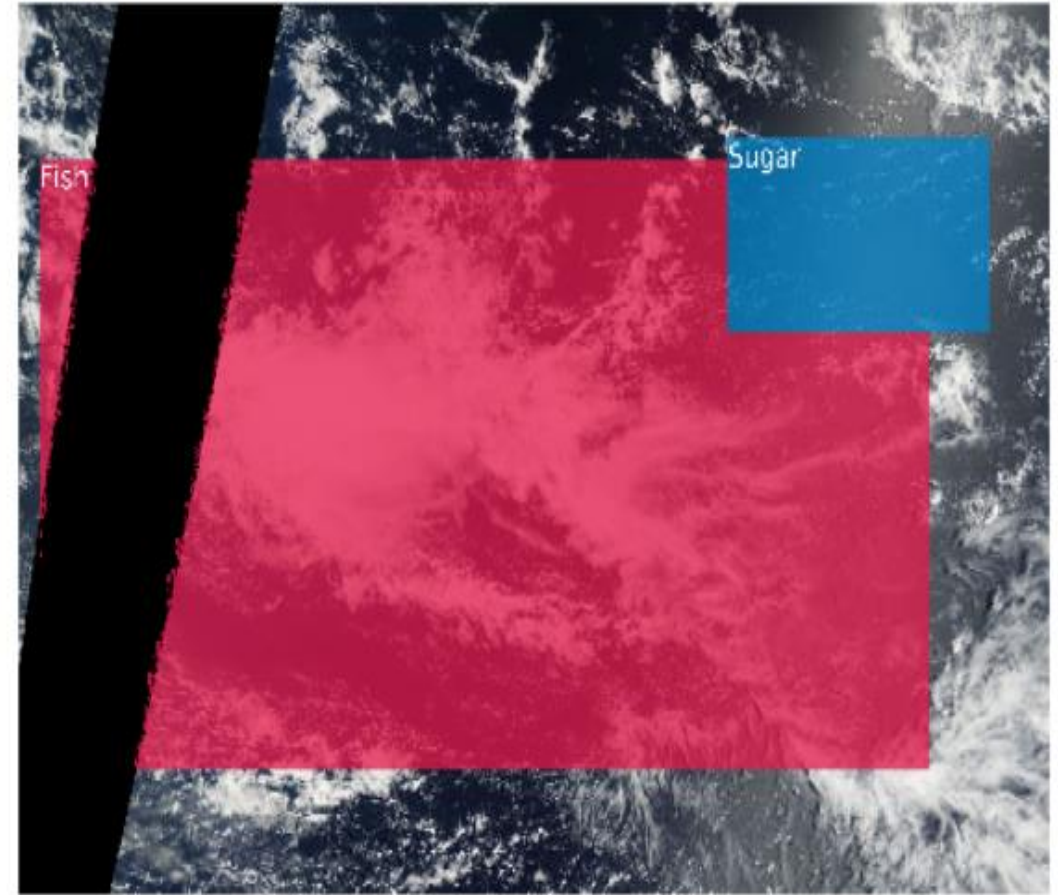
# 1. Overview

• (2) 입력 데이터

Fish Gravel Sugar



Fish Sugar



# 2. Solution

## (1) 1<sup>st</sup> Solution Summary

1<sup>st</sup> solution

### Network

- Model A: Unet with classification head
- Model B: FPN or Unet, no classification head

### Backbones

- Resnet34
- Efficientnet-b1
- Resnext101\_32x8d\_wsl
- Resnext101\_32x16d\_wsl

# 2. Solution

## (1) 1<sup>st</sup> Solution Summary

1<sup>st</sup> solution

### Loss

- Classification part: BCE
- Segmentation part:  $\text{BCE} * 0.75 + \text{DICE} * 0.25$

### Optimizer

- AdamW, weight decay 0.01
- encoder learning rate 0.000025
- decoder learning rate 0.00025
- OneCycle scheduler, shallow models 30 epochs, models 15 epochs

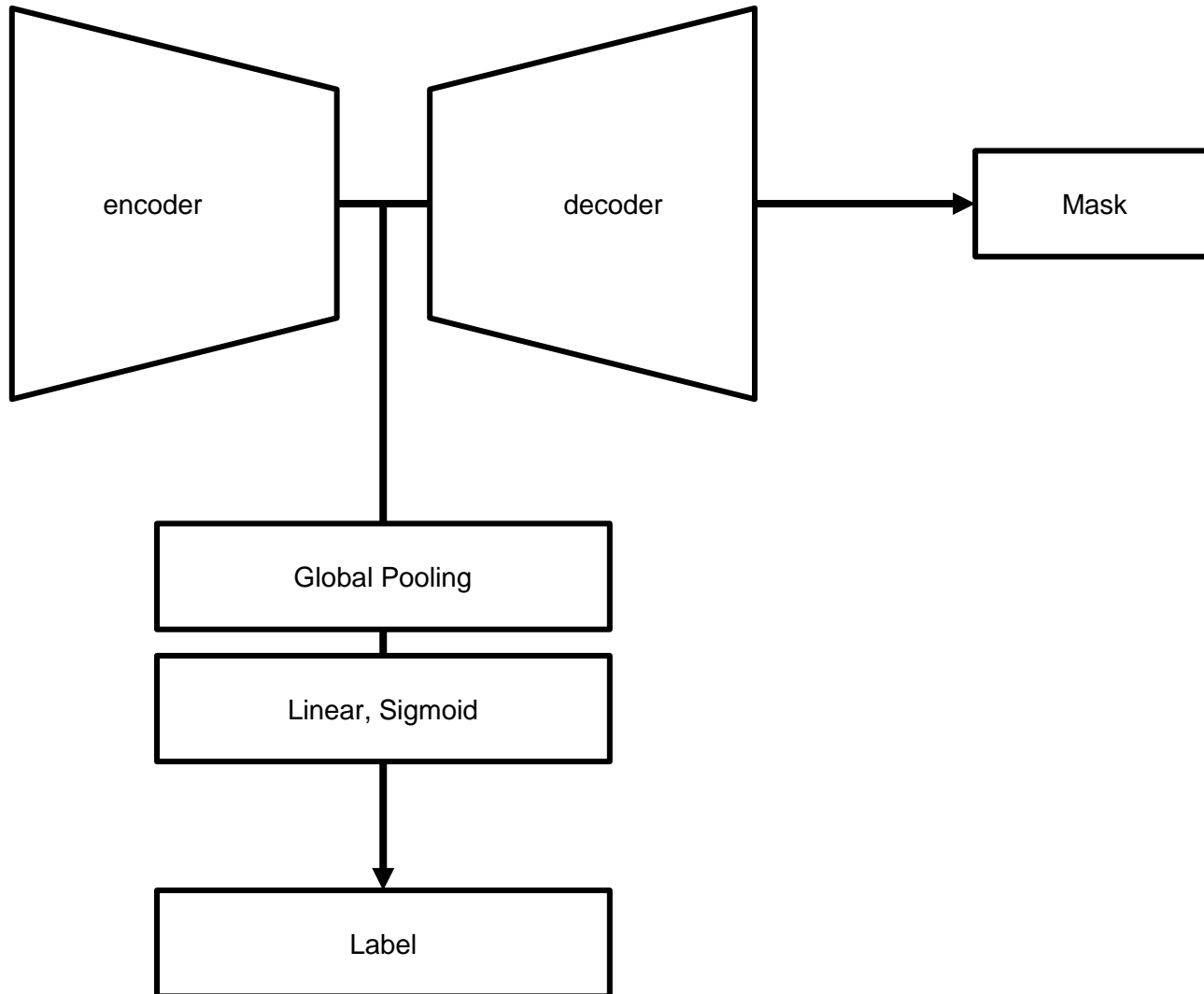
### Augmentation

- Common: hflip, vflip, shift/scale/rotate, gid distortion, channel shuffle, invert, to gray
- Modal A: random crop, size 384
- Model B: full-size, size 384, 544, 576, 768

# 2. Solution

(1) 1<sup>st</sup> Solution Summary

1<sup>st</sup> soltuion



# 2. Solution

## (1) 1<sup>st</sup> Solution Summary

1<sup>st</sup> soltuion

Other

- 추가적으로 Classification head와는 별도로 probability를 활용하여 segmentation을 classifier로 사용하였다.
- K개의 top probability pixel을 classification probability로 사용하였다. (ex K=17500)

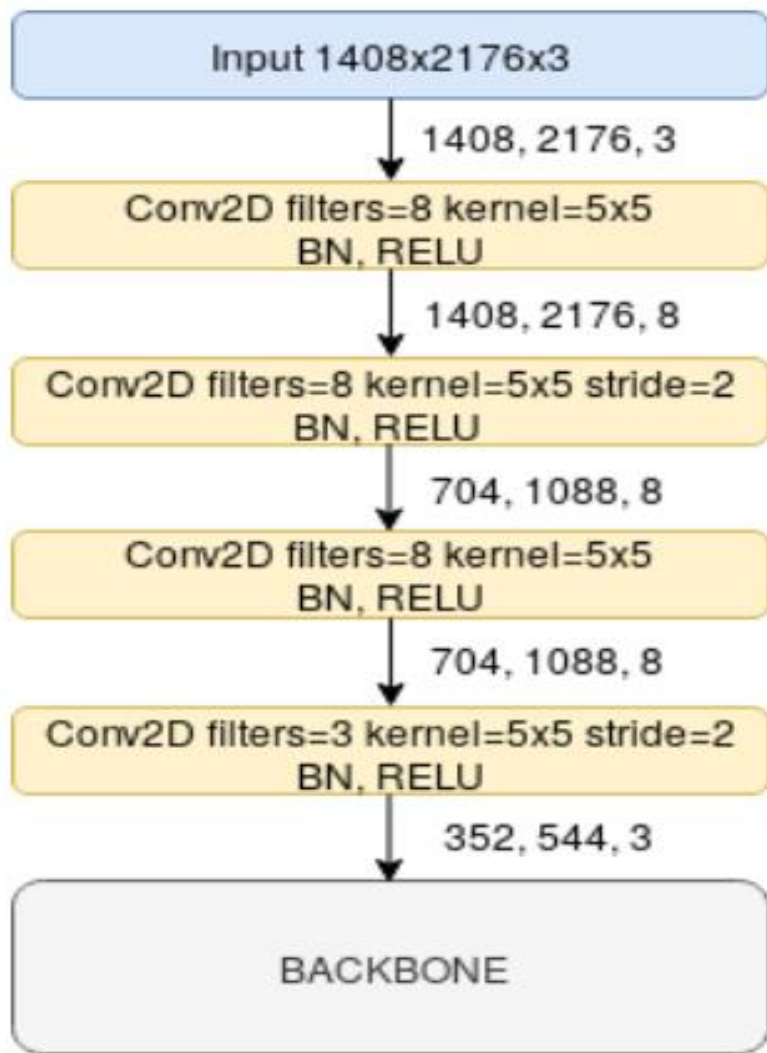
```
cls_probabilities = np.sort(mask_probabilities.reshape(4, -1), axis=1)
cls_probabilities = np.mean(cls_probabilities[:, -17500:], axis=1)
```

- Max probability를 positive prediction으로 사용
- 이미지당 적어도 1개 이상의 객체가 있으므로, max probability를 이용하여, 적어도 1개 이상의 물체를 탐지하도록 하였다.

```
cls_probabilities[np.argmax(cls_probabilities)] = 1
```

# 2. Solution

## (1) 2<sup>nd</sup> Solution Summary



- 이미지를 resize (compress) 할 때 conv layer를 사용

2<sup>nd</sup> solution



# 2. Solution

(1) 2<sup>nd</sup> Solution Summary

2<sup>nd</sup> soltuion

	Backbone		Input image dim		Image preprocess	
	efficientnetb1	seresnext50	1408 x2176	1280 x 1920	None	CLAHE
m1	✓		✓		✓	
m2	✓		✓			✓
m3	✓			✓	✓	
m4	✓			✓		✓
m5		✓	✓		✓	
m6		✓	✓			✓
m7		✓		✓	✓	
m8		✓		✓		✓



# 2. Solution

## (1) 2<sup>nd</sup> Solution Summary

2<sup>nd</sup> solution

Triplet scheme (top\_score\_threshold, min\_contour\_area, bottom\_score\_threshold)

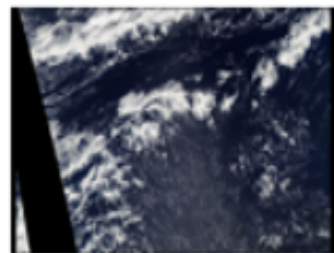
```
classification_mask = predicted > top_score_threshold
mask = predicted.copy()
mask[classification_mask.sum(axis=(1,2,3)) < min_contour_area, :, :, :] =
np.zeros_like(predicted[0])
mask = mask > bot_score_threshold
return mask
```

# 2. Solution

(1) 6<sup>th</sup> Solution Summary

6<sup>th</sup> soltuion

## 1st-stage training(pre-training)

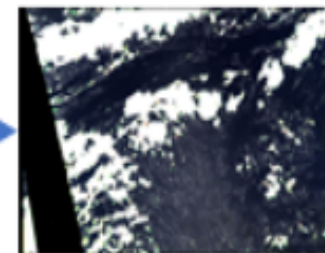


Resize(350, 525),  
PadIfNeeded(352,544, border\_mode=0)  
ShiftScaleRotate(0.5,0, 0, border\_mode=0)  
HorizontalFlip(),  
VerticalFlip()

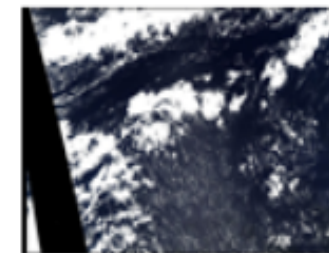
EfficientNet-B4

Decoder  
(fpn or unet)

Prediction



Label



Classifier  
(nn.Linear)

Loss  
(BCE\*0.2 + DICE\*0.8)

# 2. Solution

(1) 6<sup>th</sup> Solution Summary

6<sup>th</sup> soltuion

