



딥러닝 분반 (CV 프로젝트)

# Human Protein Atlas Image Classification

13기 최해윤 · 14기 김종민 · 14기 이우준 · 14기 최대원

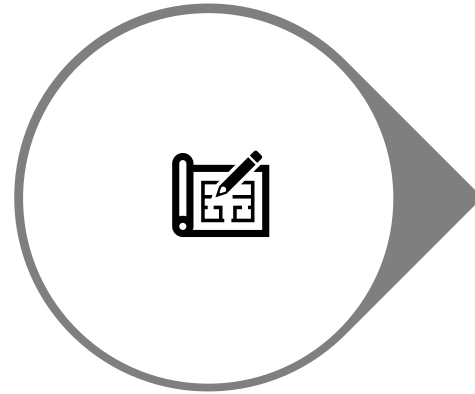


## 1. 주제 및 배경지식



## 2. 데이터

- EDA
- 전처리 및 최종 데이터



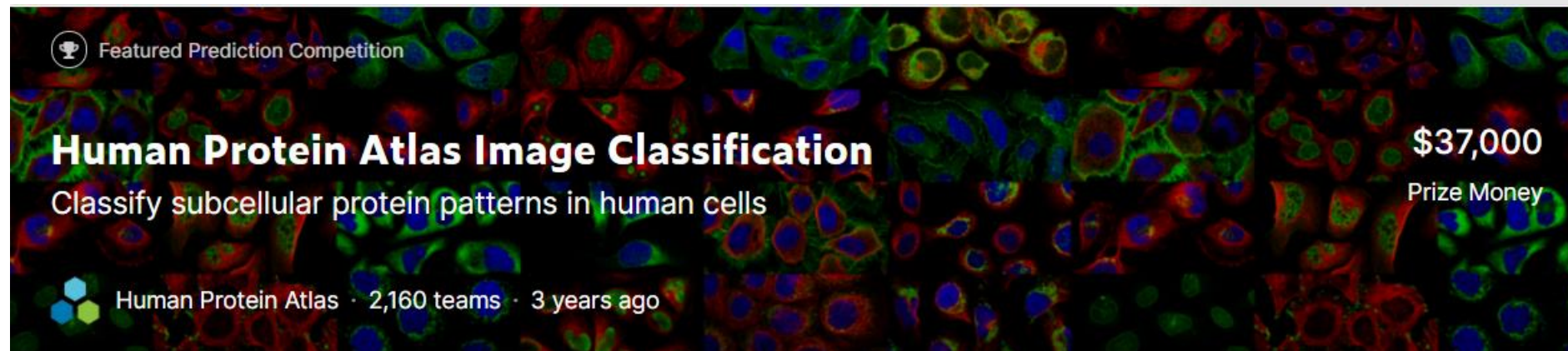
## 3. 모델

- 모델 간 비교
- Filter 조절(InceptionResNetV2)



## 4. 결과 및 결론

- 최종 모델 submission
- Discussion



**목적:** microscope image에서 mixed patterns of protein을 분류

- |                              |                                   |
|------------------------------|-----------------------------------|
| 0. Nucleoplasm               | 14. Microtubules                  |
| 1. Nuclear membrane          | 15. Microtubule ends              |
| 2. Nucleoli                  | 16. Cytokinetic bridge            |
| 3. Nucleoli fibrillar center | 17. Mitotic spindle               |
| 4. Nuclear speckles          | 18. Microtubule organizing center |
| 5. Nuclear bodies            | 19. Centrosome                    |
| 6. Endoplasmic reticulum     | 20. Lipid droplets                |
| 7. Golgi apparatus           | 21. Plasma membrane               |
| 8. Peroxisomes               | 22. Cell junctions                |
| 9. Endosomes                 | 23. Mitochondria                  |
| 10. Lysosomes                | 24. Aggresome                     |
| 11. Intermediate filaments   | 25. Cytosol                       |
| 12. Actin filaments          | 26. Cytoplasmic bodies            |
| 13. Focal adhesion sites     | 27. Rods & rings                  |

\* 28가지의 단백질 존재

**Macro-F1:**

is the harmonic mean between precision and recall, where the average is calculated per label and then averaged across all labels. If  $p_j$  and  $r_j$  are the precision and recall for all  $\lambda_j \in h(x_i)$  from  $\lambda_j \in y_i$ , the macro-F1 is

$$\text{Macro-F1} = \frac{1}{Q} \sum_{j=1}^Q \frac{2 \times p_j \times r_j}{p_j + r_j}$$

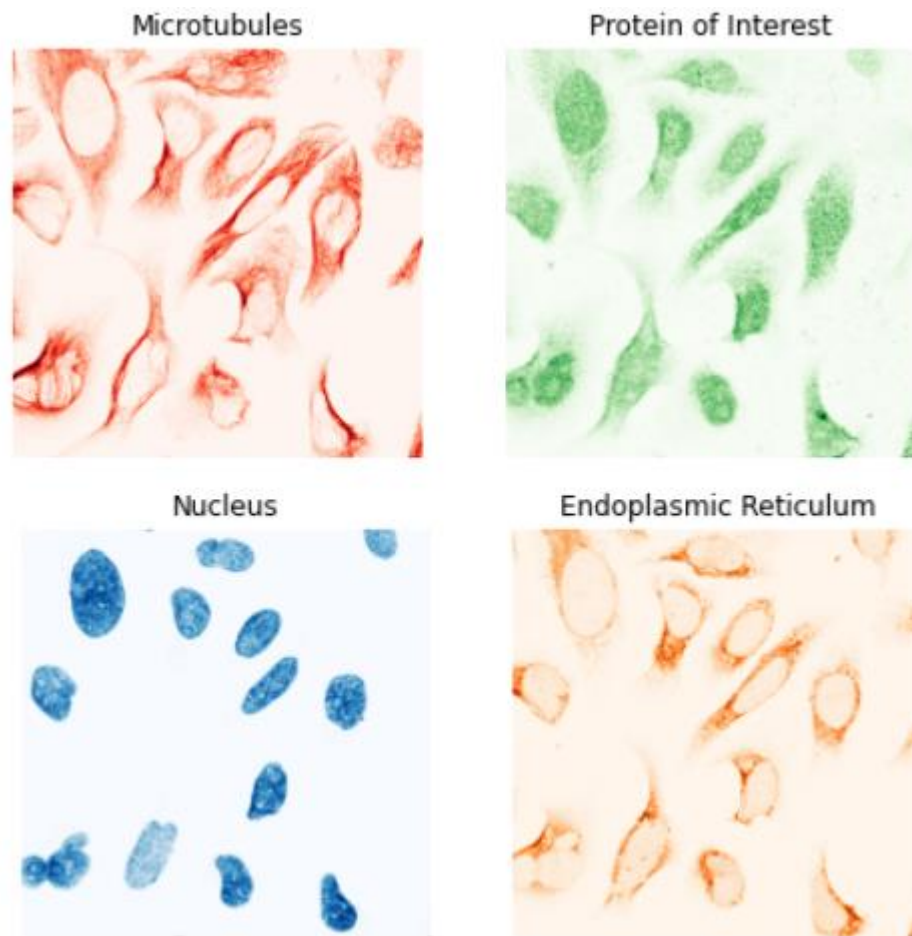


- 31,072개의 train data

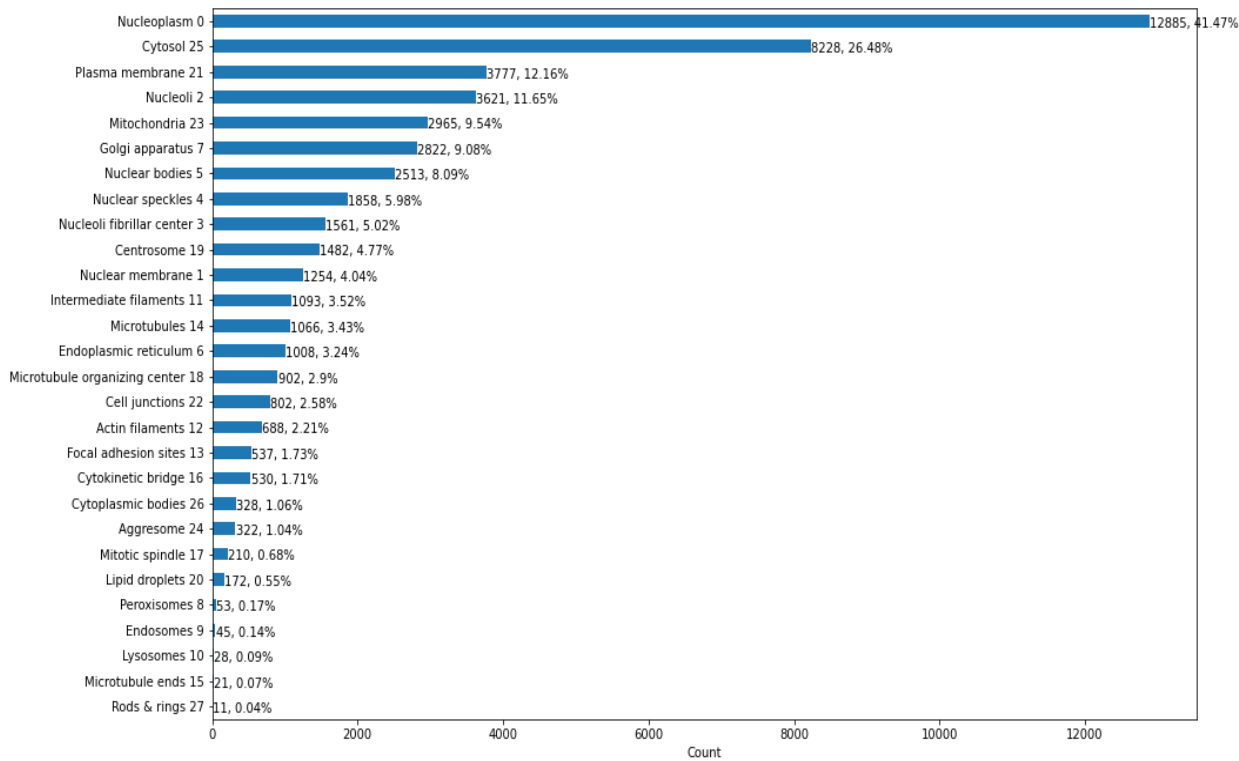
|       | Id                                   | Target  |
|-------|--------------------------------------|---------|
| 0     | 00070df0-bbc3-11e8-b2bc-ac1f6b6435d0 | 16 0    |
| 1     | 000a6c98-bb9b-11e8-b2b9-ac1f6b6435d0 | 7 1 2 0 |
| 2     | 000a9596-bbc4-11e8-b2bc-ac1f6b6435d0 | 5       |
| 3     | 000c99ba-bba4-11e8-b2b9-ac1f6b6435d0 | 1       |
| 4     | 001838f8-bbca-11e8-b2bc-ac1f6b6435d0 | 18      |
| ...   | ...                                  | ...     |
| 31067 | ffed4430-bbac-11e8-b2ba-ac1f6b6435d0 | 21      |
| 31068 | fff0a998-bbae-11e8-b2ba-ac1f6b6435d0 | 5       |
| 31069 | fff189d8-bbab-11e8-b2ba-ac1f6b6435d0 | 7       |
| 31070 | fffd7e0-bbc4-11e8-b2bc-ac1f6b6435d0  | 25 2 21 |
| 31071 | fffe0ffe-bbc0-11e8-b2bb-ac1f6b6435d0 | 2 0     |

31072 rows × 2 columns

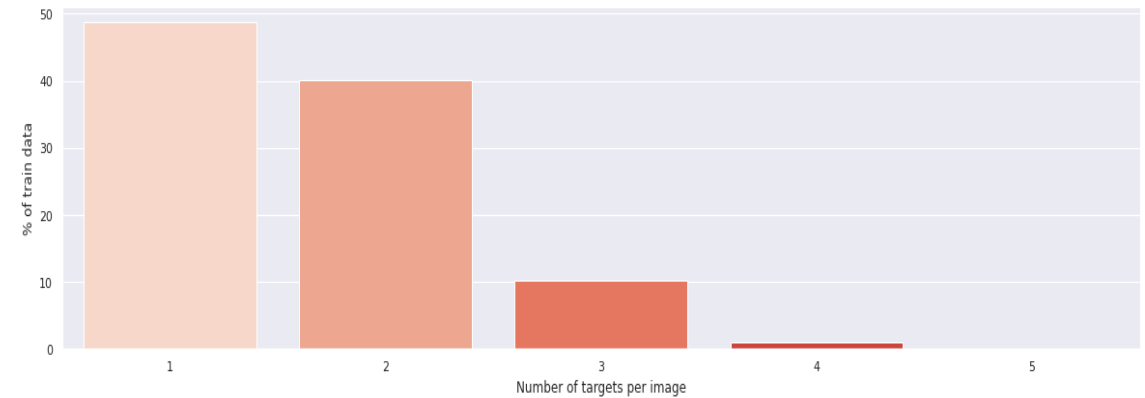
각 데이터는 R,G,B,Y 4개의 이미지로 구성  
(green: protein of interest, blue: 핵, red: 미세관, yellow: 소포내막)



- 단백질 빈도

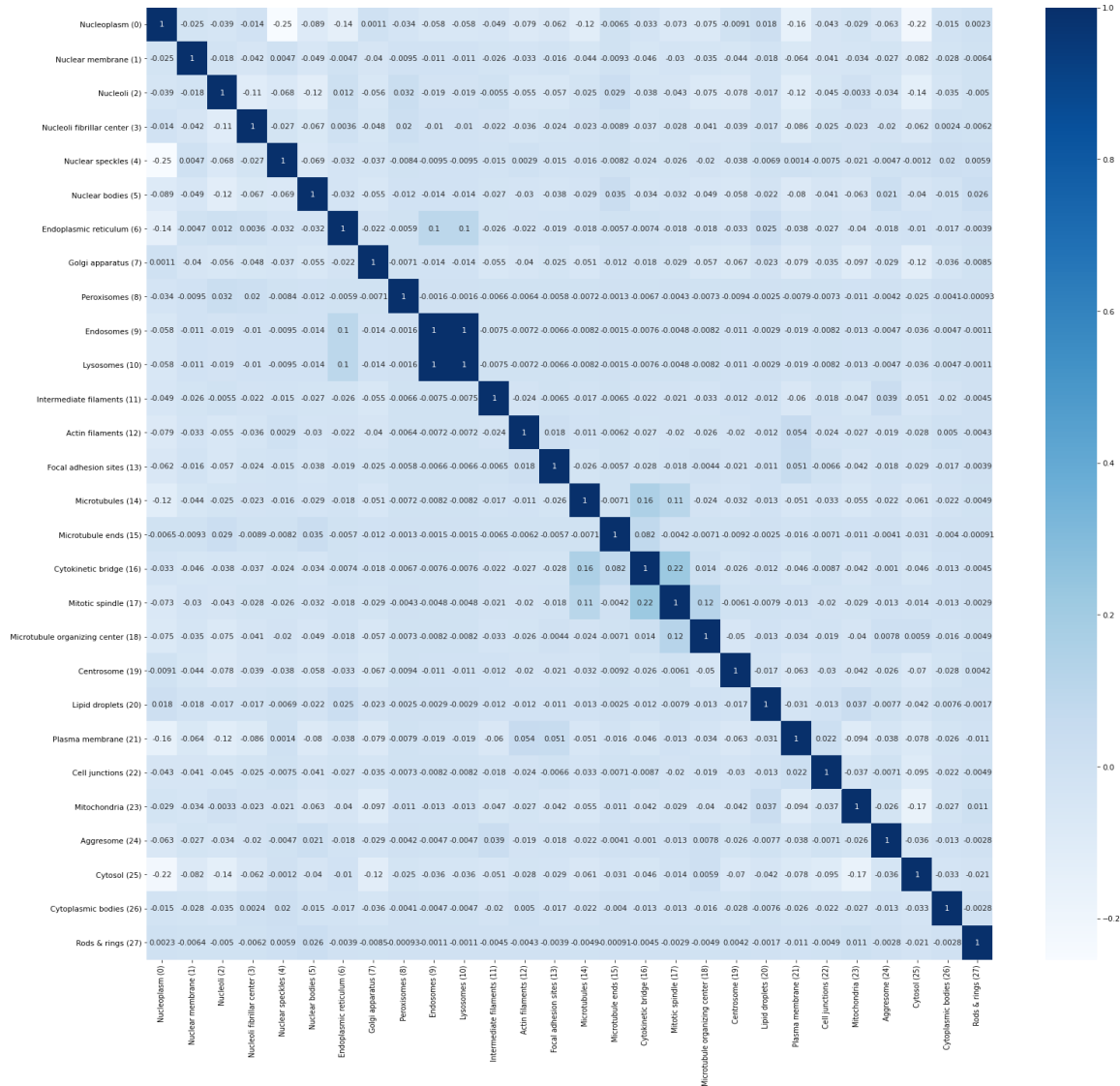


- 데이터 imbalance 문제

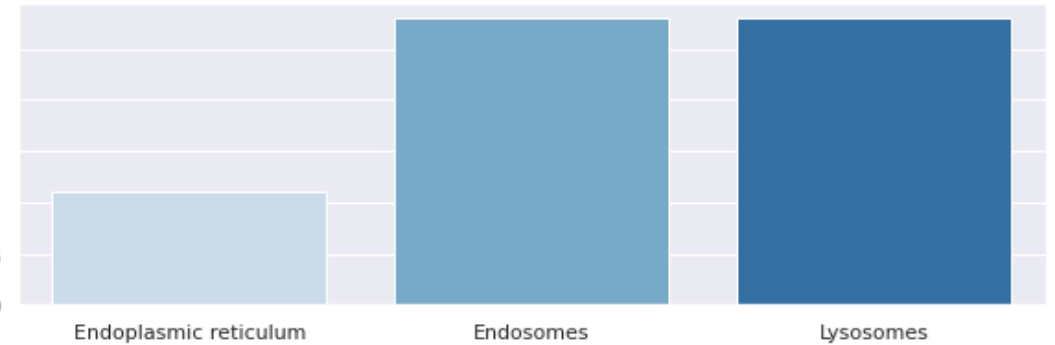


|   | image_count | pct_of_dataset |
|---|-------------|----------------|
| 1 | 15126       | 48.680484      |
| 2 | 12485       | 40.180870      |
| 3 | 3160        | 10.169928      |
| 4 | 299         | 0.962281       |
| 5 | 2           | 0.006437       |

- 단백질 간 Correlation



Counts in train data



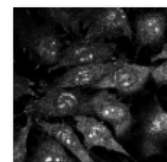
Lysomes & Endosomes  
상관관계가 높음

Lysomes & Endosomes은 항상 동시 관찰

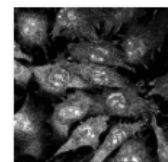


Albumentations Library 이용하여 **데이터 증강**

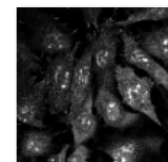
```
def strong_aug(p=1):  
    return Compose([  
        RandomRotate90(),  
        Flip(),  
        Transpose(),  
        OneOf([  
            IAAGaussianNoise(),  
            GaussNoise(),  
        ], p=0.2),  
        OneOf([  
            MotionBlur(p=.2),  
            MedianBlur(blur_limit=3, p=.1),  
            Blur(blur_limit=3, p=.1),  
        ], p=0.2),  
        ShiftScaleRotate(shift_limit=0.0625, scale_limit=0.2,  
                           rotate_limit=45, p=.2),  
        OneOf([  
            OpticalDistortion(p=0.3),  
            GridDistortion(p=.1),  
            IAAPiecewiseAffine(p=0.3),  
        ], p=0.2),  
        OneOf([  
            CLAHE(clip_limit=2),  
            IAASharpen(),  
            IAAEmboss(),  
            RandomContrast(),  
            RandomBrightness(),  
        ], p=0.3),  
    ], p=p)
```



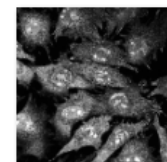
183d51d4-bbc5-  
11e8-b2bc-ac1f6  
b6435d0\_green



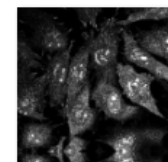
target\_5\_augmen  
tedaug\_04241\_18  
3d51d4-bbc5-11  
e8-b2bc-ac1f6b...



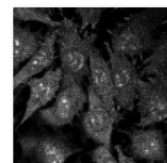
target\_5\_augmen  
tedaug\_04243\_18  
3d51d4-bbc5-11  
e8-b2bc-ac1f6b...



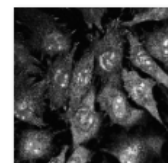
target\_5\_augmen  
tedaug\_04245\_18  
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e8-b2bc-ac1f6b...



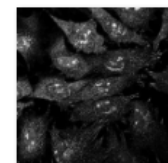
target\_5\_augmen  
tedaug\_04247\_18  
3d51d4-bbc5-11  
e8-b2bc-ac1f6b...



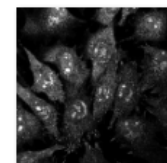
target\_5\_augmen  
tedaug\_04249\_18  
3d51d4-bbc5-11  
e8-b2bc-ac1f6b...



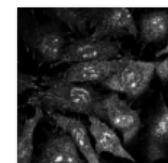
target\_5\_augmen  
tedaug\_04251\_18  
3d51d4-bbc5-11  
e8-b2bc-ac1f6b...



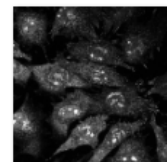
target\_5\_augmen  
tedaug\_04253\_18  
3d51d4-bbc5-11  
e8-b2bc-ac1f6b...



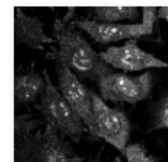
target\_5\_augmen  
tedaug\_04255\_18  
3d51d4-bbc5-11  
e8-b2bc-ac1f6b...



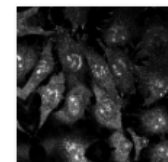
target\_5\_augmen  
tedaug\_04257\_18  
3d51d4-bbc5-11  
e8-b2bc-ac1f6b...



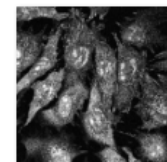
target\_5\_augmen  
tedaug\_04259\_18  
3d51d4-bbc5-11  
e8-b2bc-ac1f6b...



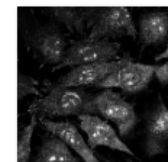
target\_5\_augmen  
tedaug\_04261\_18  
3d51d4-bbc5-11  
e8-b2bc-ac1f6b...



target\_5\_augmen  
tedaug\_04263\_18  
3d51d4-bbc5-11  
e8-b2bc-ac1f6b...



target\_5\_augmen  
tedaug\_04265\_18  
3d51d4-bbc5-11  
e8-b2bc-ac1f6b...



target\_5\_augmen  
tedaug\_04267\_18  
3d51d4-bbc5-11  
e8-b2bc-ac1f6b...

(기존 31072개 데이터 + 11539개 데이터)

## 단백질을 딥러닝 모델을 통해 예측

### HOW?

- ① Fine-tuning 기법을 통한 모델링 -> 여러가지 모델을 훈련, 평가 후 비교
- ② 데이터가 R,G,B,Y 로 주어져 있음-> 채널 (channel)을 여러가지 방식으로 구성 후 비교



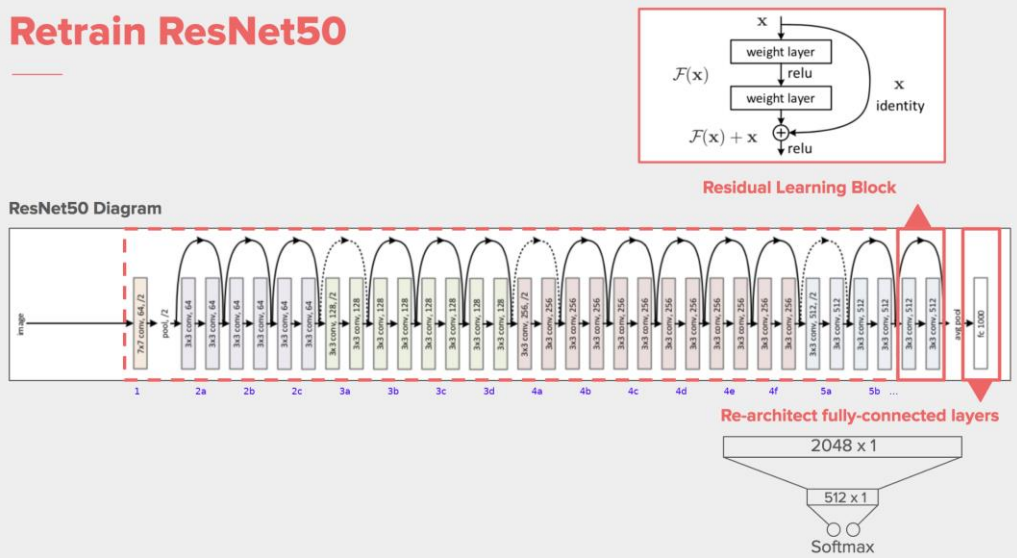
# 1) 모델 간 비교



## 3. 모델

03

①ResNet50 fine-tune  
(R/2+Y/2, G/2+Y/2, B)



Model: "model"

| Layer (type)                              | Output Shape          | Param #  |
|---|-----------------------|----------|
| input_2 (InputLayer)                      | [(None, 299, 299, 3)] | 0        |
| batch_normalization (Batch Normalization) | (None, 299, 299, 3)   | 12       |
| resnet50 (Functional)                     | (None, 10, 10, 2048)  | 23587712 |
| conv2d (Conv2D)                           | (None, 10, 10, 32)    | 65568    |
| flatten (Flatten)                         | (None, 3200)          | 0        |
| dropout (Dropout)                         | (None, 3200)          | 0        |
| dense (Dense)                             | (None, 1024)          | 3277824  |
| dropout_1 (Dropout)                       | (None, 1024)          | 0        |
| dense_1 (Dense)                           | (None, 28)            | 28700    |

Total params: 26,959,816  
Trainable params: 26,906,690  
Non-trainable params: 53,126

Epoch 7/10  
4528/4528 [=====] - 1419s 313ms/step - loss: 0.1729 - binary\_accuracy: 0.9393 - f1: 0.1290 - val\_loss: 0.1660 - val\_binary\_accuracy: 0.9416 - val\_f1: 0.1753  
Epoch 00007: val\_loss improved from 0.16707 to 0.16598, saving model to best\_model3.h5

Epoch 8/10  
4528/4528 [=====] - 1403s 310ms/step - loss: 0.1721 - binary\_accuracy: 0.9394 - f1: 0.1298 - val\_loss: 0.1721 - val\_binary\_accuracy: 0.9393 - val\_f1: 0.1687

Epoch 00008: val\_loss did not improve from 0.16598

Epoch 9/10  
4528/4528 [=====] - 1387s 306ms/step - loss: 0.1728 - binary\_accuracy: 0.9392 - f1: 0.1290 - val\_loss: 0.1685 - val\_binary\_accuracy: 0.9410 - val\_f1: 0.1742

Epoch 00009: val\_loss did not improve from 0.16598

Epoch 10/10  
4528/4528 [=====] - 1397s 309ms/step - loss: 0.1754 - binary\_accuracy: 0.9383 - f1: 0.1248 - val\_loss: 0.1700 - val\_binary\_accuracy: 0.9408 - val\_f1: 0.1725

Epoch 00010: val\_loss did not improve from 0.16598

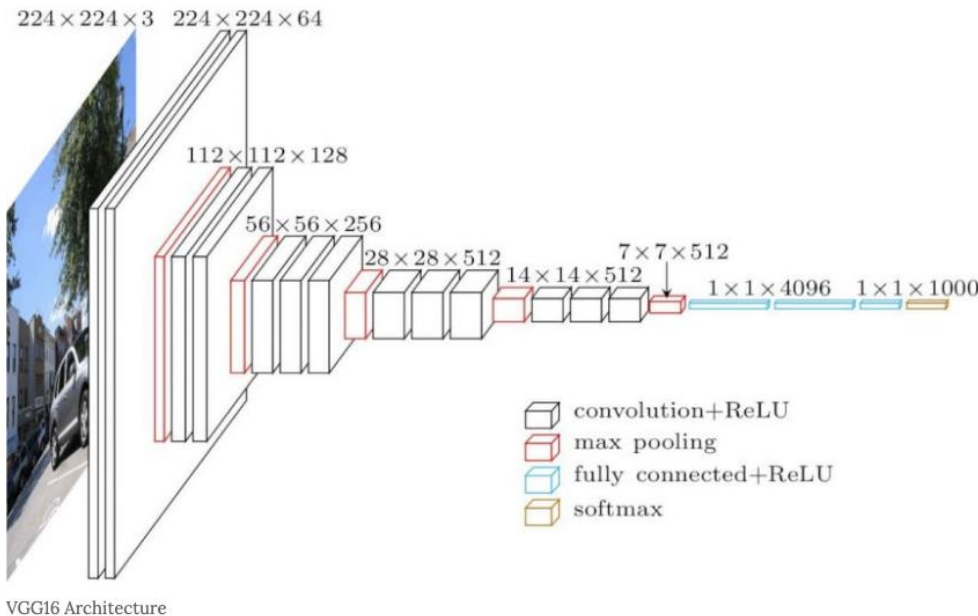
# 1) 모델 간 비교



## 3. 모델

03

②VGG16 fine-tune  
(R/2+Y/2, G/2+Y/2, B)



Model: "model\_1"

| Layer (type)                                  | Output Shape          | Param #  |
|---|-----------------------|----------|
| input_4 (InputLayer)                          | [(None, 299, 299, 3)] | 0        |
| batch_normalization_204 (Batch Normalization) | (None, 299, 299, 3)   | 12       |
| vgg16 (Functional)                            | (None, 9, 9, 512)     | 14714688 |
| conv2d_204 (Conv2D)                           | (None, 9, 9, 32)      | 16416    |
| flatten_1 (Flatten)                           | (None, 2592)          | 0        |
| dropout_2 (Dropout)                           | (None, 2592)          | 0        |
| dense_2 (Dense)                               | (None, 1024)          | 2655232  |
| dropout_3 (Dropout)                           | (None, 1024)          | 0        |
| dense_3 (Dense)                               | (None, 28)            | 28700    |
| Total params: 17,415,048                      |                       |          |
| Trainable params: 17,415,042                  |                       |          |
| Non-trainable params: 6                       |                       |          |

Epoch 5/10  
4528/4528 [=====] - 1451s 320ms/step - loss: 0.2207 - binary\_accuracy: 0.9155 - f1: 0.0659 - val\_loss: 0.2176 - val\_binary\_accuracy: 0.9165 - val\_f1: 0.0776

Epoch 00005: val\_loss improved from 0.21793 to 0.21763, saving model to best\_model2.h5

Epoch 6/10  
4528/4528 [=====] - 1647s 364ms/step - loss: 0.2192 - binary\_accuracy: 0.9162 - f1: 0.0655 - val\_loss: 0.2176 - val\_binary\_accuracy: 0.9166 - val\_f1: 0.0782

Epoch 00006: val\_loss did not improve from 0.21763

Epoch 7/10  
4528/4528 [=====] - 1769s 391ms/step - loss: 0.2199 - binary\_accuracy: 0.9158 - f1: 0.0660 - val\_loss: 0.2176 - val\_binary\_accuracy: 0.9166 - val\_f1: 0.0783

Epoch 00007: val\_loss did not improve from 0.21763

Epoch 8/10  
4528/4528 [=====] - 2197s 485ms/step - loss: 0.2195 - binary\_accuracy: 0.9162 - f1: 0.0655 - val\_loss: 0.2177 - val\_binary\_accuracy: 0.9166 - val\_f1: 0.0776

Epoch 00008: val\_loss did not improve from 0.21763

Epoch 00008: ReduceLROnPlateau reducing learning rate to 1.0000000474974514e-05.

# 1) 모델 간 비교

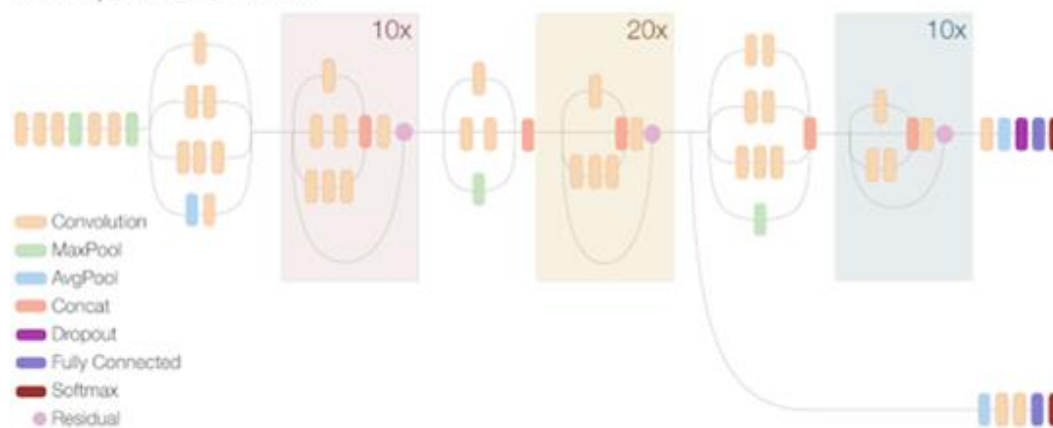


## 3. 모델

03

③ InceptionResNetV2  
fine-tune  
(R/2+Y/2, G/2+Y/2, B)

Compressed View



Schematic diagram of Inception-ResNet-v2

Model: "model"

| Layer (type)                                  | Output Shape          | Param #  |
|---|-----------------------|----------|
| input_2 (InputLayer)                          | [(None, 299, 299, 3)] | 0        |
| batch_normalization_203 (Batch Normalization) | (None, 299, 299, 3)   | 12       |
| inception_resnet_v2 (Functional)              | (None, 8, 8, 1536)    | 54336736 |
| conv2d_203 (Conv2D)                           | (None, 8, 8, 32)      | 49184    |
| flatten (Flatten)                             | (None, 2048)          | 0        |
| dropout (Dropout)                             | (None, 2048)          | 0        |
| dense (Dense)                                 | (None, 1024)          | 2098176  |
| dropout_1 (Dropout)                           | (None, 1024)          | 0        |
| dense_1 (Dense)                               | (None, 28)            | 28700    |
| Total params: 56,512,808                      |                       |          |
| Trainable params: 2,176,066                   |                       |          |
| Non-trainable params: 54,336,742              |                       |          |

```
Epoch 00005: val_loss improved from 0.14463 to 0.14262, saving model to best_model.h5
Epoch 6/10
4528/4528 [=====] - 1425s 315ms/step - loss: 0.1422 - binary_accuracy: 0.9502 - f1: 0.1671 - val_loss: 0.1321 - val_binary_accuracy: 0.9534 - val_f1: 0.2381
Epoch 00006: val_loss improved from 0.14262 to 0.13214, saving model to best_model.h5
Epoch 7/10
4528/4528 [=====] - 1400s 309ms/step - loss: 0.1408 - binary_accuracy: 0.9509 - f1: 0.1678 - val_loss: 0.1452 - val_binary_accuracy: 0.9492 - val_f1: 0.2157
Epoch 00007: val_loss did not improve from 0.13214
Epoch 8/10
4528/4528 [=====] - 1407s 311ms/step - loss: 0.1467 - binary_accuracy: 0.9489 - f1: 0.1621 - val_loss: 0.2111 - val_binary_accuracy: 0.9269 - val_f1: 0.1030
Epoch 00008: val_loss did not improve from 0.13214
Epoch 9/10
4528/4528 [=====] - 1440s 318ms/step - loss: 0.1504 - binary_accuracy: 0.9476 - f1: 0.1549 - val_loss: 0.1417 - val_binary_accuracy: 0.9506 - val_f1: 0.2224
Epoch 00009: val_loss did not improve from 0.13214
Epoch 00009: ReduceLROnPlateau reducing learning rate to 0.000100000000474974513.
```

## 2) 채널 조절(InceptionResNetV2 fine-tune)



### 3. 모델

03

#### ① (R/2+Y/2, G/2+Y/2, B)

```
Epoch 00005: val_loss improved from 0.14463 to 0.14262, saving model to best_model.h5
Epoch 6/10
4528/4528 [=====] - 1425s 315ms/step - loss: 0.1422 - binary_accuracy: 0.9502 - f1: 0.1671 - val_loss: 0.1321 - val_binary_accuracy: 0.9534 - val_f1: 0.2381

Epoch 00006: val_loss improved from 0.14262 to 0.13214, saving model to best_model.h5
Epoch 7/10
4528/4528 [=====] - 1400s 309ms/step - loss: 0.1408 - binary_accuracy: 0.9509 - f1: 0.1678 - val_loss: 0.1452 - val_binary_accuracy: 0.9492 - val_f1: 0.2157

Epoch 00007: val_loss did not improve from 0.13214
Epoch 8/10
4528/4528 [=====] - 1407s 311ms/step - loss: 0.1467 - binary_accuracy: 0.9489 - f1: 0.1621 - val_loss: 0.2111 - val_binary_accuracy: 0.9269 - val_f1: 0.1030

Epoch 00008: val_loss did not improve from 0.13214
Epoch 9/10
4528/4528 [=====] - 1440s 318ms/step - loss: 0.1504 - binary_accuracy: 0.9476 - f1: 0.1549 - val_loss: 0.1417 - val_binary_accuracy: 0.9506 - val_f1: 0.2224

Epoch 00009: val_loss did not improve from 0.13214
Epoch 00009: ReduceLROnPlateau reducing learning rate to 0.000100000000474974513.
```

#### ② (R=0, G, B=0)

```
Epoch 17/20
3019/3019 [=====] - 810s 268ms/step - loss: 0.1095 - binary_accuracy: 0.9606 - f1: 0.2055 - val_loss: 0.1081 - val_binary_accuracy: 0.9617 - val_f1: 0.3130

Epoch 00017: val_loss did not improve from 0.10740
Epoch 18/20
3019/3019 [=====] - 753s 250ms/step - loss: 0.1084 - binary_accuracy: 0.9610 - f1: 0.2098 - val_loss: 0.1079 - val_binary_accuracy: 0.9620 - val_f1: 0.3136

Epoch 00018: val_loss did not improve from 0.10740
Epoch 19/20
3019/3019 [=====] - 745s 247ms/step - loss: 0.1074 - binary_accuracy: 0.9614 - f1: 0.2104 - val_loss: 0.1050 - val_binary_accuracy: 0.9627 - val_f1: 0.3216

Epoch 00019: val_loss improved from 0.10740 to 0.10505, saving model to best_model_sig_g-only.h5
Epoch 20/20
3019/3019 [=====] - 781s 259ms/step - loss: 0.1065 - binary_accuracy: 0.9618 - f1: 0.2113 - val_loss: 0.1051 - val_binary_accuracy: 0.9627 - val_f1: 0.3190

Epoch 00020: val_loss did not improve from 0.10505
```

## 2) 채널 조절(InceptionResNetV2 fine-tune)



### 3. 모델

03

#### ③ (R,G,B)

```
Epoch 1/10
4528/4528 [=====] - 14804s 3s/step - loss: 0.1820 - binary_accuracy: 0.9381 - f1: 0.1267 - val_loss: 0.1557 - val_binary_accuracy: 0.9468 - val_f1: 0.1968

Epoch 00001: val_loss improved from inf to 0.15575, saving model to best_model.h5
Epoch 2/10
4528/4528 [=====] - 14754s 3s/step - loss: 0.1484 - binary_accuracy: 0.9490 - f1: 0.1571 - val_loss: 0.1450 - val_binary_accuracy: 0.9489 - val_f1: 0.2223

Epoch 00002: val_loss improved from 0.15575 to 0.14497, saving model to best_model.h5
Epoch 3/10
4528/4528 [=====] - 22215s 5s/step - loss: 0.1392 - binary_accuracy: 0.9515 - f1: 0.1692 - val_loss: 0.1296 - val_binary_accuracy: 0.9534 - val_f1: 0.2456

Epoch 00003: val_loss improved from 0.14497 to 0.12958, saving model to best_model.h5
Epoch 4/10
4528/4528 [=====] - 14368s 3s/step - loss: 0.1349 - binary_accuracy: 0.9527 - f1: 0.1748 - val_loss: 0.1402 - val_binary_accuracy: 0.9490 - val_f1: 0.2389

Epoch 00004: val_loss did not improve from 0.12958
```

#### ④ (B/2+Y/2, R/2+Y/2, G)

```
Epoch 17/20
3019/3019 [=====] - 969s 321ms/step - loss: 0.1003 - binary_accuracy: 0.9633 - f1: 0.2211 - val_loss: 0.0998 - val_binary_accuracy: 0.9640 - val_f1: 0.3374

Epoch 00017: val_loss did not improve from 0.09979
Epoch 18/20
3019/3019 [=====] - 1070s 354ms/step - loss: 0.1004 - binary_accuracy: 0.9633 - f1: 0.2196 - val_loss: 0.0996 - val_binary_accuracy: 0.9641 - val_f1: 0.3363

Epoch 00018: val_loss improved from 0.09979 to 0.09956, saving model to best_model_sig_rgby.h5
Epoch 19/20
3019/3019 [=====] - 855s 283ms/step - loss: 0.0989 - binary_accuracy: 0.9640 - f1: 0.2223 - val_loss: 0.0979 - val_binary_accuracy: 0.9652 - val_f1: 0.3459

Epoch 00019: val_loss improved from 0.09956 to 0.09793, saving model to best_model_sig_rgby.h5
Epoch 20/20
3019/3019 [=====] - 982s 325ms/step - loss: 0.0981 - binary_accuracy: 0.9640 - f1: 0.2225 - val_loss: 0.0975 - val_binary_accuracy: 0.9648 - val_f1: 0.3409

Epoch 00020: val_loss improved from 0.09793 to 0.09753, saving model to best_model_sig_rgby.h5
```

# 1) 최종 모델 submission



## 4. 결과 및 결론

04

최종모델:

데이터 증강 0, InceptionResNetV2 fine-tune,  
(R,G,B,Y)

### Submission 결과

| Name                       | Submitted | Wait time | Execution time | Score   |
|----------------------------|-----------|-----------|----------------|---------|
| submit_best_model_rgby.csv | just now  | 1 seconds | 0 seconds      | 0.33828 |

Complete

[Jump to your position on the leaderboard](#) ▼

|      |      |                            |  |         |     |    |
|------|------|----------------------------|--|---------|-----|----|
| 1465 | ▲ 50 | C_heng                     |  | 0.33888 | 3   | 3y |
| 1466 | ▼ 34 | LarsHulstaert              |  | 0.33878 | 9   | 3y |
| 1467 | ▼ 20 | masih4                     |  | 0.33829 | 6   | 3y |
| 1468 | ▲ 29 | miyashita                  |  | 0.33807 | 1   | 3y |
| 1469 | ▲ 35 | AntOn                      |  | 0.33797 | 8   | 3y |
| 1470 | ▲ 18 | newber than IG             |  | 0.33765 | 6   | 3y |
| 1471 | ▼ 76 | Krishnadasar Sudheer Kumar |  | 0.33719 | 183 | 3y |

leaderboard:  
**1468/2160**

### 최종모델:

데이터 증강 0, InceptionResNetV2 fine-tune,  
(R,G,B,Y)

비록 리더보드의 등수는 높진 않지만

- ① 1등의 f1 score가 0.59369라는 점에서 많은 차이를 보이지 않음.
- ② 최종모델의 경우 시간관계 상 20epoch에서 멈추었지만, val loss가 계속 줄고 있어서 더 훈련할 경우, 더 좋은 성능을 보였을 것.



## 2) Discussion



## 4. 결과 및 결론

04

### ① Activation function에 sigmoid 사용한 이유

Multi Class: 하나의 이미지에 하나의 객체만 존재, 각 객체는 2개 이상의 카테고리로 분류



Softmax 사용

Mutli Label: 하나의 이미지에 여러 객체가 존재. 각 객체는 2개 이상의 카테고리로 분류



Sigmoid 사용

*Q&A*