

Find Wally with CNN

```
√ import numpy as np

 import matplotlib pyplot as plt 1. matplotlib : 그래프 패키지
 import matplotlib.patches as patches 2.patches: 사각형 그려줌
 import keras.layers as layers
 import keras.optimizers as optimizers
 from keras.models import Model, load model
 from keras.callbacks import LambdaCallback, ModelCheckpoint, ReduceLROnPlateau
 import seaborn as sns 3. seaborn : 그래프 패키지
 from PIL import Image 4. PIL(Pillow): 이미지 처리 패키지
 from skimage transform import resize 5.skimage:이미지 처리 패키지
 import threading, random, os
 from keras.utils.np_utils import to_categorical
 from tensorflow.keras import optimizers
```

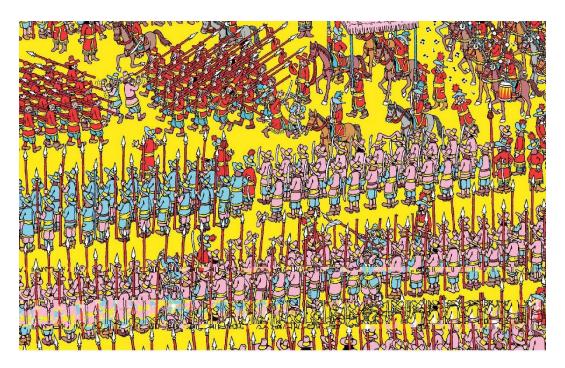
### Load DataSet 이미지 data를 가져옴

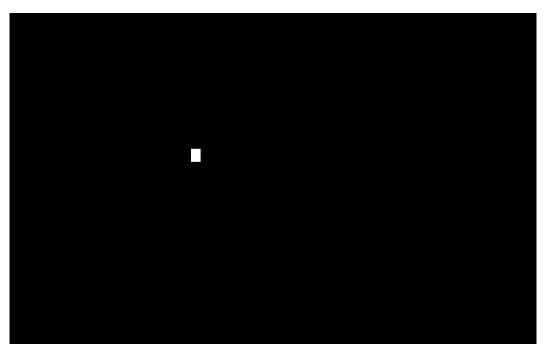
Imgs\_unit8.npy: 전체 원본 이미지

labels\_unit8.npy: 전체 원본 이미지 마스크(월리가 있는 곳만 255, 나머지 0)

waldo\_sub\_imgs\_unit8.npy : 월리 이미지

Waldo\_sub\_labels\_unit8.npy : 월리 이미지 마스크





전체 원본 이미지

전체 원본 이미지 마스크

#### Load DataSet 이미지 data를 가져옴

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월리 이미지



월리 이미지 마스크

#### Load DataSet 이미지 data를 가져옴

```
imgs = np.load('dataset/imgs_uint8.npy', allow_pickle=True).astype(np.float32) / 255.
labels = np.load('dataset/labels_uint8.npy', allow_pickle=True).astype(np.float32) / 255.
waldo_sub_imgs = np.load('dataset/waldo_sub_imgs_uint8.npy', allow_pickle=True) / 255.
waldo_sub_labels = np.load('dataset/waldo_sub_labels_uint8.npy', allow_pickle=True) / 255.

print(imgs.shape, labels.shape)
print(waldo_sub_imgs.shape, waldo_sub_labels.shape)

18개이미지, 가로 2800, 세로 1760의 컬러이미지
```

(18, 1760, 2800, 3) (18, 1760, 2800) 👉 라벨 : 원본 이미지와 같은 사이즈 이미지로

월리가 있는 곳은 1, 없는 곳은 0으로 표시

#### Data Generator 이미지를 무작위 자르고, 뒤집는다.

```
def get_slice(self, i,o): ← 랜덤으로 이미지를 자른다.
   start = random.randint(0, i-o) if self.train else (i-o)
   return slice(start, start+o)
slice_r = self.get_slice(self.ri[idx], self.ro)
   slice c = self.get slice(self.ci[idx], self.co)
   x = self.x[idx][slice r, slice c]
   y = self.y[idx][slice_r, slice_c]
   if self.train and (random.random()>0.5):
      y = y[:,::-1]
                        ◆ 0.5의 확률로 좌우를 반전
   if not self, waldo and np, sum(y) ≠0:
      return None
   return x, to_categorical(y, num_classes=2).reshape((y.shape[0] * y.shape[1], 2))
```

## Data Generator 이미지를 무작위 자르고, 뒤집는다.



## Preview Sample Pannel Images ান্য ধান ক্র

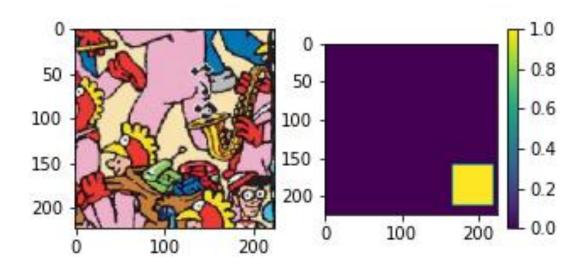
#### 1/3은 월리가 있는 이미지, 나머지는 월리가 없는 이미지

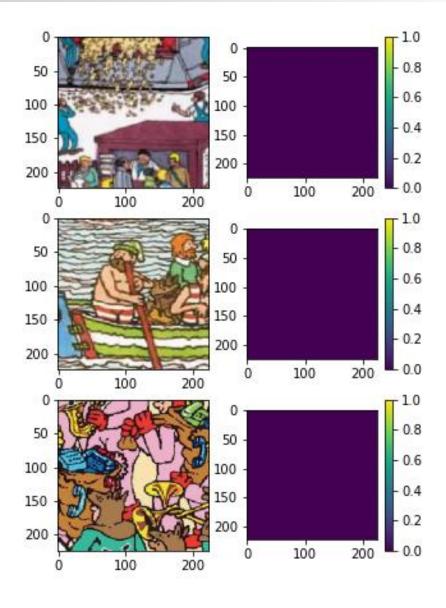
```
# waldo : not_waldo = 1 : 2 (0.34)
gen_mix = seg_gen_mix(waldo_sub_imgs, waldo_sub_labels, imgs, labels, tot_bs=4, prop=0.34)

X, y = next(gen_mix)

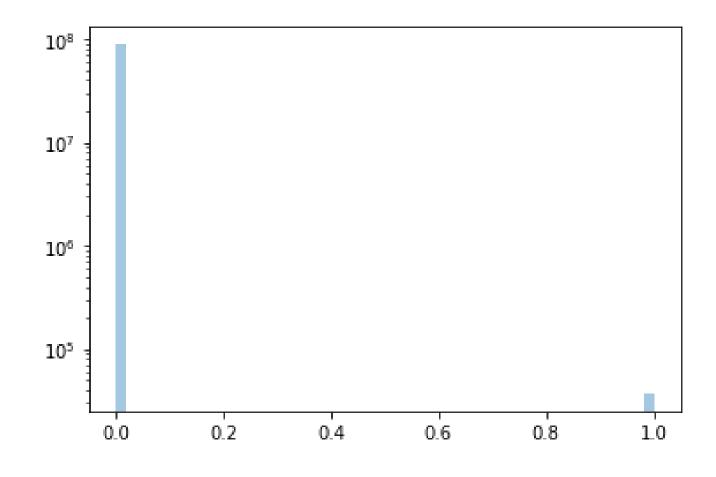
plt.figure(figsize=(5, 10))
for i, img in enumerate(X):
    plt.subplot(X.shape[0], 2, 2*i+1)
    plt.imshow(X[i])
    plt.subplot(X.shape[0], 2, 2*i+2)
    plt.colorbar()
    plt.imshow(y[i][:,1].reshape((PANNEL_SIZE, PANNEL_SIZE)))
```

# Preview Sample Pannel Images ানা ধান্ত কা





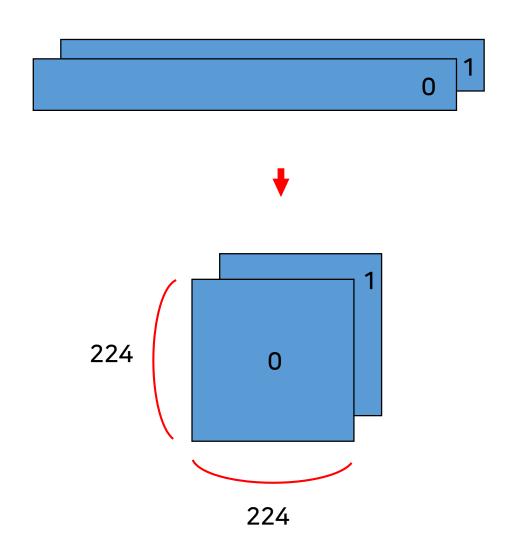
#### Plot Y-Data Distribution ਯੂਗਰੀ ਦੇਸ਼ ਬਾਹ



각 라벨에 대한 데이터셋의 수가 비슷해야 잘 학습된다.

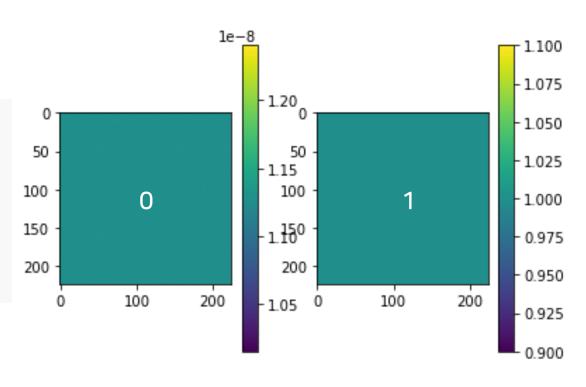
# Make Class Weights আল্লাল্ড ইম শ্বতা.

```
sample_weights = np.zeros((6, PANNEL_SIZE * PANNEL_SIZE, 2))
sample_weights[:,:,0] = 1. / freq0
sample_weights[:,:,1] = 1.
```



## Make Class Weights আল্লাল্ড ইম ক্বা

```
plt.subplot(1,2,1)
plt.imshow(sample_weights[0,:,0].reshape((224, 224)))
plt.colorbar()
plt.subplot(1,2,2)
plt.imshow(sample_weights[0,:,1].reshape((224, 224)))
plt.colorbar()
```



## Create Model

Layer (type)	Output Shape	Param #	Connected to
input_1 (InputLayer)	[(None, 224, 224, 3)	0	
conv2d (Conv2D)	(None, 224, 224, 64)	1792	input_1[0][0]
leaky_re_lu (LeakyReLU)	(None, 224, 224, 64)	0	conv2d[0][0]
max_pooling2d (MaxPooling2D)	(None, 112, 112, 64)	0	leaky_re_lu[0][0]
conv2d_1 (Conv2D)	(None, 112, 112, 128	73856	max_pooling2d[0][0]
leaky_re_lu_1 (LeakyReLU)	(None, 112, 112, 128	0	conv2d_1[0][0]
max_pooling2d_1 (MaxPooling2D)	(None, <mark>56, 56,</mark> 128)	0	leaky_re_lu_1[0][0]
conv2d_2 (Conv2D)	(None, 56, 56, 256)	295168	max_pooling2d_1[0][0]
leaky_re_lu_2 (LeakyReLU)	(None, 56, 56, 256)	0	conv2d_2[0][0]
max_pooling2d_2 (MaxPooling2D)	(None, 28, 28, 256)	0	leaky_re_lu_2[0][0]
conv2d_3 (Conv2D)	(None, 28, 28, 256)	65792	max_pooling2d_2[0][0]
leaky_re_lu_3 (LeakyReLU)	(None, 28, 28, 256)	0	conv2d_3[0][0]
max_pooling2d_3 (MaxPooling2D)	(None, 14, 14, 256)	0	leaky_re_lu_3[0][0]

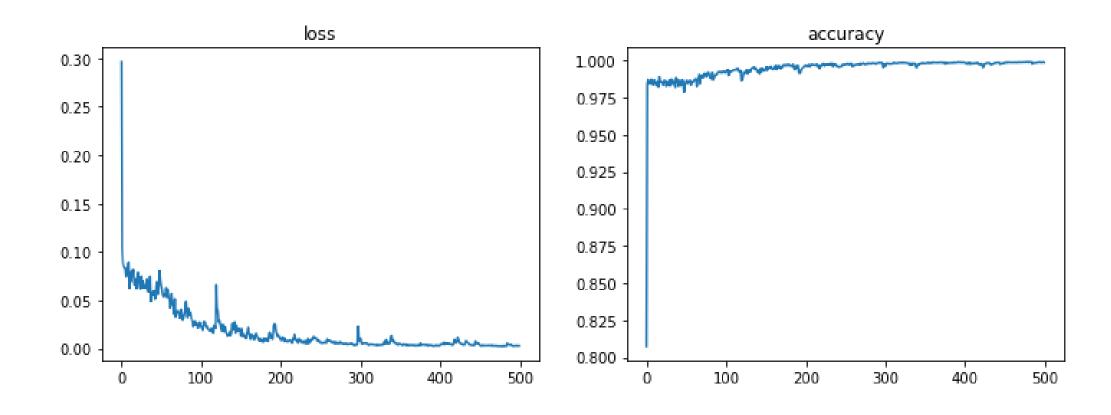
다운 샘플링 224 ▷ 112 ▷ 56 ▷ 28 ▷ 14

## Create Model

max_pooling2d_3 (MaxPooling2D)	(None, 14, 14, 256) 0	leaky_re_lu_3[0][0]
up_sampling2d (UpSampling2D)	(None, 28, 28, 256) 0	max_pooling2d_3[0][0]
conv2d_4 (Conv2D)	(None, <mark>28, 28</mark> , 256) 590080	up_sampling2d[0][0]
activation (Activation)	(None, 28, 28, 256) 0	conv2d_4[0][0]
add (Add)	(None, 28, 28, 256) 0	activation[0][0] max_pooling2d_2[0][0]
up_sampling2d_1 (UpSampling2D)	(None, <mark>56, 56</mark> , 256) 0	add[0][0]
conv2d_5 (Conv2D)	(None, 56, 56, 128) 295040	up_sampling2d_1[0][0]
activation_1 (Activation)	(None, 56, 56, 128) 0	conv2d_5[0][0]
add_1 (Add)	(None, 56, 56, 128) 0	activation_1[0][0] max_pooling2d_1[0][0]
up_sampling2d_2 (UpSampling2D)	(None, <mark>112, 112</mark> , 128 0	add_1[0][0]
conv2d_6 (Conv2D)	(None, 112, 112, 64) 73792	up_sampling2d_2[0][0]
activation_2 (Activation)	(None, 112, 112, 64) 0	conv2d_6[0][0]
add_2 (Add)	(None, 112, 112, 64) 0	activation_2[0][0] max_pooling2d[0][0]
up_sampling2d_3 (UpSampling2D)	(None, <mark>224, 224</mark> 64) 0	add_2[0][0]
conv2d_7 (Conv2D)	(None, 224, 224, 2) 130	up_sampling2d_3[0][0]
reshape (Reshape)	(None, 50176, 2) 0	conv2d_7[0][0]
activation_3 (Activation)	(None, 50176, 2) 0	reshape[0][0]

업 샘플링 14 ▷ 28 ▷ 56 ▷ 112 ▷ 224

# Train



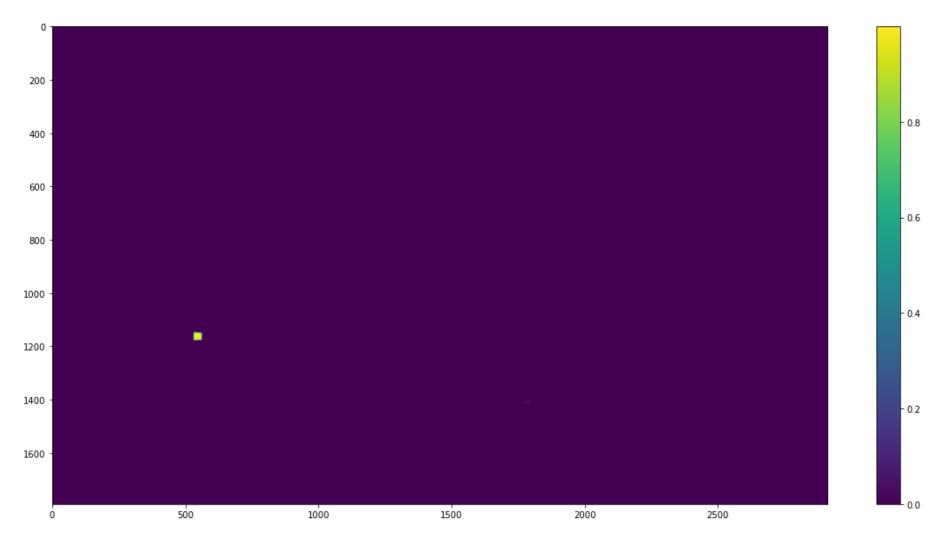
#### **Evaluation**

```
이미지 변경 가능 (ex. O1.jpg)

img_filename = '02.jpg'
test_img = np.array(Image.open(os.path.join('test_imgs', img_filename)).resize((2800, 1760), Image.NEAREST)).astype(np.float32) / 255.

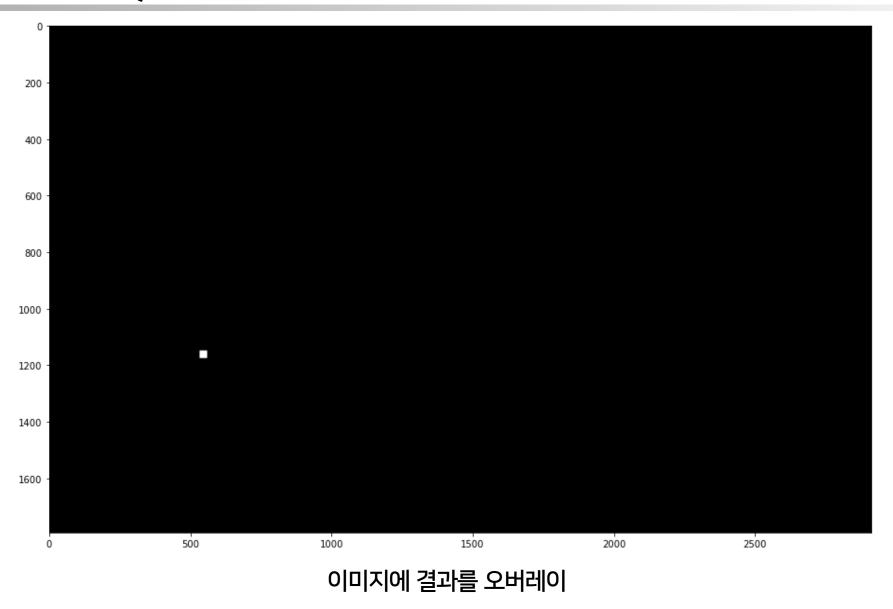
plt.figure(figsize=(20, 10))
plt.imshow(test_img)
```

# **Predict**

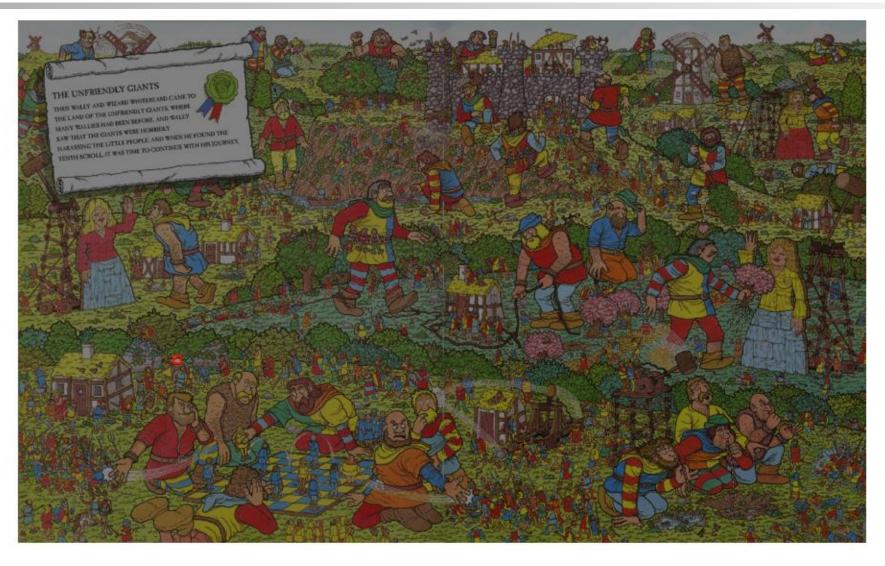


월리가 있는 곳은 1 나머지는 0으로 표시

# Make Overlay for Result



# Final Result



월리가 있는 곳에 빨간 사각형으로 표시