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## 4. Appendix. EfficientNetB6, 1000 epochs

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### 참고자료

- <https://blog.naver.com/beyondlegend/222644092397>
- <https://blog.naver.com/charzim0611/222948860899>
- <https://content.iospress.com/articles/journal-of-intelligent-and-fuzzy-systems/ifs210925>
- [https://www.tensorflow.org/tutorials/images/transfer\\_learning?hl=ko](https://www.tensorflow.org/tutorials/images/transfer_learning?hl=ko)

```
In [ ]: from google.colab import drive
drive.mount("/content/gdrive/")
```

Mounted at /content/gdrive/

```
In [ ]: # 기본
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import math
import os
import shutil
import glob
import cv2
import random

# 경고
import warnings
warnings.filterwarnings('ignore')

# 그래프 설정
plt.rcParams['figure.figsize'] = 20, 10
plt.rcParams['axes.unicode_minus'] = False

# tensor
import tensorflow as tf

# Model, Layers
from tensorflow.keras.models import Sequential
from tensorflow.keras.applications import EfficientNetB6
from tensorflow.keras.applications import InceptionResNetV2
from tensorflow.keras.applications import InceptionV3
from tensorflow.keras.applications import Xception
from tensorflow.keras.utils import Sequence
from tensorflow.keras.layers import Flatten, Dropout, LeakyReLU, Activation, Dense, GlobalAveragePooling2D, BatchNormalization

# optimizer
from keras.optimizers import Adagrad, Adadelta, Adam

# 모델 시각화
from tensorflow.keras.utils import plot_model

# Call backs
from tensorflow.keras.callbacks import ModelCheckpoint, EarlyStopping, LearningRateScheduler, ReduceLROnPlateau

# GPU check
from tensorflow.python.client import device_lib

# torch
import torch
import torch.optim as optim
import torchvision.datasets as datasets
from torch.utils.data import DataLoader, Dataset
import torchvision.transforms as transforms

# 컴퓨터에 있는 GPU 정보들을 가져온다.
# gpu가 있다면...

os.environ["CUDA_VISIBLE_DEVICES"]="0"
gpus = tf.config.experimental.list_physical_devices('GPU')
if gpus:
    try:
        tf.config.experimental.set_memory_growth(gpus[0], True)
    except RuntimeError as e:
        print(e)
```

```
In [ ]: !nvidia-smi
```

|   |          |               |               |                      |              |          |  |                      |            |
|---|----------|---------------|---------------|----------------------|--------------|----------|--|----------------------|------------|
| NVIDIA-SMI 525.85.12      Driver Version: 525.85.12      CUDA Version: 12.0 |          |               |               |                      |              |          |  |                      |            |
| GPU Name  |          | Persistence-M |               | Bus-Id               |              | Disp.A   |  | Volatile Uncorr. ECC |            |
| Fan   | Temp     | Perf          | Pwr:Usage/Cap | Memory-Usage         |              | GPU-Util |  | Compute M. MIG M.    |            |
| 0   | Tesla T4 |               | Off           | 00000000:00:04.0 Off |              |          |  | 0                    |            |
| N/A   | 40C      | P8            | 11W / 70W     | 3MiB / 15360MiB      |              | 0%       |  | Default N/A          |            |
| Processes:  |          |               |               |                      |              |          |  |                      |            |
| GPU   | GI       | CI            | PID           | Type                 | Process name |          |  |                      | GPU Memory |
|   | ID       | ID            |               |                      |              |          |  |                      | Usage      |
| No running processes found  |          |               |               |                      |              |          |  |                      |            |

```
In [ ]: # image size & batch
image_size = 256          # 리사이징할 이미지 사이즈
batch = image_size // 16  # 이미지사이즈 기준 배치사이즈 ex. 32 // 16 == 2, 512 // 16 == 32
channel = 3               # 인풋 채널 수
```

```
Out[ ]: ('/content/gdrive/MyDrive/dataset/train',
         '/content/gdrive/MyDrive/dataset/test')
```

```
Found 777 files belonging to 10 classes.  
Using 700 files for training.
```

```
Found 777 files belonging to 10 classes.
Using 77 files for validation.
```

Found 330 files belonging to 10 classes.

## 라벨 클래스 10개 (10-1 , 10-2 통합)

## # 라벨

```
class_names = train_ds.class_names
print(class_names)
```

```
['0', '1', '2', '3', '4', '5', '6', '7', '8', '9']
```

```
In [ ]: # 라벨 수
num_classes = len(class_names)
num_classes
```

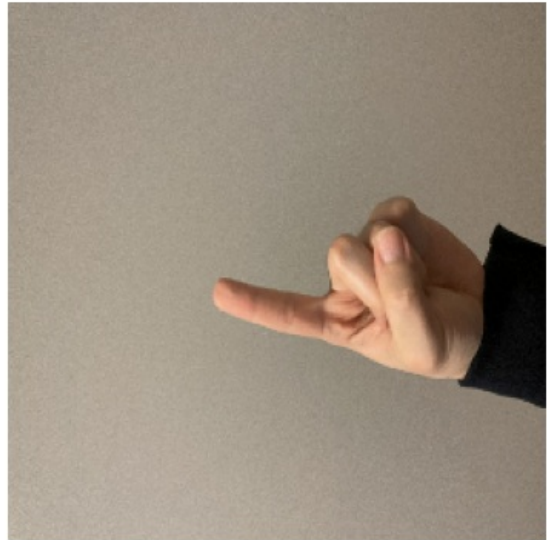
```
Out[ ]: 10
```

```
In [ ]: # train 데이터 확인
plt.figure(figsize=(10, 10))
for images, labels in train_ds.take(1):
    for i in range(4):
        ax = plt.subplot(2, 2, i + 1)
        plt.imshow(images[i].numpy().astype("uint8"))
        plt.title(class_names[labels[i]])
        plt.axis("off")
```

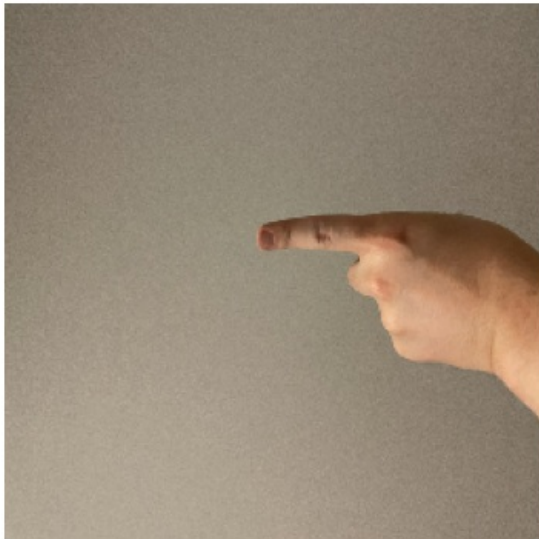
2



1



1



4



```
In [ ]: # 프리페치/
AUTOTUNE = tf.data.AUTOTUNE

train_ds = train_ds.cache().shuffle(1000).prefetch(buffer_size=AUTOTUNE)
val_ds = val_ds.cache().prefetch(buffer_size=AUTOTUNE)
```

```
In [ ]: # 전이학습
# base_model = Xception(weights = 'imagenet', include_top = False, input_shape = (image_size, image_size, channels))
# base_model = InceptionResNetV2(weights = 'imagenet', include_top = False, input_shape = (image_size, image_size, channels))
base_model = EfficientNetB6(weights = 'imagenet', include_top = False, input_shape = (image_size, image_size, channels))
# base_model = tf.keras.applications.MobileNetV2(weights='imagenet', include_top=False, input_shape = (image_size, image_size, channels))
```

```
Downloading data from https://storage.googleapis.com/keras-applications/efficientnetb6_notop.h5
165234480/165234480 [=====] - 1s 0us/step
```

```
In [ ]: # Preprocessing
```

```
preprocess_input = tf.keras.applications.xception.preprocess_input
```

```
In [ ]: # 모델 학습 여부
base_model.trainable = False
print(len(base_model.layers)) # 베이스모델의 층 갯수 확인

667
```

```
In [ ]: # 모델에서 n층까진 w값을 조정하지 않고, 그 다음부터 w값을 조정
for layer in base_model.layers[:100]:
    layer.trainable = True

# 학습가능한 변수층 수
len(base_model.trainable_variables)
```

```
Out[ ]: 0
```

```
In [ ]: # 특징 추출
image_batch, label_batch = next(iter(train_ds))
feature_batch = base_model(image_batch)
print(feature_batch.shape)

global_average_layer = tf.keras.layers.GlobalAveragePooling2D()
feature_batch_average = global_average_layer(feature_batch)
print(feature_batch_average.shape)

prediction_layer = tf.keras.layers.Dense(num_classes, activation='softmax')
prediction_batch = prediction_layer(feature_batch_average)
print(prediction_batch.shape)

(16, 8, 8, 2304)
(16, 2304)
(16, 10)
```

```
In [ ]: # Pretrained Model
inp = tf.keras.Input(shape=(image_size, image_size, channel))

data_aug = Sequential([tf.keras.layers.RandomFlip("horizontal"),
                        tf.keras.layers.RandomRotation(0.2),
                        tf.keras.layers.RandomZoom(0.2)
                      ])(inp)

x = preprocess_input(data_aug)

x = base_model(x)
x = global_average_layer(x)

x = Dense(image_size * 4, activation = LeakyReLU(alpha=0.2))(x)
x = Dropout(0.5)(x)
x = Dense(image_size * 2, activation = LeakyReLU(alpha=0.2))(x)
x = Dropout(0.5)(x)
x = Dense(image_size, activation = LeakyReLU(alpha=0.2))(x)
x = Dropout(0.5)(x)

outp = Dense(num_classes, activation='softmax')(x)
model = tf.keras.Model(inp, outp)
```

```
In [ ]: # optimizer = tf.keras.optimizers.SGD(momentum = True, nesterov = True)
# optimizer = tf.keras.optimizers.RMSprop()
optimizer = tf.keras.optimizers.Adam(beta_1 = 0.9, beta_2 = 0.999)

model.compile(loss = "sparse_categorical_crossentropy",
              optimizer = optimizer,
              metrics = ['accuracy'])
```

```
In [ ]: # 모델 플롯
plot_model(model, show_layer_names=True, show_trainable = True, show_layer_activations=True, show_shapes=True)
```

```
Out[ ]:
```

|   |            |         |                       |
|---|------------|---------|-----------------------|
| T | input_2    | input:  | [(None, 256, 256, 3)] |
|   | InputLayer | output: | [(None, 256, 256, 3)] |



|   |            |         |                     |
|---|------------|---------|---------------------|
| T | sequential | input:  | (None, 256, 256, 3) |
|   | Sequential | output: | (None, 256, 256, 3) |



|   |                 |         |                     |
|---|-----------------|---------|---------------------|
| T | tf.math.truediv | input:  | (None, 256, 256, 3) |
|   | TFOpLambda      | output: | (None, 256, 256, 3) |



|   |                  |         |                     |
|---|------------------|---------|---------------------|
| T | tf.math.subtract | input:  | (None, 256, 256, 3) |
|   | TFOpLambda       | output: | (None, 256, 256, 3) |



|    |                |         |                     |
|----|----------------|---------|---------------------|
| NT | efficientnetb6 | input:  | (None, 256, 256, 3) |
|    | Functional     | output: | (None, 8, 8, 2304)  |



|   |                          |         |                    |
|---|--------------------------|---------|--------------------|
| T | global_average_pooling2d | input:  | (None, 8, 8, 2304) |
|   | GlobalAveragePooling2D   | output: | (None, 2304)       |



|   |         |             |         |              |
|---|---------|-------------|---------|--------------|
| T | dense_1 |             | input:  | (None, 2304) |
|   | Dense   | leaky_re_lu | output: | (None, 1024) |



|   |         |         |              |
|---|---------|---------|--------------|
| T | dropout | input:  | (None, 1024) |
|   | Dropout | output: | (None, 1024) |



|   |         |               |         |              |
|---|---------|---------------|---------|--------------|
| T | dense_2 |               | input:  | (None, 1024) |
|   | Dense   | leaky_re_lu_1 | output: | (None, 512)  |



|   |           |         |             |
|---|-----------|---------|-------------|
| T | dropout_1 | input:  | (None, 512) |
|   | Dropout   | output: | (None, 512) |



|   |         |               |         |             |
|---|---------|---------------|---------|-------------|
| T | dense_3 |               | input:  | (None, 512) |
|   | Dense   | leaky_re_lu_2 | output: | (None, 256) |



|   |           |         |             |
|---|-----------|---------|-------------|
| T | dropout_2 | input:  | (None, 256) |
|   | Dropout   | output: | (None, 256) |



|   |         |         |         |             |
|---|---------|---------|---------|-------------|
| T | dense_4 |         | input:  | (None, 256) |
|   | Dense   | softmax | output: | (None, 10)  |

```
In [ ]: # 모델 구조 확인
model.summary()

Model: "model"

Layer (type)                Output Shape                Param #
=====
input_2 (InputLayer)        [(None, 256, 256, 3)]      0

sequential (Sequential)      (None, 256, 256, 3)         0

tf.math.truediv (TFOpLambda) (None, 256, 256, 3)         0

tf.math.subtract (TFOpLambda) (None, 256, 256, 3)         0

efficientnetb6 (Functional)  (None, 8, 8, 2304)          40960143

global_average_pooling2d (GlobalAveragePooling2D) (None, 2304)                0

dense_1 (Dense)              (None, 1024)                2360320

dropout (Dropout)            (None, 1024)                0

dense_2 (Dense)              (None, 512)                 524800

dropout_1 (Dropout)          (None, 512)                 0

dense_3 (Dense)              (None, 256)                 131328

dropout_2 (Dropout)          (None, 256)                 0

dense_4 (Dense)              (None, 10)                  2570

=====
Total params: 43,979,161
Trainable params: 3,019,018
Non-trainable params: 40,960,143

In [ ]: # 학습모델을 저장할 경로
path = PATH + '/models/pretrained/'

In [ ]: # 폴더 생성 및 체크
if os.path.isdir(path) :
    pass
else :
    os.makedirs(os.path.join(path))

In [ ]: # parameters 세팅
epoch_n = 1000

In [ ]: # 스케줄러
def lr_step_decay(epoch, lr):
    drop_rate = 0.899 # 학습률 드랍 비율
    epochs_drop = round(math.sqrt(epoch_n))
    initial_lr = 0.001 # 사전학습시 학습률 낮게
    lr = initial_lr * math.pow(drop_rate, math.floor(epoch / epochs_drop))
    return lr

scheduler = LearningRateScheduler(lr_step_decay, verbose=1)

In [ ]: # 조기 종료
early_stop = tf.keras.callbacks.EarlyStopping(
    monitor='val_loss', # 모니터링 지표표
    min_delta = 0.001,
    patience = round(math.sqrt(epoch_n)), # 기준횟수
    restore_best_weights = True, # 최상 weights 복원
    verbose=1)

In [ ]: # ReduceLROnPlateau
reduce_lr = ReduceLROnPlateau(monitor = 'val_accuracy', patience = 1, verbose = 1)

In [ ]: # histroy 초기화
model_history = []

In [ ]: # model_fit
history = model.fit(train_ds, validation_data = val_ds,
                    workers = 8,
                    epochs = epoch_n,
```

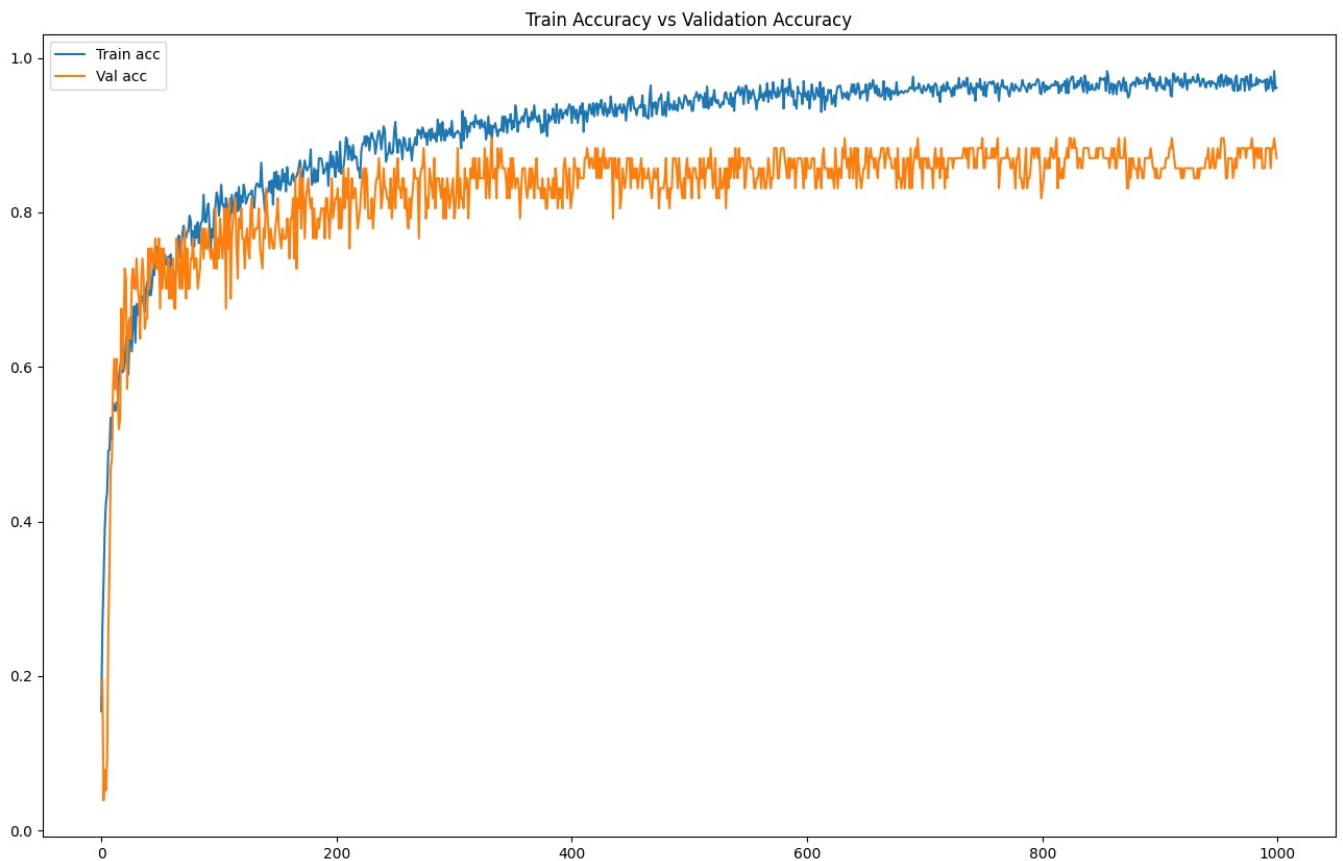
```
callbacks = [scheduler, reduce_lr]
)

model_history.append(history)
```

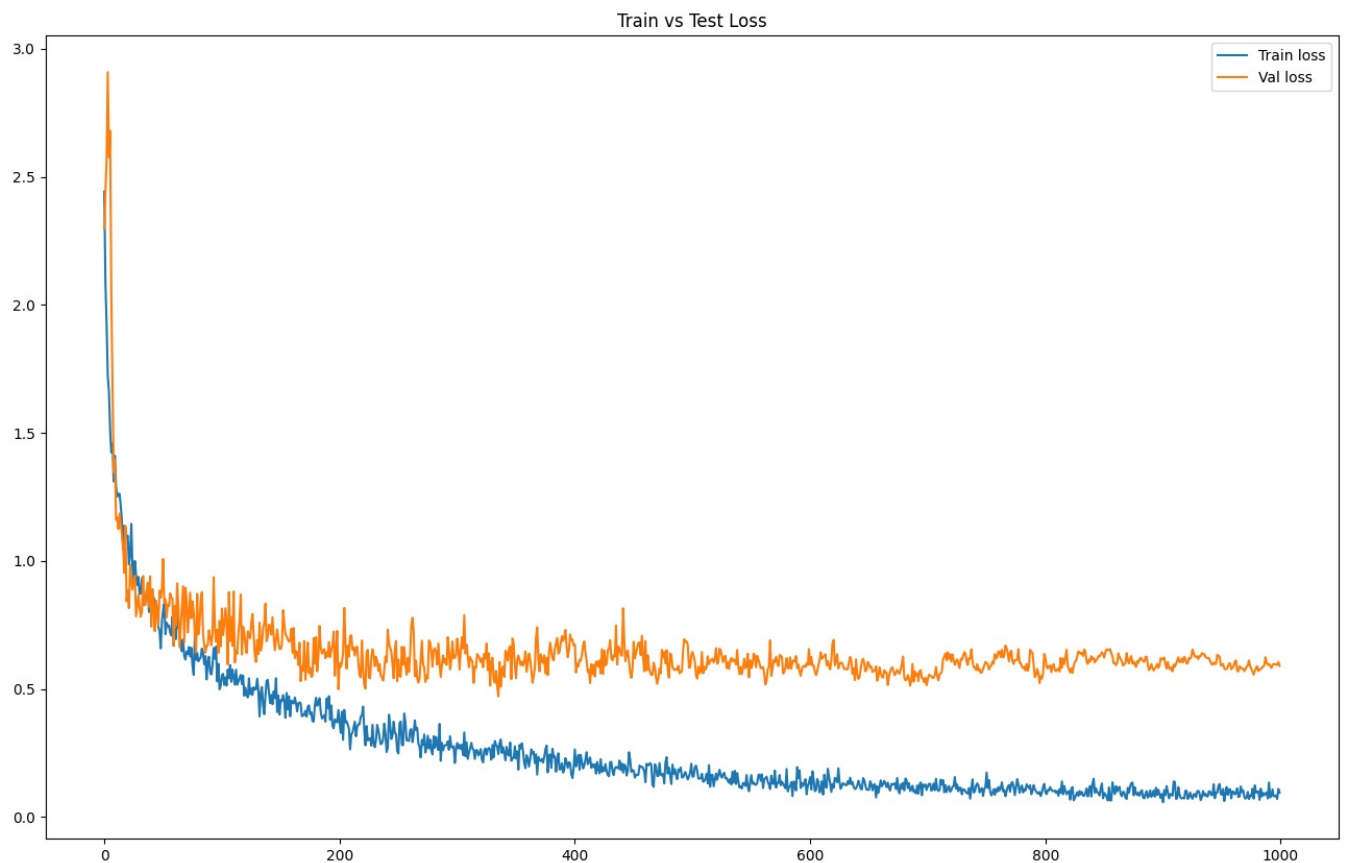
## [학습 로그 부분 삭제]

```
In [ ]: # 리스트를 추출한다.
loss_list = history.history['loss']
accuracy_list = history.history['accuracy']
val_loss_list = history.history['val_loss']
val_accuracy_list = history.history['val_accuracy']
```

```
In [ ]: # 정확도 그래프
plt.figure(figsize=(16, 10))
plt.title('Train Accuracy vs Validation Accuracy')
plt.plot(accuracy_list, label='Train acc')
plt.plot(val_accuracy_list, label='Val acc')
plt.legend()
plt.show()
```



```
In [ ]: # 손실률
plt.figure(figsize=(16, 10))
plt.title('Train vs Test Loss')
plt.plot(loss_list, label='Train loss')
plt.plot(val_loss_list, label='Val loss')
plt.legend()
plt.show()
```



```
In [ ]: # 테스트셋 적용
model.evaluate(test_ds)
```

21/21 [=====] - 37s 1s/step - loss: 1.8002 - accuracy: 0.7667

```
Out[ ]: [1.800195455511475, 0.7666666507720947]
```

```
In [ ]: # 중지
//
```

```
File "<ipython-input-36-9402994718cf>", line 2
    //
    ^
SyntaxError: invalid syntax
```

```
In [ ]: # 모델 훈련 계속하기
fine_tune_epochs = 50
total_epochs = epoch_n + fine_tune_epochs

history_fine1 = model.fit(train_ds, validation_data = val_ds,
                           epochs = total_epochs,
                           initial_epoch = history.epoch[-1],
                           callbacks = [scheduler, early_stop]
                           )
```

```
In [ ]: loss_list += history_fine1.history['loss']
accuracy_list += history_fine1.history['accuracy']
val_loss_list += history_fine1.history['val_loss']
val_accuracy_list += history_fine1.history['val_accuracy']
```

```
In [ ]: # 테스트셋 적용
model.evaluate(test_ds)
```

```
In [ ]: # 모델 훈련 계속하기
fine_tune_epochs = 50
total_epochs = epoch_n + fine_tune_epochs

history_fine2 = model.fit(train_ds, validation_data = val_ds,
                           epochs = total_epochs,
                           initial_epoch = history.epoch[-1],
                           callbacks = [scheduler, early_stop]
                           )
```

```
In [ ]:
```

```
In [ ]:
```

```
In [ ]:
```



## 이하 휴지통 (초기 모델 및 데이터 전처리, not TFDS)

```
In [ ]: ## 이미지 처리
## heif 확장자 처리
# !pip install pillow-heif
# from PIL import Image
# from pillow_heif import register_heif_opener
# register_heif_opener()

In [ ]: ## 하위 폴더 들의 이름을 가져온다(결과 데이터)
# categories = list(os.walk(PATH + 'train'))[0][1]

In [ ]: # categories

In [ ]: # num_classes = len(categories)
# num_classes

In [ ]: ## image size
# image_size = 128
# channel = 3

In [ ]: ## Imgdagen
# idg = ImageDataGenerator(horizontal_flip = True,
#                             vertical_flip = True,
#                             rotation_range = 0.2,
#                             height_shift_range = 0.2,
#                             width_shift_range = 0.2,
#                             )

In [ ]: # 매우 느리고 무거움..
# X = []
# y = []
# averArr = np.zeros(shape=(image_size, image_size, channel))

# for idx, category in enumerate(categories):
#     # one-hot
#     label = [0 for i in range(num_classes)]
#     label[idx] = 1

#     image_dir = PATH + '/train/' + category
#     files = glob.glob(image_dir + '/*')

#     for i, j in enumerate(files):
#         # 원본 저장
#         img_org = Image.open(j)
#         img1 = img_org.resize((image_size, image_size))
#         data = np.asarray(img1) / 255
#         averArr += np.asarray(img_org.resize((image_size, image_size)))

#         X.append(data)
#         y.append(label)

#         # Augmentation
#         for k in range(0,2):
#             img2 = np.expand_dims(img1, axis = 0)
#             datagen = idg.flow(img2)
#             aug_img = next(datagen)
#             img2 = np.squeeze(aug_img)
#             data = np.asarray(img2) / 255

#             X.append(data)
#             y.append(label)

#         if i == 199:
#             averArr = (np.trunc((averArr / 200))) / 255
#             X.append(averArr)
#             averArr = np.zeros(shape=(image_size, image_size, channel))

#     print(category, 'done')

# X = np.array(X)
# y = np.array(y)
```

```
In [ ]: # X.shape, y.shape
```

```
In [ ]: # len(X)
```

```

In [ ]: # len(X)

In [ ]: # # X 샘플확인
# plt.figure(figsize=(16,10))
# for g in range(len(X[:16])):
#     plt.subplot(4,4,g+1)
#     plt.imshow(X[g])
#     plt.axis('off')

In [ ]: # # 전이학습
# base_model = Xception(weights = 'imagenet', include_top = False, input_shape = (image_size, image_size, channels))
# # base_model = InceptionResNetV2(weights = 'imagenet', include_top = False, input_shape = (image_size, image_size, channels))
# # base_model = EfficientNetV2S(weights = 'imagenet', include_top = False, input_shape = (image_size, image_size, channels))

In [ ]: # # 모델 학습 여부
# base_model.trainable = True
# print(len(base_model.layers)) # 베이스모델의 층 갯수 확인

In [ ]: # # 모델에서 n층까진 w값을 조정하지 않고, 그 다음부터 w값을 조정
# for layer in base_model.layers[:100]:
#     layer.trainable = False

# # 학습가능한 변수층 수
# len(base_model.trainable_variables)

In [ ]: # model = Sequential()
# model.add(base_model)

# model.add(Flatten())

# model.add(Dense(image_size * 2, activation = LeakyReLU(alpha=0.2)))
# model.add(Dropout(0.5))

# model.add(Dense(image_size * 4, activation = LeakyReLU(alpha=0.2)))
# model.add(Dropout(0.5))

# model.add(Dense(image_size * 2, activation = LeakyReLU(alpha=0.2)))
# model.add(Dropout(0.5))

# model.add(Dense(num_classes, activation='softmax'))

# # optimizer = tf.keras.optimizers.SGD(lr, momentum = True, nesterov = True)
# # optimizer = tf.keras.optimizers.RMSprop(lr, rho = 0.9)
# optimizer = tf.keras.optimizers.Adam(beta_1 = 0.9, beta_2 = 0.999)

# model.compile(loss = "categorical_crossentropy",
#               optimizer = optimizer,
#               metrics = ['accuracy'])

# model.summary()

In [ ]: # # 모델 플롯
# plot_model(model, show_layer_names= True, show_trainable = True, show_layer_activations= True, show_shapes= True)

In [ ]: # # 학습모델을 저장할 경로
# path = PATH + '/models/pretrained/'

In [ ]: # # 폴더 생성 및 체크
# if os.path.isdir(path) :
#     pass
# else :
#     os.makedirs(os.path.join(path))

In [ ]: # # 데이터로더
# class DataLoader(Sequence):

#     def __init__(self, x_set, y_set, batch_size, shuffle=False):
#         self.x, self.y = x_set, y_set
#         self.batch_size = batch_size
#         self.shuffle = shuffle
#         self.on_epoch_end()

#     def __len__(self):
#         return math.ceil(len(self.x) / self.batch_size)

#     def __getitem__(self, idx):
#         indices = self.indices[idx * self.batch_size : (idx + 1) * self.batch_size]

#         batch_x = [self.x[i] for i in indices]
#         batch_y = [self.y[i] for i in indices]

```

```
#         return np.array(batch_x), np.array(batch_y)

#     def on_epoch_end(self):
#         self.indices = np.arange(len(self.x))
#         if self.shuffle == True:
#             np.random.shuffle(self.indices)
```

```
In [ ]: # 모델 체크포인트
# path1 = path + '{epoch}-{val_loss}.h5'
# path2 = path + '/best_model.h5'

# # 저장 콜백
# call1 = ModelCheckpoint(filepath = path1, monitor = 'val_loss', save_best_only = True) # 용량주의
# call2 = ModelCheckpoint(filepath = path2, monitor = 'val_loss', save_best_only = True) # 용량주의
```

```
In [ ]: # # parameters 세팅
# epoch_n = 1000
# batch = 32
```

```
In [ ]: # # 스케줄러
# def lr_step_decay(epoch, lr):
#     drop_rate = 0.9876 # 학습률 드랍 비율
#     epochs_drop = round(math.sqrt(epoch_n))
#     initial_lr = 0.001
#     lr = initial_lr * math.pow(drop_rate, math.floor(epoch / epochs_drop))
#     return lr

# scheduler = LearningRateScheduler(lr_step_decay, verbose=1)
```

```
In [ ]: # # 조기 종료
# early_stop = tf.keras.callbacks.EarlyStopping(
#     monitor='val_loss', # 모니터링 지표표
#     min_delta = 0.001,
#     patience = round(math.sqrt(epoch_n)), # 기준횟수
#     restore_best_weights = True, # 최상 weights 복원
#     verbose=1)
```

```
In [ ]: # # Train, Val
# X_train, X_test, y_train, y_test = train_test_split(X, y,
#     random_state = 1,
#     shuffle = True,
#     test_size = 0.2)

# X_train.shape, X_test.shape, y_train.shape, y_test.shape
```

```
In [ ]: # model_history = []

# # Data loader
# train_loader = Dataloader(X_train, y_train, batch, shuffle = True)
# test_loader = Dataloader(X_test, y_test, batch)

# history = model.fit(train_loader, validation_data = test_loader,
#     epochs = epoch_n,
#     workers = 8,
#     callbacks= [scheduler, early_stop]
# )

# model_history.append(history)
```

```
In [ ]: # len(model_history)
```

```
In [ ]: # # Val 정확도 확인
# a1 = model.evaluate(X_test, y_test)
# print(f' 테스트 손실률 : {a1[0]}')
# print(f' 테스트 정확도 : {a1[1]}')
```

```
In [ ]: # # 리스트를 추출
# loss_list = history.history['loss']
# accuracy_list = history.history['accuracy']
# val_loss_list = history.history['val_loss']
# val_accuracy_list = history.history['val_accuracy']
```

```
In [ ]: # # Acc
# plt.figure(figsize= (16, 10))
# plt.title('Train Accuracy vs Validation Accuracy')
# plt.plot(accuracy_list, label='Train acc')
# plt.plot(val_accuracy_list, label= 'Val acc')
# plt.legend()
# plt.show()
```

```
In [ ]: # # Loss
# plt.figure(figsize= (16, 10))
# plt.title('Train vs Test Loss')
# plt.plot(loss_list, label='Train loss')
# plt.plot(val_loss_list, label='Val loss')
# plt.legend()
# plt.show()
```

```
In [ ]: # len(y_train), len(y_test)
```

```
In [ ]: # # 모델 세이브
# model.save(PATH + 'models/pretrained/best_model.h5')
```

```
In [ ]:
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