CNNの実習:キャグルの樹木の葉の病気の診断

https://www.kaggle.com/c/plant-pathology-2020-fgvc7

Plant Pathology 2020 - FGVC7

Identify the category of foliar diseases in apple trees



リンゴの木の葉のイメージで病気の種類を 判別するキャグルコンペティション

Plant Pathology 2020 - FGVC7

Identify the category of foliar diseases in apple trees



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Dataset Description

Given a photo of an apple leaf, can you accurately assess its health? This competition will challenge you to distinguish between leaves which are healthy, those which are infected with apple rust, those that have apple scab, and those with more than one disease.

Files

train.csv

- image_id: the foreign key
- · combinations: one of the target labels
- healthy: one of the target labels
- rust: one of the target labels
- scab: one of the target labels

Files

3645 files

Size

823.79 MB

Type

jpg, csv

License

Subject to Competition Rules

リンゴの葉の写真が与えられたとき、あなたはその葉の健康状態を正確に判断できますか? このコンテストでは、健康な葉、さび病に感染している葉、リンゴ<mark>かさぶた</mark>に感染している 葉、複数の病気に感染している葉を見分けることに挑戦します。

データセットの概要



- Rust, scab, combination, healthyなど4つのタイプを判別
- 個々の画像サイズは(1365×2048)
- 学習画像の件数は1821件
- 性能評価指標はROC-AUC

Files

train.csv

- image_id: the foreign key
- combinations: one of the target labels
- healthy: one of the target labels
- rust: one of the target labels
- scab: one of the target labels

images

A folder containing the train and test images, in jpg format.

test.csv

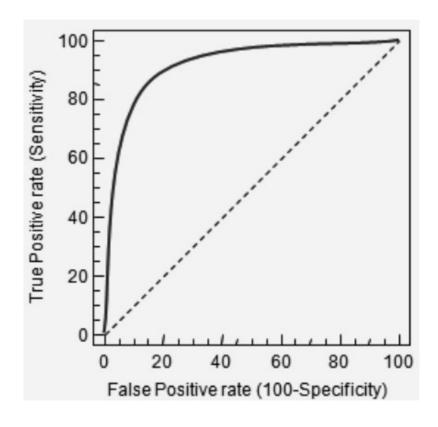
• image_id: the foreign key

sample_submission.csv

- image_id: the foreign key
- combinations: one of the target labels
- healthy: one of the target labels
- rust: one of the target labels
- scab: one of the target labels

ROC-AUCとは

- ROC 曲線(Receiver Operation Characteristic Curve)とそれに基づいたAUC(Area Under Curve)スコアは、バイナリ分類の予測性能測定に使われる重要な指標である。
- 一般的に医学分野で多く使用されますが、機械学習の分類モデルの予測性能を判断する指標として重要な評価指標でもあります。



- ROC曲線は、FPR(False Positive Rate)が変化すると TPR(True Positive Rate)がどのように変化するかを示す曲線です。FPRをX軸に、TPRをY軸にすると、FPRの変化によるTPRの変化が曲線の形で現れます。
- 分類の性能指標として使用されるのは、ROC曲線の**面積**に基づいた**AUC値**で決定します。AUC値はROC曲線の下の面積を求めたもので、一般的に1に近いほど良い数値です。

ROC曲線

FPRの変化に伴うTPRの変化曲線

TPRはTrue Positive Rateの略で、これは再現率(感度)を表します。

$$TPR=TP / (FN + TP)$$

FPRは、実際のNegativeを誤ってPositiveに予測した割合を表します。

$$FPR = FP / (FP + TN)$$

- 閾値を1にするとFPRは0になります。
- ・ 閾値をOにするとFPRは1である。

予測した値 (Predict class) Positive (1) Negative(0) Negative Negative **Negative Positive** Negative(0) (True Negative) (False Positive) FP TN 実際の値 Positive Negative Positive Positive (Ground (False Negative) (True Positive) FN TP Truth)

Evaluation



Submissions are evaluated on mean column-wise ROC AUC. In other words, the score is the average of the individual AUCs of each predicted column.

Submission File

For each image_id in the test set, you must predict a probability for each target variable. The file should contain a header and have the following format:

```
image_id,
test_0,0.25,0.25,0.25
test_1,0.25,0.25,0.25
test_2,0.25,0.25,0.25
etc.
```





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Code

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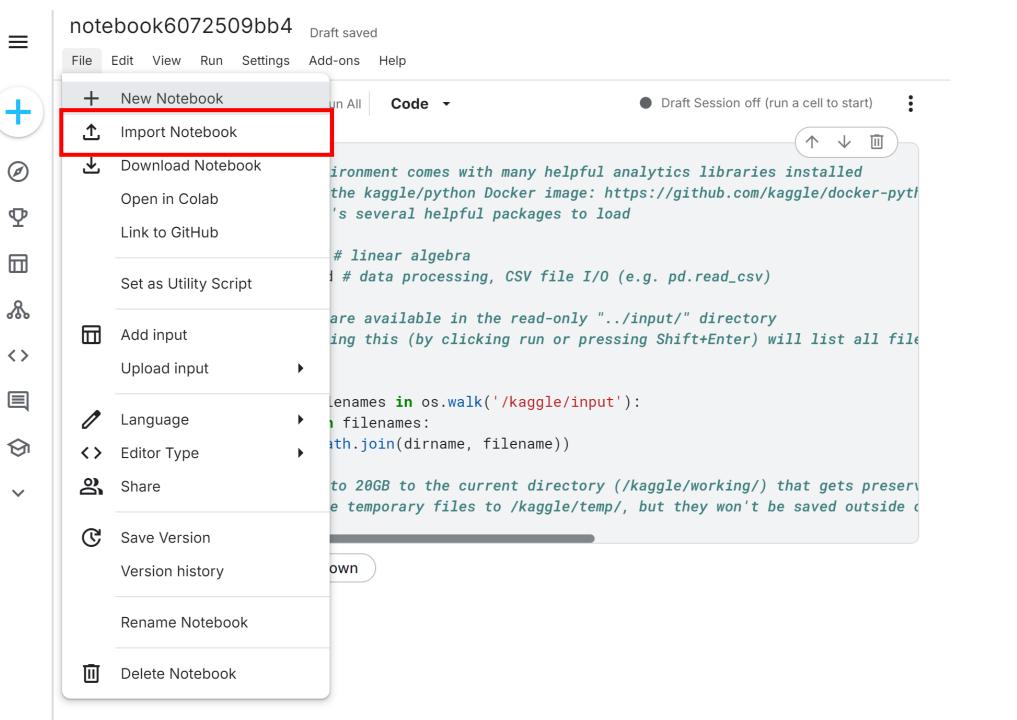
GPU

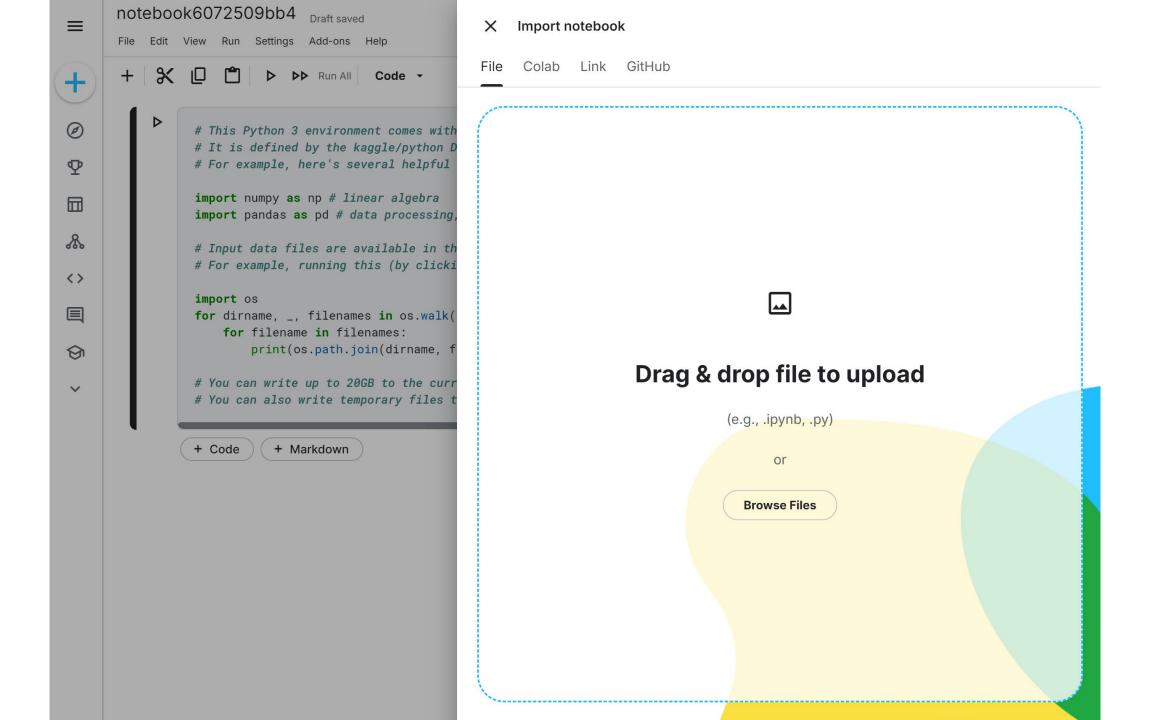
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Plant_Pathology.ipynb

X

```
plant_Pathology_01 Draft save
File Edit View Run Set<mark>i</mark>ngs Add-ons Help
                       ▶▶ Run All Markdown ▼
                                                               Draft Session off (run a cell to start)
         + Markdown
          + Code
          import numpy as np
          import pandas as pd
          import os
          for dirname, _, filenames in os.walk('/kaggle/input'):
              for filename in filenames:
                  print(os.path.join(dirname, filename))
          import numpy as np
          import pandas as pd
          import os
          test_df = pd.read_csv("../input/plant-pathology-2020-fgvc7/test.csv")
          train_df = pd.read_csv("../input/plant-pathology-2020-fgvc7/train.csv")
          train_df.head()
```

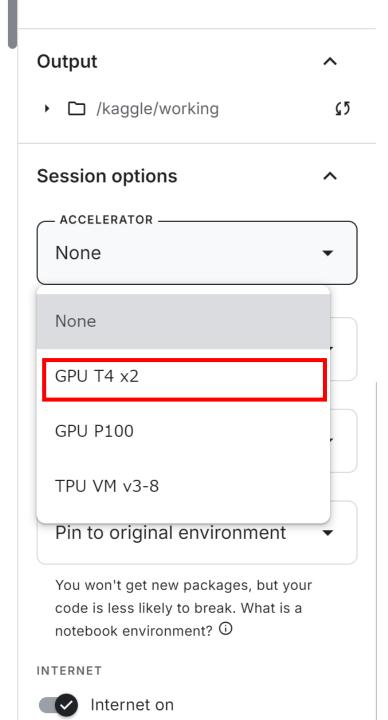
```
import numpy as np
import pandas as pd

import os
for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))
```

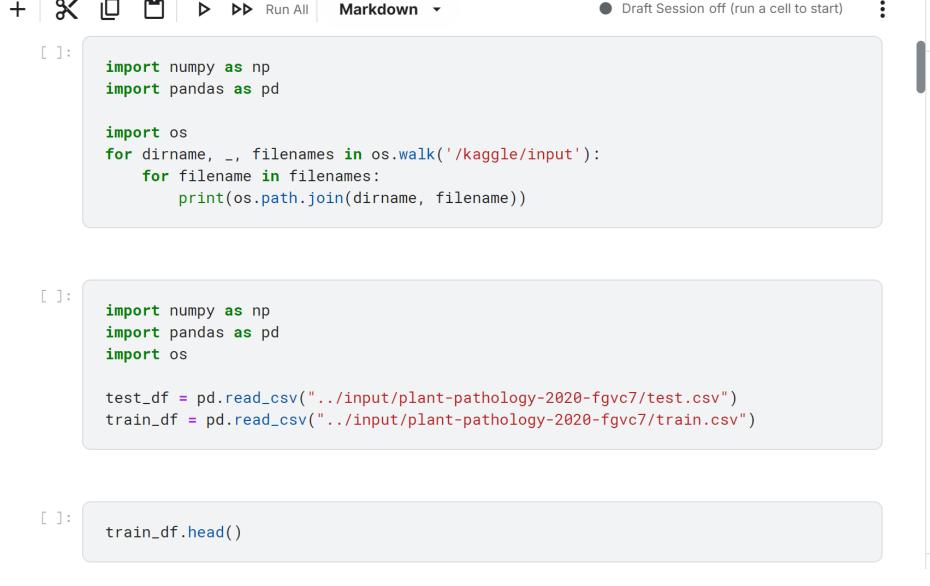
```
import numpy as np
import pandas as pd
import os

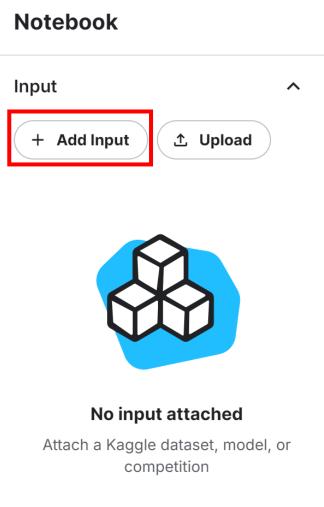
test_df = pd.read_csv("../input/plant-pathology-2020-fgvc7/test.csv")
train_df = pd.read_csv("../input/plant-pathology-2020-fgvc7/train.csv")
```

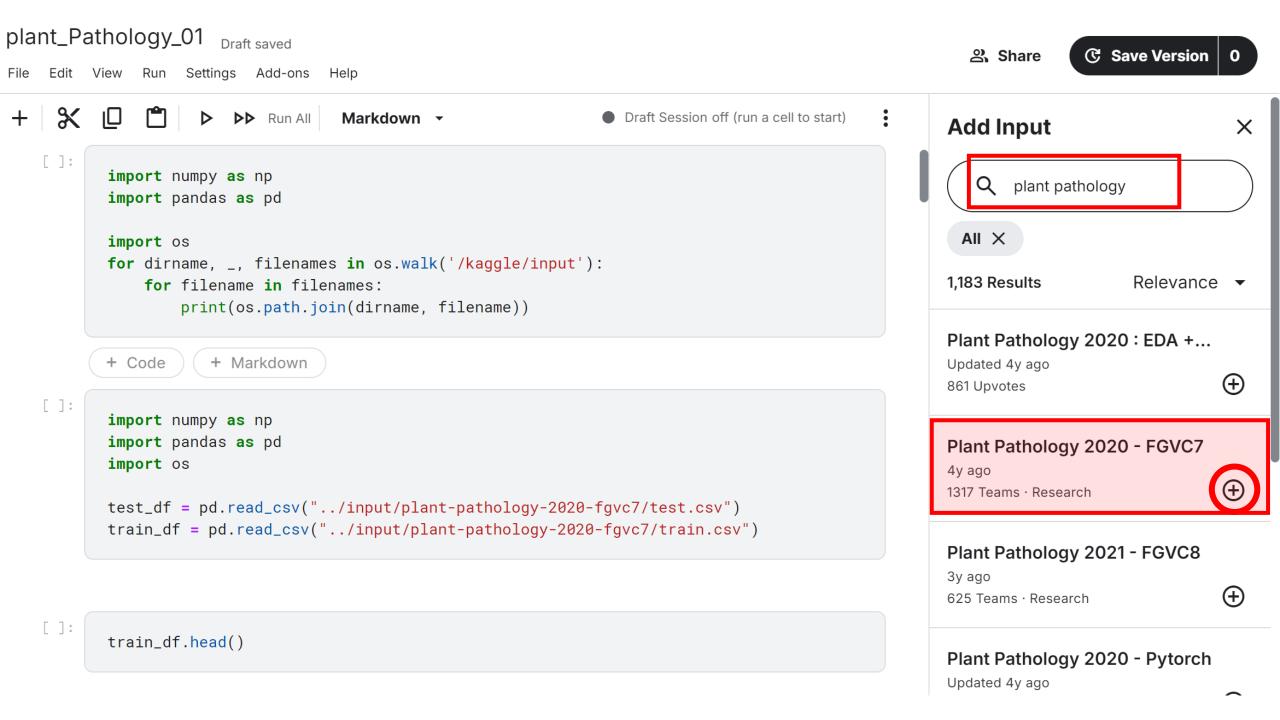
```
train_df.head()
```











All X OS rname

Please read the competition rules

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any late, lost, damaged, misdirected, incomplete, illegible, undeliverable, or destroyed Submissions or entry materials due to system errors, failed, incomplete or garbled computer or other telecommunication transmission malfunctions, hardware or software failures of any kind, lost or unavailable network connections, typographical or system/human errors and failures, technical malfunction(s) of any telephone network or lines, cable connections, satellite transmissions, servers or providers, or computer equipment, traffic congestion on the Internet or at the Competition Website, or any combination thereof, which may limit a participant's ability to participate.

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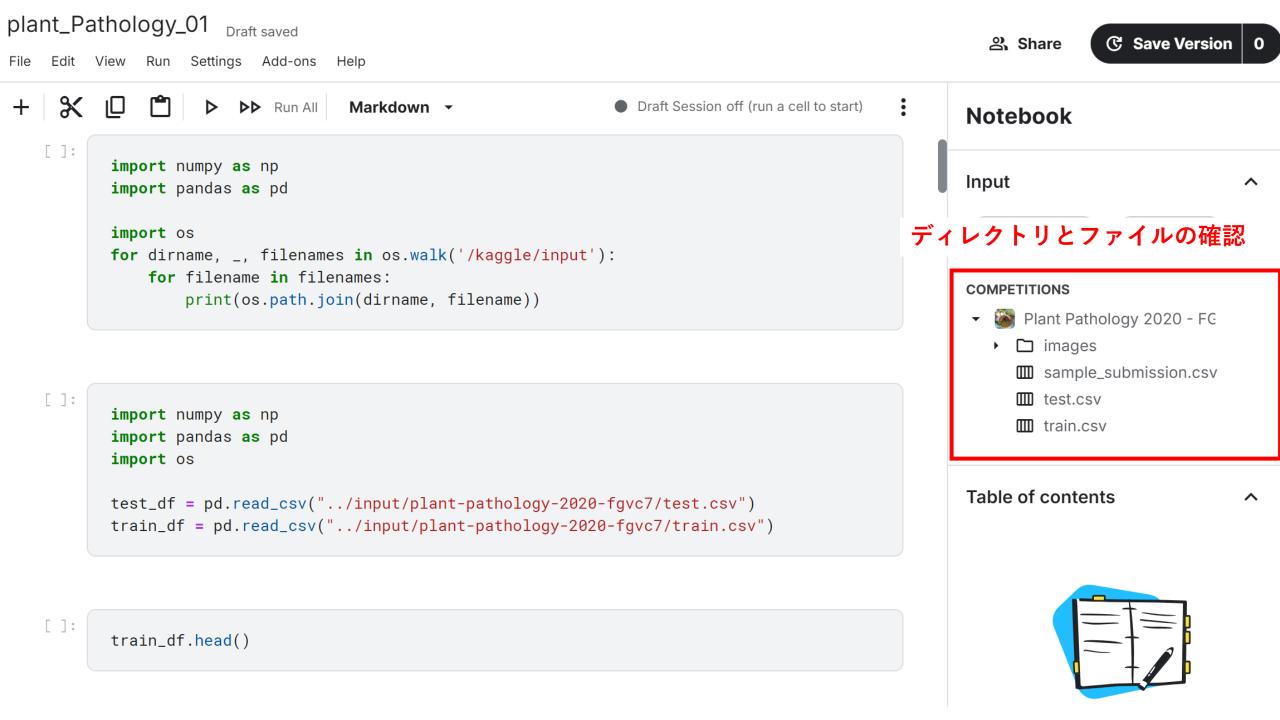
= p

df =

df.he

OS

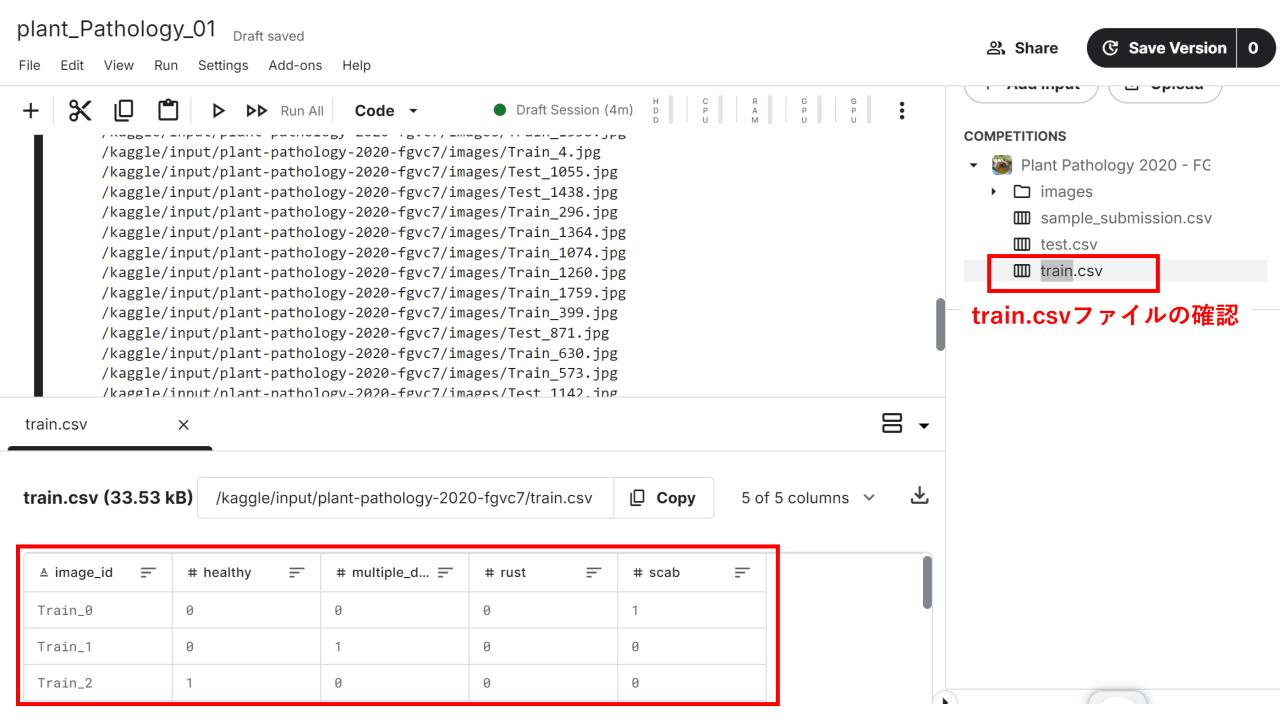
mu

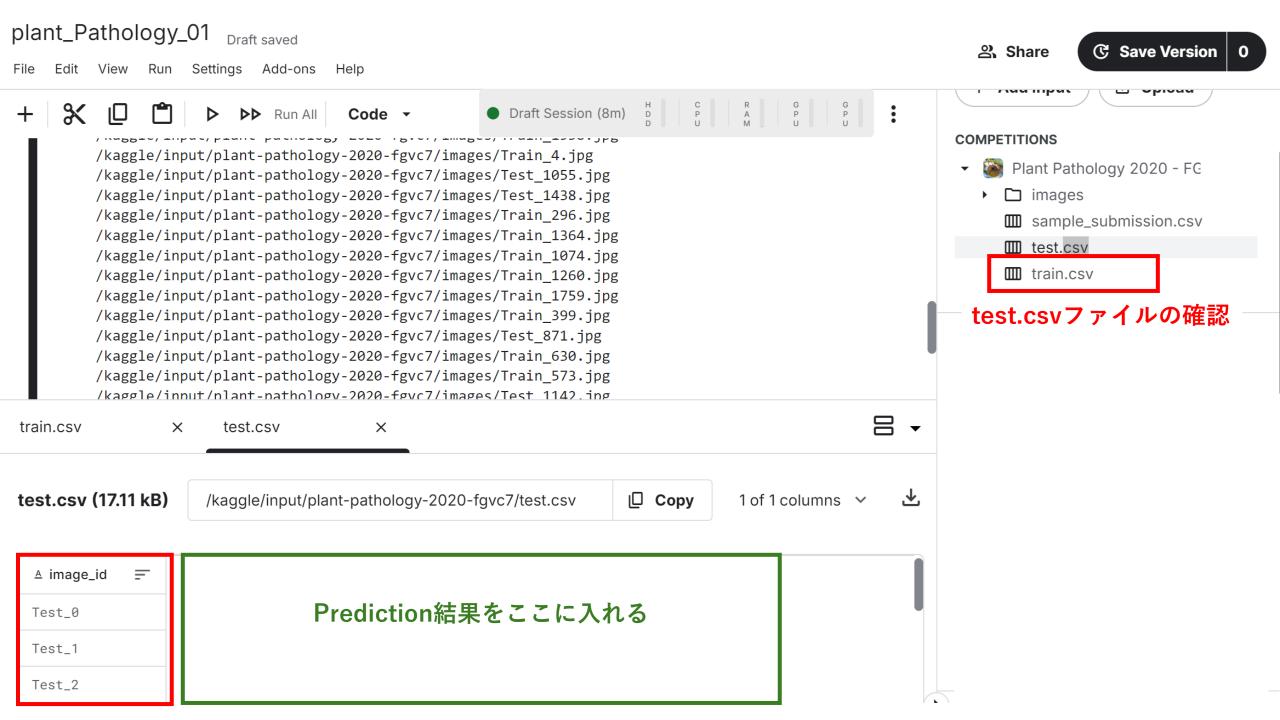


一緒に入っている

```
import numpy as np
  import pandas as pd
  import os
  for dirname, _, filenames in os.walk('/kaggle/input'):
      for filename in filenames:
          print(os.path.join(dirname, filename))
/kaggle/input/plant-pathology-2020-fgvc7/sample submission.csv
/kaggle/input/plant-pathology-2020-fgvc7/train.csv
/kaggle/input/plant-pathology-2020-fgvc7/test.csv
/kaggle/input/plant-pathology-2020-fgvc7/images/Test_1743.jpg
                                                                 imagesのディレクトリに
/kaggle/input/plant-pathology-2020-fgvc7/images/Test_262.jpg
                                                                 TrainとTestファイルが
```

/kaggle/input/plant-pathology-2020-fgvc7/images/Train 1524.jpg





```
[2]:
       import numpy as np
                                         testとtrainファイルをpandasで読み込む
       import pandas as pd
       import os
       test_df = pd.read_csv("../input/plant-pathology-2020-fgvc7/test.csv")
       train_df = pd.read_csv("../input/plant-pathology-2020-fgvc7/train.csv")
                  + Markdown
       + Code
train_df.head()
                                         trainデータフレームを確認
[3]:
       image_id healthy multiple_diseases rust scab
         Train 0
                   0
                                     0
         Train_1
                   0
                                     0
                                         0
                                               LabelがOne hot encodingされている
         Train_2
         Train_3
                   0
         Train 4
```

```
# healthy, multiple_diseases, rust, scab 列を合計しsumを作ってsumがしより大きいか, 0かを確認
train_df['sum'] = train_df['healthy'] + train_df['multiple_diseases'] + train_df['rust'] + train_df['scab']
train_df[(train_df['sum'] > 1) | (train_df['sum']==0)]
```

image_id healthy multiple_diseases rust scab sum

One hot encodingされているLabelにエラーがあるか確認する

```
pd.set_option("max_colwidth", 100)

IMAGE_DIR = '/kaggle/input/plant-pathology-2020-fgvc7/images'
train_df['path'] = IMAGE_DIR + '/' + train_df['image_id'] + '.jpg'
train_df.head()
```

画像の絶対パスをDataFrameのpathに作成する

| | image_id | healthy | multiple_diseases | rust | scab | sum | path |
|---|----------|---------|-------------------|------|------|-----|---|
| 0 | Train_0 | 0 | 0 | 0 | 1 | 1 | /kaggle/input/plant-pathology-2020-fgvc7/images/Train_0.jpg |
| 1 | Train_1 | 0 | 1 | 0 | 0 | 1 | /kaggle/input/plant-pathology-2020-fgvc7/images/Train_1.jpg |
| 2 | Train_2 | 1 | 0 | 0 | 0 | 1 | /kaggle/input/plant-pathology-2020-fgvc7/images/Train_2.jpg |
| 3 | Train_3 | 0 | 0 | 1 | 0 | 1 | /kaggle/input/plant-pathology-2020-fgvc7/images/Train_3.jpg |
| 4 | Train_4 | 1 | 0 | 0 | 0 | 1 | /kaggle/input/plant-pathology-2020-fgvc7/images/Train_4.jpg |

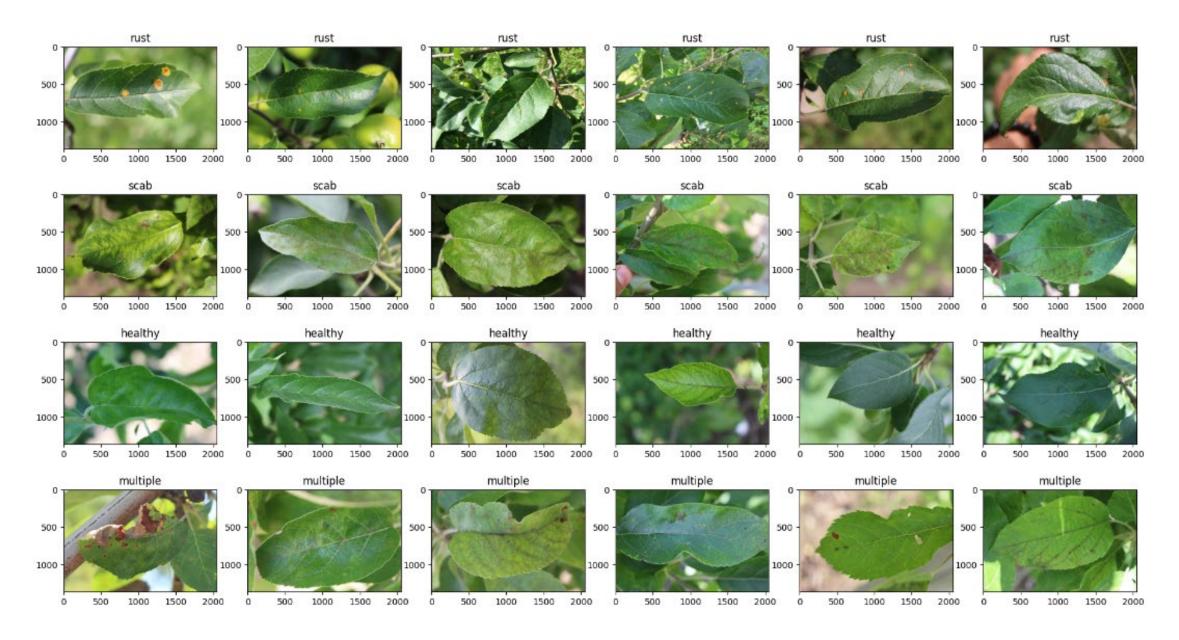
```
def get_label(x):
   if x['healthy'] == 1:
       return 'healthy'
   elif x['multiple_diseases'] == 1:
       return 'multiple_diseases'
   elif x['rust'] == 1:
       return 'rust'
   elif x['scab'] == 1:
       return 'scab'
   else: return 'None'
train_df['label'] = train_df.apply(lambda x:get_label(x), axis=1)
train_df.head()
                                                        学習画像のlabelを分かりやすく作成する
```

| | image_id | healthy | multiple_diseases | rust | scab | sum | path | label |
|---|----------|---------|-------------------|------|------|-----|---|-------------------|
| 0 | Train_0 | 0 | 0 | 0 | 1 | 1 | /kaggle/input/plant-pathology-2020-fgvc7/images/Train_0.jpg | scab |
| 1 | Train_1 | 0 | 1 | 0 | 0 | 1 | /kaggle/input/plant-pathology-2020-fgvc7/images/Train_1.jpg | multiple_diseases |
| 2 | Train_2 | 1 | 0 | 0 | 0 | 1 | /kaggle/input/plant-pathology-2020-fgvc7/images/Train_2.jpg | healthy |
| 3 | Train_3 | 0 | 0 | 1 | 0 | 1 | /kaggle/input/plant-pathology-2020-fgvc7/images/Train_3.jpg | rust |
| 4 | Train_4 | 1 | 0 | 0 | 0 | 1 | /kaggle/input/plant-pathology-2020-fgvc7/images/Train_4.jpg | healthy |

```
##labelごとに学習イメージが何枚か数える
   print('train shape:', train_df.shape)
   print('labelごとの学習画像数:')
   train_df['label'].value_counts()
 train shape: (1821, 8)
 labelごとの学習画像数:
: label
 rust
                    622
 scab
                    592
 healthy
                    516
                         他のlabelに比べてデータ件数が少ない
 multiple_diseases
                    91
 Name: count, dtype: int64
```

元画像の可視化

```
(1365, 2048, 3)
import seaborn as sns
                                                                                                              (1365, 2048, 3)
import matplotlib.pyplot as plt
                                                                                                              (1365, 2048, 3)
import cv2
                                                                                                              (1365, 2048, 3)
%matplotlib inline
                                                                                                              (1365, 2048, 3)
def show_grid_images(image_path_list, augmentor=None, ncols=4, title=None):
                                                                                                              (1365, 2048, 3)
    figure, axs = plt.subplots(figsize=(22, 4), nrows=1, ncols=ncols)
                                                                                                              (1365, 2048, 3)
    for i in range(ncols):
                                                                                                              (1365, 2048, 3)
        image = cv2.cvtColor(cv2.imread(image_path_list[i]), cv2.COLOR_BGR2RGB)
                                                                                                              (1365, 2048, 3)
        if augmentor is not None:
                                                                                                              (1365, 2048, 3)
            image = augmentor(image=image)['image']
                                                                                                              (1365, 2048, 3)
        axs[i].imshow(image)
                                                                                                              (1365, 2048, 3)
       #axs[i].axis('off')
                                                                                                              (1365, 2048, 3)
        axs[i].set_title(title)
                                                                                                              (1365, 2048, 3)
        print(image.shape)
                                                                                                              (1365, 2048, 3)
                                                                                                              (1365, 2048, 3)
rust_image_list = train_df[train_df['label'] == 'rust']['path'].iloc[:6].tolist()
                                                                                                              (1365, 2048, 3)
scab_image_list = train_df[train_df['label']=='scab']['path'].iloc[:6].tolist()
                                                                                                              (1365, 2048, 3)
healthy_image_list = train_df[train_df['label'] == 'healthy']['path'].iloc[:6].tolist()
                                                                                                              (1365, 2048, 3)
multiple_image_list = train_df[train_df['label'] == 'multiple_diseases']['path'].iloc[:6].tolist()
                                                                                                              (1365, 2048, 3)
                                                                                                              (1365, 2048, 3)
show_grid_images(rust_image_list, ncols=6, title='rust')
show_grid_images(scab_image_list, ncols=6, title='scab')
                                                                                                              (1365, 2048, 3)
show_grid_images(healthy_image_list, ncols=6, title='healthy')
                                                                                                              (1365, 2048, 3)
show_grid_images(multiple_image_list, ncols=6, title='multiple')
                                                                                                              (1365, 2048, 3)
```

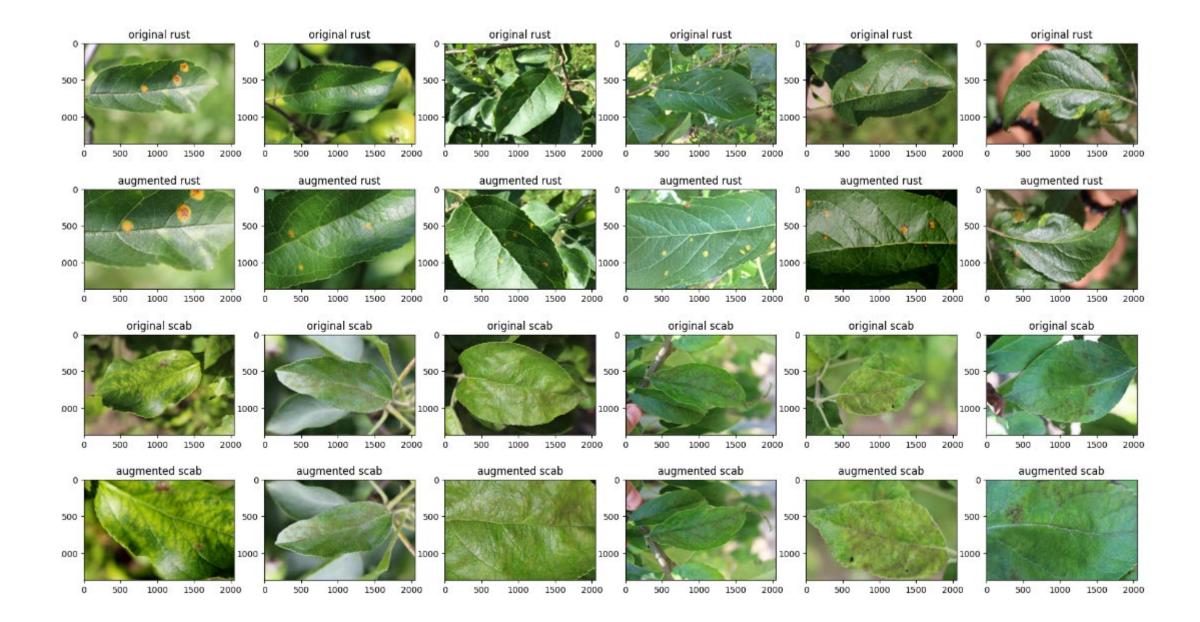


色が大事な情報になる

Augmentationを適用

- *カットアウトのようなノイズは、葉の病原菌斑点と混同される可能性があるため、使用しない。
- * 全体画像が緑系であり、病原菌斑点が特定の色(オレンジ,ブラウンなど)を持っているため、色の変化は適用しない。
- * 全体的に判別したい葉が画像全体の中央に位置しているので、スケール等の適用を検討。

```
import albumentations as A
augmentor_01 = A.Compose([
   A.HorizontalFlip(p=0.5),
                                自分が自由に工夫して適用する
   A. VerticalFlip(p=0.5),
   A.ShiftScaleRotate(scale_limit=(0.7, 0.9), p=0.5, rotate_limit=30),
   A.RandomBrightnessContrast(brightness_limit=(-0.2, 0.2), contrast_limit=(-0.2, 0.2), p=0.5),
   A.Blur(p=0.2)
])
show_grid_images(rust_image_list, augmentor=None, ncols=6, title='original rust')
show_grid_images(rust_image_list, augmentor=augmentor_01, ncols=6, title='augmented rust')
show_grid_images(scab_image_list, augmentor=None, ncols=6, title='original scab')
show_grid_images(scab_image_list, augmentor=augmentor_01, ncols=6, title='augmented scab')
```



SequenceベースのDataset生成

- 従来はimage sizeが高さと幅が同じでしたが、今回は高さと幅が違う場合があることを考慮してimage_sizeをtupleで入力する。
- opencvのresize()は引数で画像サイズを入力しますが、横x縦(幅x高さ)の概念で入力します。画像配列の場合は行x列(高さx幅)なので、resize()呼び出し時にこれを考慮する。

* キャグルコンペティションにテストデータの結果をsubmitするため、 テスト日付のLabelがない。 そのため、Datasetのlabel_batchの値が Noneになる可能性があるため、コードの修正が必要。

```
from sklearn.model_selection import train_test_split
from tensorflow.keras.utils import Sequence
import sklearn
import cv2
#入力引数image_filenames、labelsは、全てnumpy arrayで入る。image sizeは(高さ、幅)で修正。
class Plant Dataset(Sequence):
   def __init__(self, image_filenames, labels, image_size=(224, 224), batch_size=64,
               augmentor=None, shuffle=False, pre_func=None): tuple
       1.1.1
       パラメータの説明
       image_filenames: opencvでimageをロードするファイルの絶対パス。
       labels: 該当するimageのラベル
       batch_size: __getitem__(self, index)呼び出しごとに取り込むデータのバッチ数。
       augmentor: albumentationsのオブジェクト。
       shuffle: 学習データの場合、epoch終了時ごとにデータを混ぜるかどうかを決める。
       # オブジェクト牛成引数として入ってきた値をオブジェクト内部変数に割り当てる。
       self.image_filenames = image_filenames
       self.labels = labels
       self.image_size = image_size
       self.batch size = batch size
       self.augmentor = augmentor
       self.pre_func = pre_func
       # train dataの場合
       self.shuffle = shuffle
       if self.shuffle:
          # オブジェクト生成する時一回Shuffleする
          #self.on_epoch_end()
          pass
```

```
# Sequenceを継承するDatasetは batch_size 単位で入力データを処理する
# __len__()は全体データ数が与えられた時、batch_size単位でデータを何回返すかを表す
def __len__(self):
   return int(np.ceil(len(self.image_filenames) / self.batch_size))
def __getitem__(self, index):
   # indexは何回目のbatchかを表す.
   # batch sizeごとに順次的にデータを持ってくるように記述する。
   image_name_batch = self.image_filenames[index*self.batch_size:(index+1)*self.batch_size]
   if self.labels is not None:
       label_batch = self.labels[index*self.batch_size:(index+1)*self.batch_size]
   # label_batchが Noneになる場合
   else:
       label batch = None
   # もし、オブジェクト生成引数にalbumentationで作ったaugmentorが与えられた場合、下記のようにaugmentorを利用してimage変換する。
   # albumentationsは個々のimageしか変換できないので、batch_sizeだけ割り当てられたimage_name_batchを1件ずつiterationしながら変換
   # image_batch 配列は float32 に設定。
   image_batch = np.zeros((image_name_batch.shape[0], self.image_size[0], self.image_size[1], 3), dtype='float32')
   for image_index in range(image_name_batch.shape[0]):
       image = cv2.cvtColor(cv2.imread(image_name_batch[image_index]), cv2.COLOR_BGR2RGB)
       if self.augmentor is not None:
           image = self.augmentor(image=image)['image']
       #resize適用. opencvのresizeは (幅, 縦)の概念. 配列は (縦, 幅)になるので注意しながらopencvのresizeに引数を入力する.
       image = cv2.resize(image, (self.image_size[1], self.image_size[0]))
       # もし preprocessing_inputが pre_func引数に入るとこれを利用し scalingを適用.
       if self.pre_func is not None:
           image = self.pre_func(image)
       image_batch[image_index] = image
   if self.labels is None:
       return image_batch
   else:
       return image_batch, label_batch
```

学習データ用DataFrameから学習用/検証用画像の絶対パスとLabelを抽出し、これをDatasetとして作成。

- すでに学習用DataFrameに'healthy'、'multiple_diseases'、'rust'、'scab'の順にワンホットエンコーディング(One hot encoding)されています。
- キャグルでテストデータ予測した結果を'healthy','multiple_diseases','rust','scab'の順番で提出を要求するので、これを別途 に再度ワンホットエンコーディングしてはいけません。
- Augmentation は先に生成した augmentor_01 を適用。pre_funcはxception用のPreprocessing関数を適用。





sample_df = pd.read_csv('/kaggle/input/plant-pathology-2020-fgvc7/sample_submission.csv')
sample_df.head()

| :[: | | image_id | healthy | multiple_diseases | rust | scab |
|-----|---|----------|---------|-------------------|------|------|
| | 0 | Test_0 | 0.25 | 0.25 | 0.25 | 0.25 |
| | 1 | Test_1 | 0.25 | 0.25 | 0.25 | 0.25 |
| | 2 | Test_2 | 0.25 | 0.25 | 0.25 | 0.25 |
| | 3 | Test_3 | 0.25 | 0.25 | 0.25 | 0.25 |
| | 4 | Test_4 | 0.25 | 0.25 | 0.25 | 0.25 |

```
from sklearn.model_selection import train_test_split
def get_train_valid train_df, valid_size=0.2, random_state=2021):
    train_path = train_df['path'].values
    train_label = train_df[['healthy', 'multiple_diseases', 'rust', 'scab']].values
    tr_path, val_path, tr_label, val_label = train_test_split(train_path, train_label, test_size=valid_size, random_state=random_state)
    print('tr_path shape:', tr_path.shape, 'tr_label shape:', tr_label.shape, 'val_path shape:', val_path.shape, 'val_label shape:', val_label.shape)
    return tr_path, val_path, tr_label, val_label
           + Markdown
+ Code
from tensorflow.keras.applications.xception import preprocess_input as xcp_preprocess_input
from tensorflow.keras.applications.efficientnet import preprocess_input as eff_preprocess_input
# image size는 224x224로 Dataset 생성.
IMAGE\_SIZE = (224, 224)
BATCH_SIZE = 64
tr_path, val_path, tr_label, val_label = get_train_valid(train_df, valid_size=0.2, random_state=2021)
tr_ds = Plant_Dataset(tr_path, tr_label, image_size=IMAGE_SIZE, batch_size=BATCH_SIZE,
                          augmentor=augmentor_01, shuffle=True, pre_func=xcp_preprocess_input)
val_ds = Plant_Dataset(val_path, val_label, image_size=IMAGE_SIZE, batch_size=BATCH_SIZE,
                      augmentor=None, shuffle=False, pre_func=xcp_preprocess_input)
tr_image_batch, tr_label_batch = next(iter(tr_ds))
val_image_batch, val_label_batch = next(iter(val_ds))
print(tr_image_batch.shape, val_image_batch.shape, tr_label_batch.shape, val_label_batch.shape)
print(tr_image_batch[0], val_image_batch[0])
```

```
### create model() 関数を生成
* resnet50v2, xception, efficientnetb0~b7 などの Pretrained モデルを生成
from tensorflow.keras.models import Sequential , Model
from tensorflow.keras.layers import Input, Dense , Conv2D , Dropout , Flatten , Activation, MaxPooling2D , GlobalAveragePooling2D
from tensorflow.keras.optimizers import Adam , RMSprop
from tensorflow.keras.regularizers import 12
from tensorflow.keras.layers import BatchNormalization
from tensorflow.keras.callbacks import ReduceLROnPlateau , EarlyStopping , ModelCheckpoint , LearningRateScheduler
from tensorflow.keras.metrics import AUC
from tensorflow.keras.applications import Xception, ResNet50V2, EfficientNetB0, EfficientNetB1, EfficientNetB2, EfficientNetB3
from tensorflow.keras.applications import EfficientNetB4, EfficientNetB5, EfficientNetB6, EfficientNetB7
import tensorflow as tf
def create_model(model_type='efficientnetb0', in_shape=(224, 224, 3), n_classes=4):
    input_tensor = Input(shape=in_shape)
    if model_type == 'resnet50v2':
        base_model = tf.keras.applications.ResNet50V2(include_top=False, weights='imagenet', input_tensor=input_tensor)
    elif model_type == 'xception':
        base_model = tf.keras.applications.Xception(include_top=False, weights='imagenet', input_tensor=input_tensor)
    elif model_type == 'efficientnetb0':
        base_model = tf.keras.applications.EfficientNetB0(include_top=False, weights='imagenet', input_tensor=input_tensor)
    elif model_type == 'efficientnetb1':
        base_model = tf.keras.applications.EfficientNetB1(include_top=False, weights='imagenet', input_tensor=input_tensor)
    elif model_type == 'efficientnetb2':
        base_model = tf.keras.applications.EfficientNetB2(include_top=False, weights='imagenet', input_tensor=input_tensor)
```

```
elif model_type == 'efficientnetb3':
    base_model = tf.keras.applications.EfficientNetB3(include_top=False, weights='imagenet', input_tensor=input_tensor)
elif model_type == 'efficientnetb4':
    base_model = tf.keras.applications.EfficientNetB4(include_top=False, weights='imagenet', input_tensor=input_tensor)
elif model_type == 'efficientnetb5':
    base_model = tf.keras.applications.EfficientNetB5(include_top=False, weights='imagenet', input_tensor=input_tensor)
elif model_type == 'efficientnetb6':
    base_model = tf.keras.applications.EfficientNetB6(include_top=False, weights='imagenet', input_tensor=input_tensor)
elif model_type == 'efficientnetb7':
    base_model = tf.keras.applications.EfficientNetB7(include_top=False, weights='imagenet', input_tensor=input_tensor)
```

```
x = base_model.output
x = GlobalAveragePooling2D()(x)
x = Dense(1024, activation='relu')(x)
x = Dropout(0.5)(x)
preds = Dense(units=n_classes, activation='softmax')(x)
model = Model(inputs=input_tensor, outputs=preds)

return model
```

xception モデルを生成して学習を行う。

- image sizeは224x224で生成。
- Learning Rate SchedulerはReduceLROnPlateauで、初期Learning Rateは0.0001に設定。
- epochsは10回のみ設定。
- metricsはROC-AUCを設定

```
Epoch 2/10
23/23 -
                          46s 2s/step - auc 1: 0.9517 - loss: 0.5511 - val auc 1: 0.9217 - val loss: 0.8791 - learning rate: 1.0000e-04
Epoch 3/10
                          47s 2s/step - auc 1: 0.9722 - loss: 0.3987 - val auc 1: 0.9471 - val_loss: 0.6718 - learning_rate: 1.0000e-04
23/23 -
Epoch 4/10
                         - 46s 2s/step - auc 1: 0.9891 - loss: 0.2457 - val auc 1: 0.9739 - val loss: 0.3922 - learning rate: 1.0000e-04
23/23 -
Epoch 5/10
23/23 -
                          47s 2s/step - auc 1: 0.9943 - loss: 0.1843 - val auc 1: 0.9744 - val loss: 0.3551 - learning rate: 1.0000e-04
Epoch 6/10
                          46s 2s/step - auc 1: 0.9950 - loss: 0.1630 - val auc 1: 0.9683 - val_loss: 0.4173 - learning_rate: 1.0000e-04
23/23 ----
Epoch 7/10
23/23 -
                          47s 2s/step - auc 1: 0.9952 - loss: 0.1520 - val auc 1: 0.9777 - val loss: 0.3000 - learning rate: 1.0000e-04
Epoch 8/10
23/23 -
                          46s 2s/step - auc 1: 0.9980 - loss: 0.1102 - val auc 1: 0.9803 - val loss: 0.3092 - learning rate: 1.0000e-04
Epoch 9/10
23/23 -
                          46s 2s/step - auc 1: 0.9983 - loss: 0.0978 - val auc 1: 0.9806 - val loss: 0.3230 - learning rate: 1.0000e-04
Epoch 10/10
23/23
                          0s 1s/step - auc 1: 0.9987 - loss: 0.0804
Epoch 10: ReduceLROnPlateau reducing learning rate to 1.9999999494757503e-05.
                           46s 2s/step auc 1: 0.9987 loss: 0.0804 - val auc 1: 0.9757 - val loss: 0.3999 - learning rate: 1.0000e-04
23/23
```

テストデータでPlantの病気を予測してキャグルに提出するsubmit csvファイルを作成します。

- テスト用DataFrameに画像パスを追加。
- テスト用Datasetを生成します。labelはテストデータでは分からないのでNoneで入力。

```
sample_df = pd.read_csv('/kaggle/input/plant-pathology-2020-fgvc7/sample_submission.csv')
sample_df.head()
```

| | image_id | healthy | multiple_diseases | rust | scab |
|---|----------|---------|-------------------|------|------|
| 0 | Test_0 | 0.25 | 0.25 | 0.25 | 0.25 |
| 1 | Test_1 | 0.25 | 0.25 | 0.25 | 0.25 |
| 2 | Test_2 | 0.25 | 0.25 | 0.25 | 0.25 |
| 3 | Test_3 | 0.25 | 0.25 | 0.25 | 0.25 |
| 4 | Test_4 | 0.25 | 0.25 | 0.25 | 0.25 |
| | | | | | |

test_df.head(10)

絶対パースを追加

| | • | - |
|---|----------|--|
| | image_id | path |
| 0 | Test_0 | /kaggle/input/plant-pathology-2020-fgvc7/images/Test_0.jpg |
| 1 | Test_1 | /kaggle/input/plant-pathology-2020-fgvc7/images/Test_1.jpg |
| 2 | Test_2 | /kaggle/input/plant-pathology-2020-fgvc7/images/Test_2.jpg |
| 3 | Test_3 | /kaggle/input/plant-pathology-2020-fgvc7/images/Test_3.jpg |
| 4 | Test_4 | /kaggle/input/plant-pathology-2020-fgvc7/images/Test_4.jpg |
| 5 | Test_5 | /kaggle/input/plant-pathology-2020-fgvc7/images/Test_5.jpg |
| 6 | Test_6 | /kaggle/input/plant-pathology-2020-fgvc7/images/Test_6.jpg |
| 7 | Test_7 | /kaggle/input/plant-pathology-2020-fgvc7/images/Test_7.jpg |
| 8 | Test_8 | /kaggle/input/plant-pathology-2020-fgvc7/images/Test_8.jpg |
| 9 | Test_9 | /kaggle/input/plant-pathology-2020-fgvc7/images/Test_9.jpg |

| <pre>IMAGE_DIR = '/kaggle/input/plant-pathology-2020-fgvc7/images'</pre> | |
|---|------|
| <pre>test_df = pd.read_csv("/input/plant-pathology-2020-fgvc7/test.org)</pre> | sv") |
| <pre>test_df['path'] = IMAGE_DIR + '/' + test_df['image_id'] + '.jpg'</pre> | |
| | |



+ Markdown

+ Code

```
preds_df = pd.DataFrame(preds)
preds_df.columns = ['healthy', 'multiple_diseases', 'rust', 'scab']
preds_df.head()
```

テスト用Datasetを生成し、これを利用してmodelのpredict()を呼び出して画像予測を実行します。

| | healthy | multiple_diseases | rust | scab |
|---|--------------|-------------------|--------------|--------------|
| 0 | 1.052231e-05 | 0.000134 | 9.998443e-01 | 1.091581e-05 |
| 1 | 4.836722e-07 | 0.000093 | 9.999059e-01 | 2.637625e-07 |
| 2 | 1.032351e-07 | 0.000002 | 7.365213e-08 | 9.999974e-01 |
| 3 | 9.992207e-01 | 0.000016 | 7.503788e-04 | 1.260714e-05 |
| 4 | 1.137724e-05 | 0.000193 | 9.997917e-01 | 4.101276e-06 |

```
# 予測した結果に基づき、別の結果DataFrameを作成。
preds_df = pd.DataFrame(preds)
preds_df.columns = ['healthy', 'multiple_diseases', 'rust', 'scab']
# テスト用DataFrameに上記で生成した結果DataFrameを合成し、これを利用してsubmit用DataFrameを生成します。
submit_df = pd.concat([test_df['image_id'], preds_df], axis = 1)
submit_df.head()
```

| | image_id | healthy | multiple_diseases | rust | scab | |
|---|----------|--------------|-------------------|--------------|--------------|--|
| 0 | Test_0 | 1.052231e-05 | 0.000134 | 9.998443e-01 | 1.091581e-05 | |
| 1 | Test_1 | 4.836722e-07 | 0.000093 | 9.999059e-01 | 2.637625e-07 | |
| 2 | Test_2 | 1.032351e-07 | 0.000002 | 7.365213e-08 | 9.999974e-01 | |
| 3 | Test_3 | 9.992207e-01 | 0.000016 | 7.503788e-04 | 1.260714e-05 | |
| 4 | Test_4 | 1.137724e-05 | 0.000193 | 9.997917e-01 | 4.101276e-06 | |

提出する最終フォーマットをcsvファイルに作成

キャグル提出用CSVを作成した後、キャグルに提出し、テスト性能を確認する

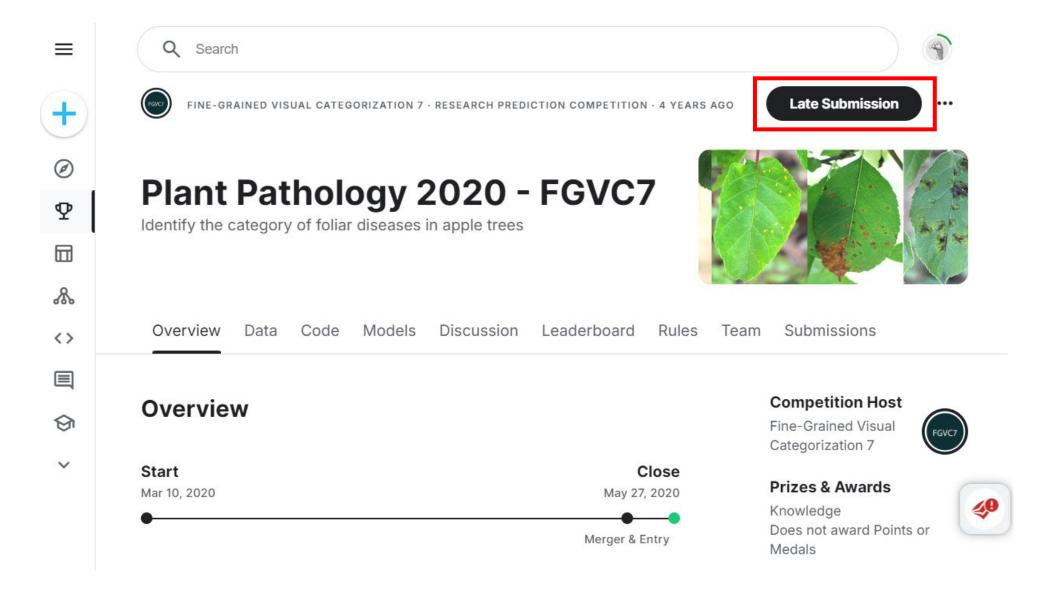
+ Code + Markdown

submit_df.to_csv('submit_01.csv', index=False)

ダウンロードする

Output (180KiB / 19.5GiB)

Kaggleへ提出



X Submit to Competition

File Upload Notebook



Plant Pathology 2020 - FGVC7

You have 100 submissions remaining today. This resets in 8 hours.



Drag and drop file to upload

(e.g., .csv, .zip, .gz, .7z)

Uploaded File

submit_01.csv (105 KiB)

性能向上

Submission and Description

Private Score (i) Public Score (i) Selected

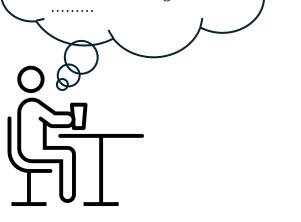
Selected

Submit_01.csv
Complete (after deadline) · now · First submission

O.91530

O.91184

学習モデルを変更 ハイパーパラメータのチューニング 学習率、最適化、Dropout、 Batch normalization Augmentationの拡張 Transfer Learning



| # | Δ | Team | Members | Score | Entries | Last | Solution |
|---|--------------|-----------------------------------|------------|---------|---------|------|----------|
| 1 | - 17 | Alipay Tian Suan Securit y Lab | | 0.98445 | 115 | 4у | |
| 2 | - 94 | Shenlan | | 0.98182 | 15 | 4y | |
| 3 | - 103 | Clay | | 0.98089 | 167 | 4y | |
| 4 | ^ 61 | Mack Tang | | 0.98064 | 28 | 4y | |
| 5 | - 9 | Profit! | 9 9 | 0.98060 | 361 | 4y | |