Mount Drive

```
from google.colab import drive
drive.mount('/content/drive'
    Mounted at /content/drive
```

Prepare the Dataset

```
Import library
```

```
!pip install tensorflowjs
                     Collecting packaging~=20.9
                            Downloading packaging-20.9-py2.py3-none-any.whl (40 kB)
                                                                                                                                              | 40 kB 7.6 MB/s
tensorflow-hub<0.13,>=0.7.0 in /usr/local/lib/python3.7/dist-packages (from tensorflowjs) (0.12.0)
                         Requirement already satisfied:
                   Requirement already satisfied: tensorflow-hubc0.13,>=0.7.0 in /usr/local/lib/python3.7/dist-packages (from tensorflowjs) (0.12.0) Requirement already satisfied: tensorflow(3,>=2.1.0 in /usr/local/lib/python3.7/dist-packages (from tensorflowjs) (2.8.2+zzcolab20220527125636) Requirement already satisfied: six(2,>=1.12.0 in /usr/local/lib/python3.7/dist-packages (from tensorflowjs) (1.15.0) Requirement already satisfied: pyparsing>=2.0.2 in /usr/local/lib/python3.7/dist-packages (from tensorflow(3,>=2.1.0->tensorflow(s)) (3.0.9) Requirement already satisfied: termcolor>=1.1.0 in /usr/local/lib/python3.7/dist-packages (from tensorflow(3,>=2.1.0->tensorflow(s)) (1.1.0) Requirement already satisfied: protobuf(3.20,>=3.9.2 in /usr/local/lib/python3.7/dist-packages (from tensorflow(3,>=2.1.0->tensorflow(s)) (3.17.3) Requirement already satisfied: gast>=0.2.1 in /usr/local/lib/python3.7/dist-packages (from tensorflow(3,>=2.1.0->tensorflow(s)) (5.5.3) Requirement already satisfied: grpcio<2.0,>=1.24.3 in /usr/local/lib/python3.7/dist-packages (from tensorflow(3,>=2.1.0->tensorflow(s)) (5.4.0) Requirement already satisfied: bpy>=2.9.0 in /usr/local/lib/python3.7/dist-packages (from tensorflow(3,>=2.1.0->tensorflow(s)) (3.1.0) Requirement already satisfied: wrapt>=1.1.0 in /usr/local/lib/python3.7/dist-packages (from tensorflow(3,>=2.1.0->tensorflow(s)) (3.1.0) Requirement already satisfied: wrapt>=1.1.0 in /usr/local/lib/python3.7/dist-packages (from tensorflow(3,>=2.1.0->tensorflow(s)) (1.1.1) Requirement already satisfied: wrapt>=1.1.1 in /usr/local/lib/python3.7/dist-packages (from tensorflow(3,>=2.1.0->tensorflow(s),>=2.1.0->tensorflow(s),>=2.1.0->tensorflow(s),>=2.1.0->tensorflow(s),>=2.1.0->tensorflow(s),>=2.1.0->tensorflow(s),>=2.1.0->tensorflow(s),>=2.1.0->tensorflow(s),>=2.1.0->tensorflow(s),>=2.1.0->tensorflow(s),>=2.1.0->tensorflow(s),>=2.1.0->tensorflow(s),>=2.1.0->tensorflow(s),>=2.1.0->tensorflow(s),>=2.1.0->tensorflow(s),>=2.1.0->tensorflow(s),>=2.1.0->tensorflow(s),>=2.1.0->tensorflow(s),>=2.1.0->tensorflo
                Requirement already satisfied: gprioc2.8, >=1.2.4, in /usr/local/lib/python3.7/dist-packages (from tensorflows), >=2.1.8-tensorflows) (1.4.6.3)
Requirement already satisfied: where the processing ==1.1 in /usr/local/lib/python3.7/dist-packages (from tensorflows), >=2.1.8-tensorflows) (1.1.4.1)
Requirement already satisfied: where the processing ==1.1 in /usr/local/lib/python3.7/dist-packages (from tensorflows), >=2.1.8-tensorflows) (1.6.8)
Requirement already satisfied: abs1.py=8.4.8 in /usr/local/lib/python3.7/dist-packages (from tensorflows), >=2.1.8-tensorflows) (2.8.8)
Requirement already satisfied: where the substance of the processing ==1.1 in /usr/local/lib/python3.7/dist-packages (from tensorflows), >=2.1.8-tensorflows) (2.8.8)
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Requirement already satisfied: tensorflow estimator(2.9)=2.8 in /usr/local/lib/python3.7/dist-packages (from tensorflows), >=2.1.8-tensorflows) (2.8.8)
Requirement already satisfied: google.pasts.9-3.8 in /usr/local/lib/python3.7/dist-packages (from tensorflows), >=2.1.8-tensorflows) (2.8.8)
Requirement already satisfied: google.pasts.9-3.1 in /usr/local/lib/python3.7/dist-packages (from tensorflows), >=2.1.8-tensorflows) (2.8.8)
Requirement already satisfied: google.pasts.9-3.1 in /usr/local/lib/python3.7/dist-packages (from tensorflows), >=2.1.8-tensorflows) (2.8.8)
Requirement already satisfied: tensorflow-in-gcs-filesystem-80.23.1 in /usr/local/lib/python3.7/dist-packages (from tensorflows), >=2.1.8-tensorflows) (4.2.8)
Requirement already satisfied: tensorflow-in-gcs-filesystem-80.23.1 in /usr/local/lib/python3.7/dist-packages (from tensorflows), >=2.1.8-tensorflows) (4.2.8)
Requirement already satisfied: tensorflow-in-gcs-filesystem-80.23.1 in /usr/local/lib/python3.7/dist-packages (from tensorflows), >=2.1.8-tensorflows) (4.2.8)
Requirement already satisfied: sheel.1.6, 9=0.23.8 in /usr/local/lib/python3.7/dist-packages (fr
                    Installing collected packages: packaging, tensorflowjs
Attempting uninstall: packaging
Found existing installation: packaging 21.3
                                    Uninstalling packaging-21.3:
                    Successfully uninstalled packaging-21.3

ERROR: pip's dependency resolver does not currently take into account all the packages that are installed. This behaviour is the source of the following dependency conflicts. datascience 0.10.6 requires folium==0.2.1, but you have folium 0.8.3 which is incompatible.
                     Successfully installed packaging-20.9 tensorflowjs-3.18.0
                    4
   import tensorflow as ti
  import subprocess
  Copy dataset from drive
  rm -rf /content/model10Class.zip
  cp -R /content/drive/MyDrive/indo food datasets/jadi/newdataset /content/
  Unzip file
  import zipfile
 # Extract the archive
 local_zip = '/content/1.zip'
zip_ref = zipfile.ZipFile(local_zip, 'r')
  zip ref.extractall('')
  zip_ref.close()
# local_zip = './rps-test-set.zip'
# zip_ref = zipfile.ZipFile(local_zip, 'r')
  # zip_ref.extractall('tmp/rps-test')
  # zip ref.close()
  Delete unused dataset
  food_classes = ['soto','pepes', 'mendoan', 'lumpia', 'martabak']
  for food_class in food_classes:
```

```
subprocess.run(["rm", "-rf", "/content/food-dataset-500/test/"+food_class])
subprocess_run(["rm" "-rf" "/content/food-dataset-500/train/"+food_class])
cp -R /content/food-dataset-500/train /content/drive/MyDrive/indo_food_datasets/jadi/food-dataset-500
```

 ${\tt ls /content/drive/MyDrive/indo_food_datasets/jadi/food-dataset-500/train/klepon \mid wc -large and the large and$

Model

```
Build Model Layer
```

```
model = tf.keras.models.Sequential([
    # Note the input shape is the desired size of the image 150x150 with 3 bytes color
    # This is the first convolution
    tf.keras.layers.Conv2D(32, (3,3), activation='relu', input_shape=(150, 150, 3)),
   tf.keras.layers.MaxPooling2D(),
# The second convolution
   tf.keras.layers.Conv2D(32, (3,3), activation='relu'),
tf.keras.layers.MaxPooling2D(),
    # The third convolution
    tf.keras.layers.Conv2D(64, (3,3), activation='relu'),
    tf.keras.layers.MaxPooling2D(),
    # The fourth convolution
    # tf.keras.layers.Conv2D(128, (3,3), activation='relu'),
    # tf.keras.lavers.MaxPooling2D(2.2),
    # Flatten the results to feed into a DNN
    tf.keras.lavers.Flatten().
    #tf.keras.layers.Dropout(0.5),
    # 512 neuron hidden layer
    tf.keras.layers.Dense(128, activation='relu'),
    tf.keras.layers.Dense(10, activation='softmax')
# Print the model summary
model.summary()
     Model: "sequential"
```

Layer (type) Output Shape conv2d (Conv2D) (None, 148, 148, 32) max_pooling2d (MaxPooling2D (None, 74, 74, 32) (None, 72, 72, 32) conv2d_1 (Conv2D) 9248 max_pooling2d_1 (MaxPooling (None, 36, 36, 32) conv2d_2 (Conv2D) (None, 34, 34, 64) max_pooling2d_2 (MaxPooling (None, 17, 17, 64) flatten (Flatten) (None, 18496) 0 2367616 dense (Dense) (None, 128) Total params: 2,397,546 Trainable params: 2,397,546 Non-trainable params: 0

Compile Model

```
# Set the training parameters
model.compile(loss = 'categorical_crossentropy', optimizer=tf.keras.optimizers.Adam(), metrics=['accuracy'])
```

Prepare the ImageDataGenerator

```
from keras_preprocessing.image import ImageDataGenerator
TRAINING_DIR = "/content/food-dataset-500/train"
training_datagen = ImageDataGenerator(
       rescale = 1./255,
rotation_range=40,
       width_shift_range=0.2,
       height_shift_range=0.2,
       shear_range=0.2,
zoom_range=0.2,
       horizontal_flip=True,
       fill_mode='nearest')
VALIDATION_DIR = "/content/food-dataset-500/test"
validation_datagen = ImageDataGenerator(rescale = 1./255)
train_generator = training_datagen.flow_from_directory(
  TRAINING DIR.
  target_size=(150,150),
  class_mode='categorical',
  batch_size=150
validation_generator = validation_datagen.flow_from_directory(
  VALIDATION DIR,
  target_size=(150,150),
class_mode='categorical',
  batch_size=150
  #batch_size=126
```

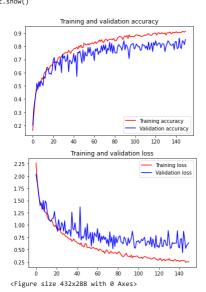
```
Found 4160 images belonging to 10 classes. Found 1000 images belonging to 10 classes.
```

Train the model and evaluate the results

```
Define Callback
class myCallback(tf.keras.callbacks.Callback):
  def on_epoch_end(self, epoch, logs={}):
   Halts the training after reaching 60 percent accuracy
      epoch (integer) - index of epoch (required but unused in the function definition below)
    logs (dict) - metric results from the training epoch
    # Check accuracy
    # if(logs.get('loss') < 0.4):
        # Stop if threshold is met
       print("\nLoss is lower than 0.4 so cancelling training!")
        self.model.stop_training = True
    if(logs.get('val\_accuracy') \,>\, 0.8 \,\, and \,\, logs.get('accuracy') \,\,>\, 0.8);\\
      # Stop if threshold is met
      \begin{array}{l} \cdot \\ \text{print("\nVal\_accuracy is higher than 0.8 so cancelling training!")} \end{array}
      self.model.stop training = True
# Instantiate class
callbacks = myCallback()
callbacks = tf.keras.callbacks.ReduceLROnPlateau(monitor='val_loss', factor=0.2,
                              patience=5, min_lr=0.001)
Train Model
history = model.fit(train_generator, epochs=150, validation_data = validation_generator, verbose = 1, validation_steps=3, callbacks=[callbacks])
     Epoch 122/150
     28/28 [=
                         Epoch 123/150
                                   ======= ] - 28s 1s/step - loss: 0.2935 - accuracy: 0.8990 - val loss: 0.7502 - val accuracy: 0.8156 - lr: 0.0010
     28/28 [=
     Enoch 124/150
     28/28 [=====
Epoch 125/150
                                 :=======] - 27s 954ms/step - loss: 0.2718 - accuracy: 0.9036 - val_loss: 0.5655 - val_accuracy: 0.8244 - lr: 0.0010
     28/28 [===
                               ========] - 27s 946ms/step - loss: 0.2950 - accuracy: 0.8945 - val_loss: 0.5584 - val_accuracy: 0.8244 - lr: 0.0010
           126/150
                                   =======] - 27s 958ms/step - loss: 0.3019 - accuracy: 0.8959 - val_loss: 0.6952 - val_accuracy: 0.7889 - lr: 0.0010
     28/28 [==
     Enoch 127/150
     28/28
                               ========] - 27s 959ms/step - loss: 0.2680 - accuracy: 0.9036 - val_loss: 0.6859 - val_accuracy: 0.8156 - lr: 0.0010
     Epoch 128/150
     28/28 [=
                                 =======] - 27s 953ms/step - loss: 0.2872 - accuracy: 0.8974 - val loss: 0.5926 - val accuracy: 0.8200 - lr: 0.0010
     Epoch 12
28/28 [=
           129/150
                                        ===] - 27s 949ms/step - loss: 0.2947 - accuracy: 0.8969 - val_loss: 0.5775 - val_accuracy: 0.8156 - lr: 0.0010
     Epoch 130/150
     28/28
                                ========] - 26s 943ms/step - loss: 0.2820 - accuracy: 0.9017 - val_loss: 0.8875 - val_accuracy: 0.7600 - lr: 0.0010
           131/150
     Epoch
     28/28 [====
                                  :======] - 27s 962ms/step - loss: 0.3160 - accuracy: 0.8851 - val_loss: 0.6898 - val_accuracy: 0.7956 - lr: 0.0010
     Epoch 132/150
28/28 [=====
                               ========] - 29s 1s/step - loss: 0.2837 - accuracy: 0.9012 - val_loss: 0.5230 - val_accuracy: 0.8200 - lr: 0.0010
     Epoch 133/150
     28/28 [
                             ========] - 27s 950ms/step - loss: 0.2760 - accuracy: 0.9029 - val_loss: 0.5399 - val_accuracy: 0.8422 - lr: 0.0010
     Epoch 134/150
     28/28 T==
                                   :======] - 27s 967ms/step - loss: 0.2810 - accuracy: 0.9005 - val loss: 0.6409 - val accuracy: 0.8311 - lr: 0.0010
     Epoch 135/150
     28/28 [=
                                  =======] - 27s 962ms/step - loss: 0.2629 - accuracy: 0.9094 - val_loss: 0.6185 - val_accuracy: 0.8156 - lr: 0.0010
     Epoch 136/150
     28/28 [===
                             ========] - 27s 963ms/step - loss: 0.2868 - accuracy: 0.8959 - val_loss: 0.6319 - val_accuracy: 0.8067 - lr: 0.0010
           137/150
     Epoch
                                  :======] - 27s 960ms/step - loss: 0.2949 - accuracy: 0.8988 - val loss: 0.6330 - val accuracy: 0.8089 - lr: 0.0010
     28/28 [===
     Fnoch 138/150
     28/28 [
                              =========] - 27s 969ms/step - loss: 0.2663 - accuracy: 0.9079 - val loss: 0.6295 - val accuracy: 0.8200 - lr: 0.0010
     Epoch 139/150
    28/28 [=====
Fnoch 140/150
                                 ========] - 27s 954ms/step - loss: 0.2767 - accuracy: 0.9029 - val loss: 0.8861 - val accuracy: 0.7756 - lr: 0.0010
                                   =======] - 27s 955ms/step - loss: 0.2762 - accuracy: 0.9038 - val_loss: 0.5534 - val_accuracy: 0.8356 - lr: 0.0010
     28/28 [==
     Epoch 141/150
     28/28
                             =========] - 27s 958ms/step - loss: 0.2554 - accuracy: 0.9111 - val_loss: 0.5333 - val_accuracy: 0.8356 - lr: 0.0010
     Epoch 142/150
     28/28 [===
                                 ========] - 27s 949ms/step - loss: 0.2702 - accuracy: 0.9022 - val loss: 0.7120 - val accuracy: 0.7822 - lr: 0.0010
           143/150
                                   ======] - 27s 962ms/step - loss: 0.2796 - accuracy: 0.9046 - val_loss: 0.7233 - val_accuracy: 0.7889 - lr: 0.0010
     Epoch 144/150
     28/28 [=
                               ========] - 27s 959ms/step - loss: 0.2698 - accuracy: 0.9077 - val_loss: 0.4356 - val_accuracy: 0.8622 - lr: 0.0010
     Epoch 145/150
     28/28 [=
                                  =======1 - 29s 1s/step - loss: 0.2755 - accuracy: 0.9036 - val loss: 0.8728 - val accuracy: 0.7956 - lr: 0.0010
     Epoch 146/150
     28/28 [==
                               ========] - 27s 965ms/step - loss: 0.2725 - accuracy: 0.9058 - val_loss: 0.7223 - val_accuracy: 0.7822 - lr: 0.0010
     Epoch 147/150
     28/28 [=
                               ========] - 27s 976ms/step - loss: 0.2376 - accuracy: 0.9132 - val_loss: 0.5238 - val_accuracy: 0.8422 - lr: 0.0010
           148/150
     Epoch
                                   =======] - 27s 971ms/step - loss: 0.2505 - accuracy: 0.9101 - val loss: 0.5822 - val accuracy: 0.8356 - lr: 0.0010
     28/28 [==
     Enoch 149/150
                               ========] - 28s 986ms/step - loss: 0.2441 - accuracy: 0.9072 - val_loss: 0.6091 - val_accuracy: 0.8133 - lr: 0.0010
     28/28 [=====
Epoch 150/150
     28/28 [==========] - 27s 969ms/step - loss: 0.2500 - accuracy: 0.9147 - val loss: 0.6381 - val accuracy: 0.8511 - lr: 0.0010
import matplotlib.pyplot as plt
# Plot the results
acc = history.history['accuracy']
val acc = history.history['val accuracy']
loss = history.history['loss']
val_loss = history.history['val_loss']
epochs = range(len(acc))
plt.plot(epochs, acc, 'r', label='Training accuracy')
plt.plot(epochs, val_acc, 'b', label='Validation accuracy')
```

```
plt.title('Training and validation accuracy')
plt.legend(loc=0)
plt.figure()

plt.plot(epochs, loss, 'r', label='Training loss')
plt.plot(epochs, val_loss, 'b', label='Validation loss')
plt.title('Training and validation loss')
plt.legend(loc=0)
plt.figure()
plt.show()
```



Model Prediction

```
## CODE BLOCK FOR NON-SAFARI BROWSERS
## SAFARI USERS: PLEASE SKIP THIS BLOCK AND RUN THE NEXT ONE INSTEAD
import numpy as np
from google.colab import files
from keras.preprocessing import image
uploaded = files.upload()
for fn in uploaded.keys():
  # predicting images
  img = image.load_img(path, target_size=(150, 150))
  x = image.img_to_array(img)
  x = np.expand_dims(x, axis=0)
  images = np.vstack([x])
classes = model.predict(images, batch_size=10)
  print(fn)
  print(classes)
import numpy as np
from google.colab import files
from keras.preprocessing import image
from pathlib import Path
predictDir = [x[\emptyset] \ for \ x \ in \ os.walk('/content/drive/MyDrive/indo\_food\_datasets/test/')]
del predictDir[0]
predictDir.sort()
presentase = 0
index = 0
for folder in predictDir:
  print('Predict food: ', folder)
  for fn in Path(folder).glob('*.png'):
    # predicting images
    path = os.path.join(folder,fn)
img = image.load_img(path, target_size=(150, 150))
    x = image.img_to_array(img)
    x = np.expand_dims(x, axis=0)
    images = np.vstack([x])
classes = model.predict(images, batch_size=10)
print(fn)
    print(classes)
    print(classes[0,index])
    if classes[0,index] >= np.max(classes) :
      presentase += 1
  presentase = presentase/10
  print('Presentase untuk {} adalah {}'.format(folder, str(presentase)))
  index += 1
  presentase = 0
     /content/drive/MyDrive/indo_food_datasets/test/tahu petis/945.png [[0. 0. 0. 0. 0. 1. 0. 0. 0. 0.]]
```

```
/content/drive/MyDrive/indo_food_datasets/test/tahu petis/952.png [[0. 1. 0. 0. 0. 0. 0. 0. 0. 0.]]
      v.v
/content/drive/MyDrive/indo_food_datasets/test/tahu petis/964.png
[[0.0.0.0.0.0.0.0.0.0]
      /content/drive/MyDrive/indo_food_datasets/test/tahu petis/967.png [[0. 0. 0. 0. 0. 0. 0. 1. 0. 0.]]
      0.0
      /content/drive/MyDrive/indo_food_datasets/test/tahu petis/970.png
[[0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00 0.000000e+00
         1.000000e+00 0.000000e+00 0.000000e+00 4.780791e-37 0.000000e+00]]
      /content/drive/MyDrive/indo_food_datasets/test/tahu petis/986.png [[0. 0. 0. 1. 0. 0. 0. 0. 0. 0.]]
      Presentase untuk /content/drive/MyDrive/indo_food_datasets/test/tahu petis adalah 0.3
      Predict food: /content/drive/MyDrive/indo_food_datasets/test/tumpeng
      /content/drive/MyDrive/indo_food_datasets/test/tumpeng/436.png [[0.\ 0.\ 0.\ 0.\ 0.\ 0.\ 0.\ 0.\ 1.]]
      /content/drive/MyDrive/indo_food_datasets/test/tumpeng/687.png
[[0. 0. 0. 0. 0. 0. 0. 1. 0. 0.]]
      /content/drive/MyDrive/indo_food_datasets/test/tumpeng/39.png
[[0. 0. 0. 0. 0. 0. 0. 1. 0. 0.]]
      /content/drive/MyDrive/indo_food_datasets/test/tumpeng/513.png [[0. 0. 0. 0. 0. 0. 0. 0. 0. 1.]]
      /content/drive/MyDrive/indo_food_datasets/test/tumpeng/295.png
[[0. 0. 0. 0. 0. 0. 0. 0. 0. 1.]]
1.0
      /content/drive/MyDrive/indo_food_datasets/test/tumpeng/167.png
      [[0. 0. 0. 0. 0. 0. 0. 0. 0. 1.]]
1.0
      //content/drive/MyDrive/indo_food_datasets/test/tumpeng/165.png
[[0. 0. 0. 0. 0. 0. 0. 1. 0. 0.]]
      0.0 (/content/drive/MyDrive/indo_food_datasets/test/tumpeng/383.png [[0.0.0.0.0.0.0.0.0.0.]]
       /content/drive/MyDrive/indo food datasets/test/tumpeng/112.png
      [[0. 0. 0. 0. 0. 0. 0. 0. 0. 1.]]
      //content/drive/MyDrive/indo_food_datasets/test/tumpeng/596.png
[[0. 0. 0. 0. 0. 0. 0. 0. 0. 1.]]
      Presentase untuk /content/drive/MyDrive/indo_food_datasets/test/tumpeng adalah 0.6
saved model path = "./saved model/{}.h5".format(int(time.time()))
model.save(saved_model_path)
!tensorflowis converter --input format=keras {saved model path} ./saved model/is/
!zip -r datasetcovercrop_plateau.zip saved_model
         adding: saved_model/js/ (stored 0%)
         adding: saved_model/js/group1-shard1of3.bin (deflated 7%) adding: saved_model/js/model.json (deflated 82%) adding: saved_model/js/group1-shard2of3.bin (deflated 7%) adding: saved_model/js/group1-shard2of3.bin (deflated 7%)
         adding: saved_model/1654502220.h5 (deflated 27%)
model = tf.keras.models.load_model("/content/saved_model/1653899443.h5")
```

Finish

✓ 2s completed at 2:58 PM