

In [1]:

```
import tensorflow as tf
```

In [2]:

```
data_dir = "data/food-101/images/"
```

In [68]:

```
from tensorflow.keras import mixed_precision  
  
mixed_precision.set_global_policy('float32')
```

In [64]:

```
from tensorflow.keras.utils import image_dataset_from_directory  
  
all_images = image_dataset_from_directory(directory=data_dir,  
                                          label_mode="categorical",  
                                          batch_size=32,  
                                          image_size=(256, 256)  
                                          )
```

Found 101000 files belonging to 101 classes.

In [65]:

```
len(all_images.class_names)
```

Out[65]:

101

In [69]:

```
from tensorflow.keras import Sequential, Model, Input  
from tensorflow.keras.layers import Dense, GlobalAveragePooling2D  
from tensorflow.keras.applications import EfficientNetB0  
  
data_aug = Sequential([  
    tf.keras.layers.RandomRotation(0.2),  
    tf.keras.layers.RandomZoom(0.2),  
    tf.keras.layers.RandomHeight(0.2),  
    tf.keras.layers.RandomWidth(0.2),  
    # tf.keras.layers.Rescaling(1.0/255) # But not for the Efficient layer as its own Rescaling Layer  
, name="Data_Augmentation_Layer")  
  
base_model = EfficientNetB0(include_top=False)  
for i in base_model.layers[:-10]:  
    i.trainable = False  
  
ip = Input(shape=(256, 256, 3), name="Input_layer")  
x = data_aug(ip)  
x = base_model(x, training=False)  
x = GlobalAveragePooling2D(name="Pooling_layer")(x)  
op = Dense(101, activation="softmax", name="output_layer")(x)  
  
food_vision_model = Model(ip, op)
```

In [70]:

```
food_vision_model.summary()
```

Model: "model_6"

Layer (type)	Output Shape	Param #
Input_layer (InputLayer)	[(None, 256, 256, 3)]	0
Data_Augmentation_Layer (Sequential)	(None, 256, 256, 3)	0
efficientnetb0 (Functional)	(None, None, None, 1280)	4049571
Pooling_layer (GlobalAveragePooling2D)	(None, 1280)	0
output_layer (Dense)	(None, 101)	129381
Total params: 4,178,952		
Trainable params: 1,022,613		
Non-trainable params: 3,156,339		

In [71]:

```
food_vision_model.compile(loss="categorical_crossentropy",
                        optimizer=tf.keras.optimizers.Adam(learning_rate=0.0001),
                        metrics=["accuracy"])
```

In [72]:

```
# Creating a model callback Function
from tensorflow.keras.callbacks import ModelCheckpoint

checkpoint_dir = "models/food_vision_model/"

checkpoint_callback = ModelCheckpoint(filepath=checkpoint_dir,
                                     save_best_only=True,
                                     save_weights_only=False,
                                     save_freq="epoch",
                                     monitor='loss')
```

In [74]:

```
food_vision_model.fit(all_images,
                    epochs=30,
                    steps_per_epoch=len(all_images),
                    callbacks=[checkpoint_callback])
```

Epoch 1/30

3157/3157 [=====] - ETA: 0s - loss: 1.9760 - accuracy: 0.5078

WARNING:absl:Function `_wrapped_model` contains input name(s) Input_layer with unsupported characters which will be renamed to input_layer in the SavedModel.

WARNING:absl:Found untraced functions such as _jit_compiled_convolution_op, _jit_compiled_convolution_op, _jit_compiled_convolution_op, _jit_compiled_convolution_op while saving (showing 5 of 8). These functions will not be directly callable after loading.

INFO:tensorflow:Assets written to: models/food_vision_model/assets

INFO:tensorflow:Assets written to: models/food_vision_model/assets

3157/3157 [=====] - 1210s 382ms/step - loss: 1.9760 - accuracy: 0.5078

Epoch 2/30

3157/3157 [=====] - ETA: 0s - loss: 1.4166 - accuracy: 0.6315

WARNING:absl:Function `_wrapped_model` contains input name(s) Input_layer with unsupported characters which will be renamed to input_layer in the SavedModel.

WARNING:absl:Found untraced functions such as _jit_compiled_convolution_op, _jit_compiled_convolution_op, _jit_compiled_convolution_op, _jit_compiled_convolution_op while saving (showing 5 of 8).

In [76]:

```
food_vision_model.save("models/food_vision_model.h5")
```

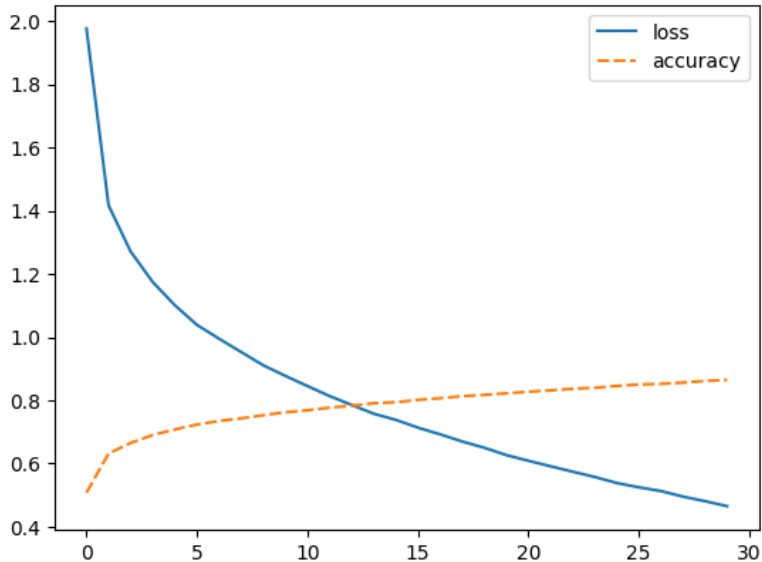
In [87]:

```
import pandas as pd
import seaborn as sns

df = pd.DataFrame(food_vision_model.history.history)
sns.lineplot(df)
```

Out[87]:

<AxesSubplot: >



In [88]:

```
test_dir = "data/food-101/test/"
```

In [90]:

```
test_images = image_dataset_from_directory(directory=test_dir,
                                          label_mode="categorical",
                                          batch_size=32,
                                          image_size=(256, 256)
                                          )
```

Found 25250 files belonging to 101 classes.

In [107]:

```
y_labels = []
for images, labels in test_images.unbatch(): # unbatch the test data and get images and labels
    y_labels.append(labels.numpy().argmax()) # append the index which has the largest value (labels are one-hot)
y_labels[:10] # check what they look like (unshuffled)
```

Out[107]:

[48, 55, 5, 11, 97, 84, 22, 72, 59, 49]

In [92]:

```
y_preds = food_vision_model.predict(test_images)
```

790/790 [=====] - 428s 543ms/step

In [94]:

```
y_classes = y_preds.argmax(axis=1)
```

In [95]:

```
y_classes
```

Out[95]:

array([51, 18, 21, ..., 36, 20, 73], dtype=int64)

In [129]:

```

import itertools
import matplotlib.pyplot as plt
import numpy as np
from sklearn.metrics import confusion_matrix

def make_confusion_matrix(y_true, y_pred, classes=None, figsize=(100, 100), text_size=15, norm=False, savefig=False):
    cm = confusion_matrix(y_true, y_pred)
    cm_norm = cm.astype("float") / cm.sum(axis=1)[:, np.newaxis] # normalize it
    n_classes = cm.shape[0] # find the number of classes we're dealing with

    fig, ax = plt.subplots(figsize=figsize)
    cax = ax.matshow(cm, cmap=plt.cm.Blues) # colors will represent how 'correct' a class is, darker == better
    fig.colorbar(cax)

    if classes:
        labels = classes
    else:
        labels = np.arange(cm.shape[0])

    # Label the axes
    ax.set(title="Confusion Matrix",
           xlabel="Predicted label",
           ylabel="True label",
           xticks=np.arange(n_classes),
           yticks=np.arange(n_classes),
           xticklabels=labels,
           yticklabels=labels)

    ax.xaxis.set_label_position("bottom")
    ax.xaxis.tick_bottom()

    plt.xticks(rotation=70, fontsize=text_size)
    plt.yticks(fontsize=text_size)

    threshold = (cm.max() + cm.min()) / 2.

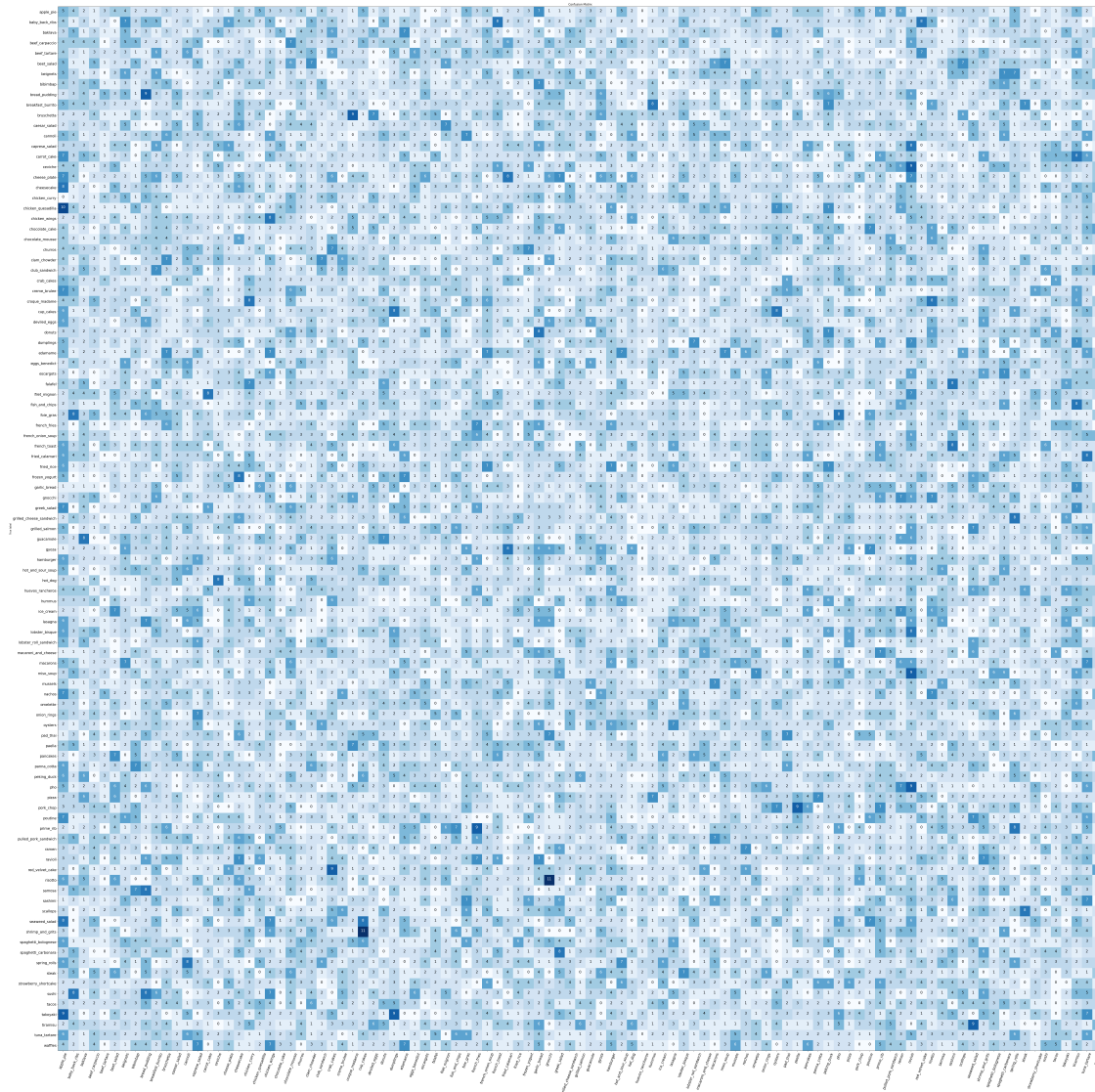
    for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
        if norm:
            plt.text(j, i, f"{cm[i, j]} ({cm_norm[i, j]*100:.1f}%)",
                     horizontalalignment="center",
                     color="white" if cm[i, j] > threshold else "black",
                     size=text_size)
        else:
            plt.text(j, i, f"{cm[i, j]}",
                     horizontalalignment="center",
                     color="white" if cm[i, j] > threshold else "black",
                     size=text_size)

    if savefig:
        fig.savefig("confusion_matrix.png")

```

In [130]:

```
make_confusion_matrix(y_labels, y_classes, test_images.class_names, savefig=True)
```



In [131]:

```
food_vision_model.evaluate(test_images)
```

790/790 [=====] - 83s 102ms/step - loss: 0.3803 - accuracy: 0.8827

Out[131]:

[0.3803199529647827, 0.8826534748077393]

In [139]:

```
def load_and_prep_image(filename, img_shape=256, scale=True):
    img = tf.io.read_file(filename)
    img = tf.io.decode_image(img)
    img = tf.image.resize(img, [img_shape, img_shape])
    if scale:
        return img/255.
    else:
        return img
```

In [175]:

```

class_names = test_images.class_names
import os
import random

plt.figure(figsize=(17, 10))

for i in range(3):
    class_name = random.choice(class_names)
    filename = random.choice(os.listdir(data_dir + "/" + class_name))
    filepath = data_dir + class_name + "/" + filename

    img = load_and_prep_image(filepath, scale=False)
    pred_prob = food_vision_model.predict(tf.expand_dims(img, axis=0))
    pred_class = class_names[pred_prob.argmax()]

    plt.subplot(1, 3, i+1)
    plt.imshow(img/255.)
    if class_name == pred_class:
        title_color = "g"
    else:
        title_color = "r"
    plt.title(f"actual: {class_name}, pred: {pred_class}, prob: {pred_prob.max():.2f}", c=title_color)
    plt.axis(False);

```

```

1/1 [=====] - 0s 29ms/step
1/1 [=====] - 0s 28ms/step
1/1 [=====] - 0s 25ms/step

```

actual: prime_rib, pred: prime_rib, prob: 0.99



actual: baklava, pred: baklava, prob: 0.85



actual: macarons, pred: macarons, prob: 1.00



In []: