

## DATA AUGMENTATION

```
from google.colab import drive
drive.mount('/content/gdrive')
```

Mounted at /content/gdrive

```
!pip install augmentor
```

```
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
Collecting augmentor
  Downloading Augmentor-0.2.12-py2.py3-none-any.whl (38 kB)
Requirement already satisfied: tqdm>=4.9.0 in /usr/local/lib/python3.9/dist-packages (from augmentor) (4.6)
Requirement already satisfied: Pillow>=5.2.0 in /usr/local/lib/python3.9/dist-packages (from augmentor) (8)
Requirement already satisfied: numpy>=1.11.0 in /usr/local/lib/python3.9/dist-packages (from augmentor) (1)
Installing collected packages: augmentor
Successfully installed augmentor-0.2.12
```

```
!pip install augmentor
# Importing necessary library
import Augmentor
# Passing the path of the image directory
p = Augmentor.Pipeline('/content/gdrive/MyDrive/Hidden hunger/Train_dataset')

# Defining augmentation parameters and generating 5 samples
p.zoom(probability = 0.5, min_factor = 0.8, max_factor = 1.5)
p.flip_top_bottom(probability=0.5)
p.sample(7000)
```

```
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
Requirement already satisfied: augmentor in /usr/local/lib/python3.9/dist-packages (0.2.12)
Requirement already satisfied: numpy>=1.11.0 in /usr/local/lib/python3.9/dist-packages (from augmentor) (1)
Requirement already satisfied: Pillow>=5.2.0 in /usr/local/lib/python3.9/dist-packages (from augmentor) (8)
Requirement already satisfied: tqdm>=4.9.0 in /usr/local/lib/python3.9/dist-packages (from augmentor) (4.6)
Initialised with 6600 image(s) found.
Output directory set to /content/gdrive/MyDrive/Hidden hunger/Train_dataset/output.Processing <PIL.Image.I
```

```
pip install split-folders
```

```
Looking in indexes: https://pypi.org/simple, https://us-python.pkg.dev/colab-wheels/public/simple/
Collecting split-folders
  Downloading split_folders-0.5.1-py3-none-any.whl (8.4 kB)
Installing collected packages: split-folders
Successfully installed split-folders-0.5.1
```

```
import splitfolders

input_folder = r'/content/gdrive/MyDrive/Hidden hunger/original dataset'
splitfolders.ratio(input_folder, output= r'/content/gdrive/MyDrive/Hidden hunger/split dataset_orig',
                    seed=42, ratio=(.8, .2),
                    group_prefix=None)
```

Copying files: 5827 files [02:17, 42.52 files/s]

```
import matplotlib.pyplot as plt
import numpy as np
import os
import PIL
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras.applications.inception_v3 import InceptionV3
```

```

from tensorflow.keras.applications.inception_v3 import preprocess_input
from tensorflow.keras import layers
from tensorflow.python.keras.layers import Dense, Flatten
from tensorflow.keras.models import Sequential
from tensorflow.keras.optimizers import Adam

train_path = '/content/gdrive/MyDrive/Hidden hunger/split dataset_orig/train'
test_path = '/content/gdrive/MyDrive/Hidden hunger/split dataset_orig/val'

from tensorflow.keras.layers import Input,Lambda,Dense,Flatten
from tensorflow.keras.models import Model
from tensorflow.keras.preprocessing import image

IMAGE_SIZE = [299,299]

# Use the Image Data Generator to import the images from the dataset
from tensorflow.keras.preprocessing.image import ImageDataGenerator

train_datagen = ImageDataGenerator(rescale = 1./255,
                                   shear_range = 0.2,
                                   zoom_range = 0.2,
                                   horizontal_flip = True
                                   )

test_datagen = ImageDataGenerator(rescale = 1./255,
                                   )

# Make sure you provide the same target size as initialied for the image size
train_set=train_datagen.flow_from_directory('/content/gdrive/MyDrive/Hidden hunger/split dataset_orig/train',tar
                                           batch_size=32,
                                           class_mode='categorical')

Found 4659 images belonging to 6 classes.

test_set = test_datagen.flow_from_directory('/content/gdrive/MyDrive/Hidden hunger/split dataset_orig/val',
                                           target_size = (299, 299),
                                           batch_size = 32,
                                           class_mode = 'categorical')

Found 1168 images belonging to 6 classes.

class_name = train_set.class_indices
print(class_name)

{'Iodine Deficiency': 0, 'Vitamin - B12 Deficiency': 1, 'Vitamin D deficiency': 2, 'Zinc Deficiency': 3,

```



```

inceptionv3_model = Sequential()
pretrained_model = InceptionV3(input_shape=(299,299,3), weights='imagenet', include_top=False,
                               pooling='avg',classes=6)
for layer in pretrained_model.layers:
    layer.trainable=False

inceptionv3_model.add(pretrained_model)
inceptionv3_model.add(Flatten())
inceptionv3_model.add(Dense(512, activation='relu'))
inceptionv3_model.add(Dense(6, activation='softmax'))

Downloading data from https://storage.googleapis.com/tensorflow/keras-applications/inception\_v3/inception\_87910968/87910968 [=====] - 5s 0us/step

```



```
inceptionv3_model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
inception_v3 (Functional)	(None, 2048)	21802784
flatten (Flatten)	(None, 2048)	0
dense (Dense)	(None, 512)	1049088
dense_1 (Dense)	(None, 6)	3078
Total params: 22,854,950		
Trainable params: 1,052,166		
Non-trainable params: 21,802,784		

```
inceptionv3_model.compile(optimizer=Adam(learning_rate=0.01), loss='categorical_crossentropy', metrics=['accuracy'])
```

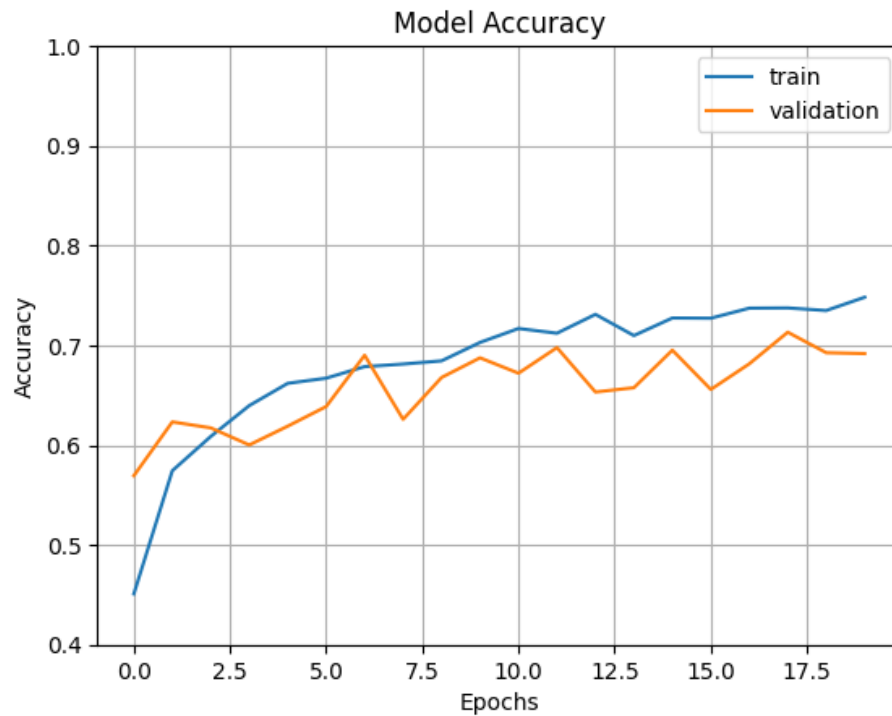
```
epochs=20
```

```
history = inceptionv3_model.fit(train_set, validation_data=test_set, epochs=epochs)
```

```
Epoch 1/20
146/146 [=====] - 150s 927ms/step - loss: 2.8875 - accuracy: 0.4510 - val_loss: 1
Epoch 2/20
146/146 [=====] - 130s 885ms/step - loss: 1.1381 - accuracy: 0.5744 - val_loss: 1
Epoch 3/20
146/146 [=====] - 131s 897ms/step - loss: 1.0548 - accuracy: 0.6087 - val_loss: 1
Epoch 4/20
146/146 [=====] - 130s 890ms/step - loss: 0.9788 - accuracy: 0.6396 - val_loss: 1
Epoch 5/20
146/146 [=====] - 131s 897ms/step - loss: 0.9026 - accuracy: 0.6619 - val_loss: 1
Epoch 6/20
146/146 [=====] - 132s 900ms/step - loss: 0.8741 - accuracy: 0.6671 - val_loss: 1
Epoch 7/20
146/146 [=====] - 127s 872ms/step - loss: 0.8702 - accuracy: 0.6787 - val_loss: 0
Epoch 8/20
146/146 [=====] - 129s 881ms/step - loss: 0.8673 - accuracy: 0.6813 - val_loss: 1
Epoch 9/20
146/146 [=====] - 127s 871ms/step - loss: 0.8501 - accuracy: 0.6845 - val_loss: 0
Epoch 10/20
146/146 [=====] - 126s 864ms/step - loss: 0.8136 - accuracy: 0.7029 - val_loss: 0
Epoch 11/20
146/146 [=====] - 125s 859ms/step - loss: 0.7748 - accuracy: 0.7169 - val_loss: 0
Epoch 12/20
146/146 [=====] - 126s 863ms/step - loss: 0.7732 - accuracy: 0.7122 - val_loss: 0
Epoch 13/20
146/146 [=====] - 126s 865ms/step - loss: 0.7521 - accuracy: 0.7311 - val_loss: 0
Epoch 14/20
146/146 [=====] - 125s 859ms/step - loss: 0.7815 - accuracy: 0.7098 - val_loss: 0
Epoch 15/20
146/146 [=====] - 125s 858ms/step - loss: 0.7348 - accuracy: 0.7274 - val_loss: 0
Epoch 16/20
146/146 [=====] - 125s 856ms/step - loss: 0.7490 - accuracy: 0.7272 - val_loss: 0
Epoch 17/20
146/146 [=====] - 125s 859ms/step - loss: 0.7162 - accuracy: 0.7373 - val_loss: 0
Epoch 18/20
146/146 [=====] - 125s 858ms/step - loss: 0.7228 - accuracy: 0.7375 - val_loss: 0
Epoch 19/20
146/146 [=====] - 127s 869ms/step - loss: 0.7247 - accuracy: 0.7349 - val_loss: 0
Epoch 20/20
146/146 [=====] - 126s 864ms/step - loss: 0.6807 - accuracy: 0.7482 - val_loss: 0
```

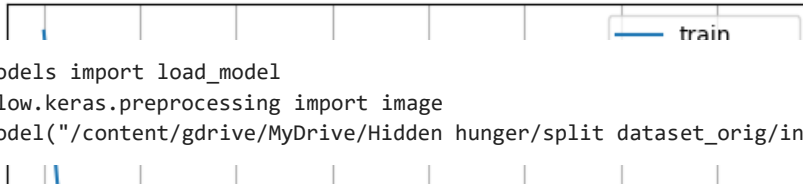
```
inceptionv3_model.save("/content/gdrive/MyDrive/Hidden hunger/split dataset_orig/inceptionv3_orig.h5")
```

```
fig1 = plt.gcf()
plt.plot(history.history['accuracy'])
plt.plot(history.history['val_accuracy'])
plt.axis(ymin=0.4,ymax=1)
plt.grid()
plt.title('Model Accuracy')
plt.ylabel('Accuracy')
plt.xlabel('Epochs')
plt.legend(['train', 'validation'])
plt.show()
```



```
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
plt.grid()
plt.title('Model Loss')
plt.ylabel('Loss')
plt.xlabel('Epochs')
plt.legend(['train', 'validation'])
plt.show()
```

## Model Loss

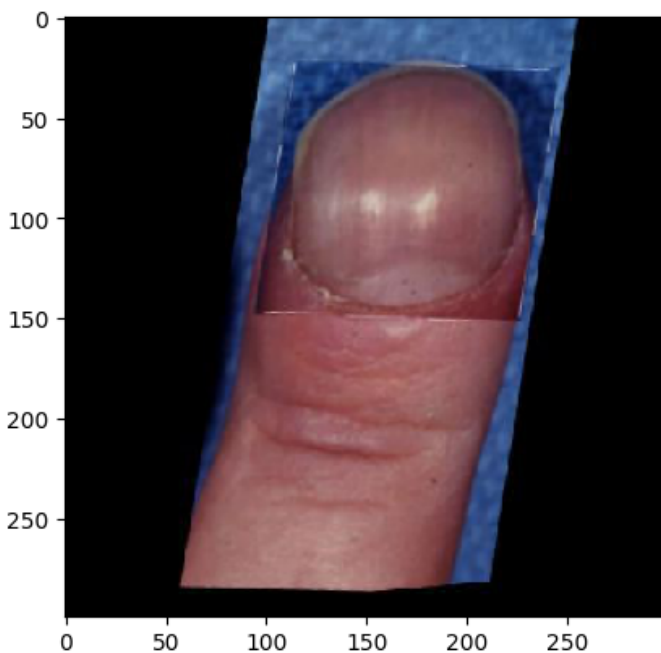


```
from keras.models import load_model
from tensorflow.keras.preprocessing import image
model=load_model("/content/gdrive/MyDrive/Hidden hunger/split dataset_orig/inceptionv3_orig.h5")
```

```
import numpy as np
def predictImage(filename,model):
    img1=image.load_img(filename,target_size=(299,299))
    plt.imshow(img1)
    Y=image.img_to_array(img1)
    X=np.expand_dims(Y,axis=0)
    pred=model.predict(X/255)
    print(pred)
    pred = np.array(pred)
    val = np.argmax(pred)
    print(val)
    if (val==0).all():
        plt.xlabel("Iodine Deficiency",fontsize=50)
    elif (val==1).all():
        plt.xlabel("Iron Deficiency",fontsize=50)
    elif (val==2).all():
        plt.xlabel("Vitamin - B12 Deficiency",fontsize=50)
    elif (val==3).all():
        plt.xlabel("Vitamin D - Deficiency",fontsize=25)
    elif (val==4).all():
        plt.xlabel("Zinc Deficiency",fontsize=50)
    elif (val==5).all():
        plt.xlabel("healthy",fontsize=25)
```

```
predictImage("/content/gdrive/MyDrive/Hidden hunger/val/Iodine Deficiency/Iodine Deficiency_original_Screen-Shot
```

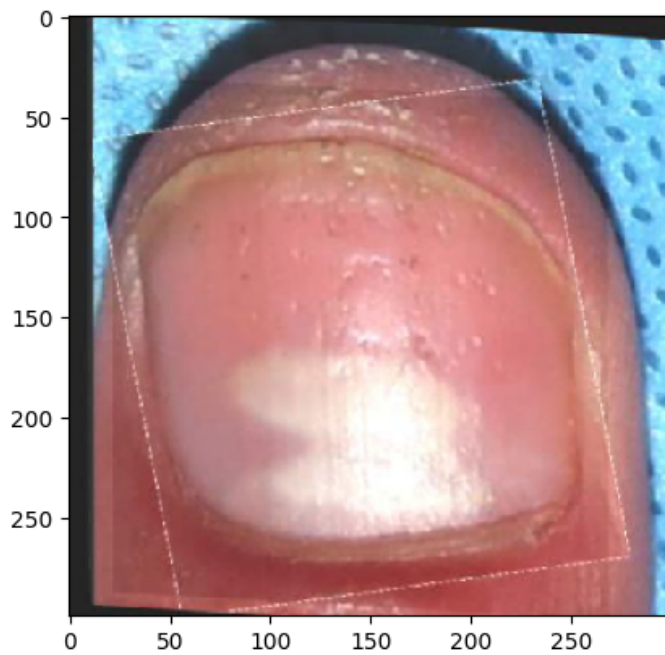
```
1/1 [=====] - 3s 3s/step
[[1.4412683e-01 1.7236900e-02 7.4900180e-01 1.7844034e-02 7.1769923e-02
 2.0544147e-05]]
2
```



in - B12 Def

```
predictImage("/content/gdrive/MyDrive/Hidden hunger/val/Iron Deficiency/Iron Deficiency_original_112_JPG.rf.4a61
```

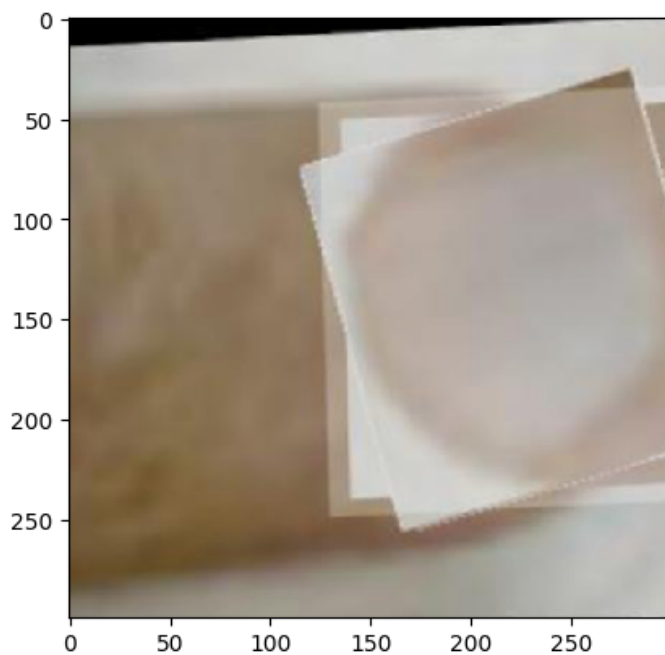
```
1/1 [=====] - 0s 47ms/step
[[1.2578721e-01 7.1040863e-01 2.5478872e-02 8.6294105e-03 1.2921669e-01
  4.7909268e-04]]
1
```



# ron Deficiency

```
predictImage("/content/gdrive/MyDrive/Hidden hunger/val/Vitamin - B12 Deficiency/Vitamin - B12 Deficiency_orig
```

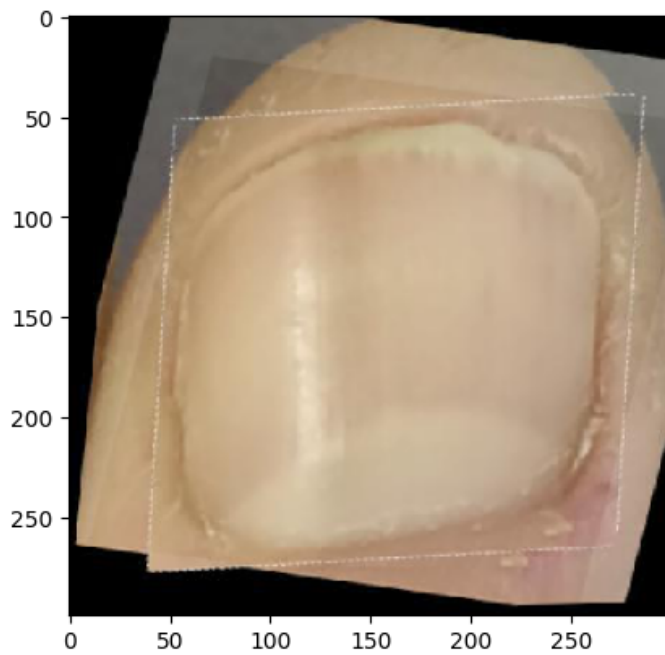
```
1/1 [=====] - 0s 45ms/step
[[3.1044530e-02 3.8101595e-02 9.2412591e-01 1.9836647e-04 8.3671941e-04
  5.6927935e-03]]
2
```



# in - B12 Def

```
predictImage("/content/gdrive/MyDrive/Hidden hunger/val/Vitamin D - Deficiency/Vitamin D - Deficiency_original_S
```

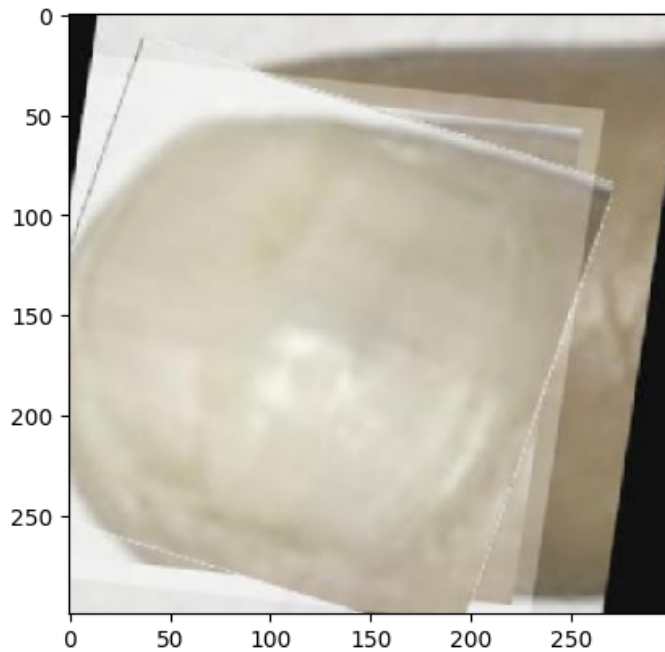
```
1/1 [=====] - 0s 28ms/step
[[0.00152076 0.29042125 0.03032159 0.5030235 0.1166429 0.05807005]]
3
```



# Vitamin D - Deficiency

```
predictImage("/content/gdrive/MyDrive/Hidden hunger/val/Zinc Deficiency/Zinc Deficiency_original_Screen-Shot-202
```

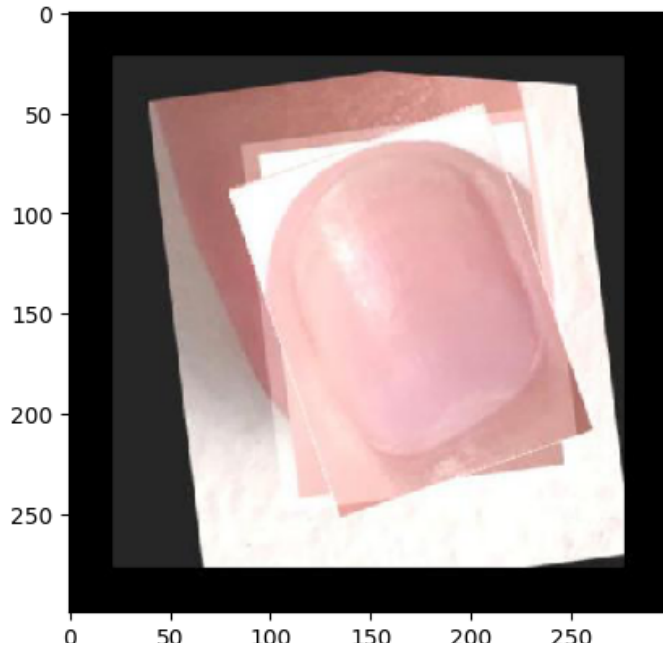
```
1/1 [=====] - 0s 27ms/step
[[0.4563299 0.04907864 0.33956146 0.0005035 0.15108983 0.00343669]]
0
```



# dine Deficier

```
predictImage("/content/gdrive/MyDrive/Hidden hunger/val/healthy/healthy_original_Screen-Shot-2021-11-15-at-12-52
```

```
1/1 [=====] - 0s 97ms/step
[[0.00148466 0.02746078 0.01187978 0.00133189 0.2960549 0.6617879 ]]
5
```



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

```
saved_model = load_model("/content/gdrive/MyDrive/Hidden hunger/split dataset_orig/inceptionv3_orig.h5")
all_y_pred = []
all_y_true = []
```

```
for i in range(len(test_set)):
    x, y = test_set[i]
    y_pred = saved_model.predict(x)

    all_y_pred.append(y_pred)
    all_y_true.append(y)
```

```
all_y_pred = np.concatenate(all_y_pred, axis=0)
all_y_true = np.concatenate(all_y_true, axis=0)
```

```
1/1 [=====] - 2s 2s/step
1/1 [=====] - 0s 62ms/step
1/1 [=====] - 0s 47ms/step
1/1 [=====] - 0s 46ms/step
1/1 [=====] - 0s 59ms/step
1/1 [=====] - 0s 59ms/step
1/1 [=====] - 0s 60ms/step
1/1 [=====] - 0s 58ms/step
1/1 [=====] - 0s 46ms/step
1/1 [=====] - 0s 57ms/step
1/1 [=====] - 0s 65ms/step
1/1 [=====] - 0s 57ms/step
1/1 [=====] - 0s 45ms/step
1/1 [=====] - 0s 45ms/step
1/1 [=====] - 0s 56ms/step
1/1 [=====] - 0s 44ms/step
1/1 [=====] - 0s 45ms/step
1/1 [=====] - 0s 55ms/step
1/1 [=====] - 0s 43ms/step
1/1 [=====] - 0s 42ms/step
1/1 [=====] - 0s 58ms/step
1/1 [=====] - 0s 56ms/step
1/1 [=====] - 0s 44ms/step
1/1 [=====] - 0s 44ms/step
1/1 [=====] - 0s 45ms/step
```



```
1/1 [=====] - 0s 44ms/step
1/1 [=====] - 0s 55ms/step
1/1 [=====] - 0s 83ms/step
1/1 [=====] - 0s 64ms/step
1/1 [=====] - 0s 79ms/step
1/1 [=====] - 0s 72ms/step
1/1 [=====] - 0s 74ms/step
1/1 [=====] - 0s 61ms/step
1/1 [=====] - 0s 75ms/step
1/1 [=====] - 0s 73ms/step
1/1 [=====] - 0s 44ms/step
1/1 [=====] - 1s 1s/step
```

```
from sklearn.metrics import confusion_matrix
import matplotlib.pyplot as plt
import seaborn as sns
import numpy as np

# compute confusion matrix
cm = confusion_matrix(all_y_true.argmax(axis=1), all_y_pred.argmax(axis=1))

# create heatmap from confusion matrix
fig, ax = plt.subplots(figsize=(6,6))
sns.heatmap(cm, annot=True, cmap="Greens", fmt="d", xticklabels=train_set.class_indices.keys(),
            yticklabels=train_set.class_indices.keys(), ax=ax)

# set axis labels and title
ax.set_xlabel('Predicted labels')
ax.set_ylabel('True labels')
ax.set_title('Confusion Matrix')
```

```
Text(0.5, 1.0, 'Confusion Matrix')
```

```
from sklearn.metrics import classification_report
```

```
# Get the predicted class labels
```

```
y_pred = np.argmax(all_y_pred, axis=1)
```

```
# Get the true class labels
```

```
y_true = np.argmax(all_y_true, axis=1)
```

```
# Compute classification report
```

```
report = classification_report(y_true, y_pred, target_names=train_set.class_indices.keys())
```

```
# Print classification report
```

```
print(report)
```

	precision	recall	f1-score	support
Iodine Deficiency	0.66	0.73	0.69	157
Vitamin - B12 Deficiency	0.59	0.49	0.54	192
Vitamin D deficiency	0.64	0.77	0.70	161
Zinc Deficiency	1.00	0.15	0.27	91
healthy	0.70	0.75	0.72	148
iron deficiency	0.75	0.84	0.79	419
accuracy			0.69	1168
macro avg	0.72	0.62	0.62	1168
weighted avg	0.71	0.69	0.67	1168

```
precision recall f1-score support
```

```
import numpy as np
```

```
from tensorflow.keras.models import load_model
```

```
from tensorflow.keras.preprocessing.image import ImageDataGenerator
```

```
# Load the saved model
```

```
model = load_model('/content/gdrive/MyDrive/Hidden hunger/split dataset_orig/inceptionv3_orig.h5')
```

```
# Create an image data generator with normalization
```

```
test_datagen = ImageDataGenerator(rescale=1./255)
```

```
# Load the test data
```

```
test_data = test_datagen.flow_from_directory(test_path, target_size=(299, 299), batch_size=32, shuffle=False)
```

```
# Use the predict method to obtain predictions on the test data
```

```
y_pred = model.predict(test_data)
```

```
# Get the predicted class labels
```

```
y_pred_classes = np.argmax(y_pred, axis=1)
```

```
# Get the true class labels
```

```
y_true = test_data.classes
```

```
# Compute the test accuracy
```

```
test_acc = np.mean(y_pred_classes == y_true)
```

```
# Use the predict method to obtain predictions on the training data
```

```
y_pred_train = model.predict(train_set)
```

```
# Get the predicted class labels
```

```
y_pred_classes_train = np.argmax(y_pred_train, axis=1)
```

```
# Get the true class labels
```

```
y_true_train = train_set.classes
```

```
# Compute the train accuracy
```

```
train_acc = np.mean(y_pred_classes_train == y_true_train)
```

```
# Print the train and test accuracy
```

```
print('Train accuracy:', train_acc)
print('Test accuracy:', test_acc)
```

```
Found 1168 images belonging to 6 classes.
37/37 [=====] - 9s 206ms/step
146/146 [=====] - 119s 816ms/step
Train accuracy: 0.22622880446447735
Test accuracy: 0.6917808219178082
```

---

✓ 1m 57s completed at 12:25 AM

