```
import tensorflow as tf
from tensorflow.keras import layers, models, Input
from tensorflow.keras.preprocessing.image import ImageDataGenerator
import matplotlib.pyplot as plt
import pickle
from tensorflow.keras import mixed_precision
# Mount Google Drive
from google.colab import drive
drive.mount('/content/drive')
# Dataset Path
DATA_DIR = "/content/drive/MyDrive/Healthy and unhealthy "
#set up data
IMG_HEIGHT = 192
IMG_WIDTH = 192
BATCH SIZE = 64
train_gen = ImageDataGenerator(
    rescale=1./255,
    validation_split=0.2,
    horizontal_flip=True,
    rotation_range=20,
    zoom_range=0.2
train_data = train_gen.flow_from_directory(
    DATA DIR.
    target_size=(IMG_HEIGHT, IMG_WIDTH),
    batch_size=BATCH_SIZE,
    class_mode='categorical',
    subset='training
)
val_data = train_gen.flow_from_directory(
    DATA_DIR,
    target_size=(IMG_HEIGHT, IMG_WIDTH),
    batch_size=BATCH_SIZE,
    class_mode='categorical',
    subset='validation'
mixed_precision.set_global_policy('mixed_float16')
print("Mixed Precision Enabled ")
    Mounted at /content/drive
     Found 11200 images belonging to 2 classes.
     Found 2800 images belonging to 2 classes.
    Mixed Precision Enabled
import tensorflow as tf
from tensorflow.keras import layers, models, Input
from tensorflow.keras.callbacks import EarlyStopping, ReduceLROnPlateau
#build binary classifier
base_model = tf.keras.applications.EfficientNetB4(
    include_top=False,
    input_shape=(IMG_HEIGHT, IMG_WIDTH, 3),
    weights='imagenet'
base_model.trainable = False # Freeze base for Phase 1
inputs = Input(shape=(IMG_HEIGHT, IMG_WIDTH, 3))
x = base_model(inputs, training=False)
x = layers.GlobalAveragePooling2D()(x)
x = layers.Dropout(0.5)(x)
outputs = layers.Dense(2, activation='softmax')(x) # binary classification
model = models.Model(inputs, outputs)
model.compile(optimizer='adam', loss='categorical_crossentropy', metrics=['accuracy'])
#Add EarlyStopping and ReduceLROnPlateau
earlystop = EarlyStopping(patience=3, restore_best_weights=True)
reduce_lr = ReduceLROnPlateau(factor=0.5, patience=2, verbose=1)
   Downloading data from <a href="https://storage.googleapis.com/keras-applications/efficientnetb4">https://storage.googleapis.com/keras-applications/efficientnetb4</a> notop.h5
     71686520/71686520
                                             - 4s Ous/step
```

```
30/04/2025, 16:07
```

```
#phase 1 train and save
history_phase1 = model.fit(
    train_data,
    validation_data=val_data,
    epochs=20,
    callbacks=[earlystop, reduce_lr]
)
model.save("health_classifier_phase1.keras")
print("Phase 1 model saved as health_classifier_phase1.keras")
```

/usr/local/lib/python3.11/dist-packages/keras/src/trainers/data_adapters/py_dataset_adapter.py:121: UserWarning: Your `P self._warn_if_super_not_called() Epoch 1/20 175/175 - 10575s 60s/step - accuracy: 0.5103 - loss: 0.7078 - val_accuracy: 0.5004 - val_loss: 0.6928 Epoch 2/20 175/175 **- 160s** 917ms/step - accuracy: 0.5136 - loss: 0.7019 - val_accuracy: 0.5075 - val_loss: 0.6923 Epoch 3/20 175/175 **- 160s** 914ms/step - accuracy: 0.5125 - loss: 0.6999 - val_accuracy: 0.5000 - val_loss: 0.6960 Epoch 4/20 - **159s** 910ms/step - accuracy: 0.5073 - loss: 0.7000 - val_accuracy: 0.5561 - val_loss: 0.6900 175/175 Epoch 5/20 175/175 - **159s** 909ms/step - accuracy: 0.5170 - loss: 0.6979 - val_accuracy: 0.5107 - val_loss: 0.6915 Epoch 6/20 **0s** 733ms/step - accuracy: 0.5082 - loss: 0.6999 175/175 Epoch 6: ReduceLROnPlateau reducing learning rate to 0.0005000000237487257. 175/175 161s 918ms/step - accuracy: 0.5083 - loss: 0.6999 - val_accuracy: 0.5411 - val_loss: 0.6900 Epoch 7/20 175/175 - **158s** 904ms/step – accuracy: 0.5209 – loss: 0.6964 – val_accuracy: 0.5082 – val_loss: 0.6911 Epoch 8/20 175/175 - **158s** 904ms/step – accuracy: 0.5183 – loss: 0.6973 – val_accuracy: 0.5382 – val_loss: 0.6890 Epoch 9/20 175/175 - 159s 911ms/step - accuracy: 0.5051 - loss: 0.6966 - val_accuracy: 0.5404 - val_loss: 0.6897 Epoch 10/20 175/175 - **0s** 724ms/step - accuracy: 0.5152 - loss: 0.6952 Epoch 10: ReduceLROnPlateau reducing learning rate to 0.0002500000118743628. 159s 908ms/step - accuracy: 0.5152 - loss: 0.6952 - val_accuracy: 0.5036 - val_loss: 0.6907 175/175 Epoch 11/20 175/175 - **159s** 910ms/step - accuracy: 0.5159 - loss: 0.6956 - val_accuracy: 0.5507 - val_loss: 0.6886 Epoch 12/20 175/175 **- 159s** 911ms/step – accuracy: 0.5125 – loss: 0.6949 – val_accuracy: 0.5111 – val_loss: 0.6906 Epoch 13/20 175/175 - **159s** 908ms/step – accuracy: 0.5222 – loss: 0.6944 – val_accuracy: 0.5554 – val_loss: 0.6885 Epoch 14/20 175/175 - **159s** 908ms/step – accuracy: 0.5250 – loss: 0.6920 – val_accuracy: 0.5507 – val_loss: 0.6888 Epoch 15/20 • **0s** 728ms/step - accuracy: **0.**5147 - loss: **0.**6943 175/175 Epoch 15: ReduceLROnPlateau reducing learning rate to 0.0001250000059371814. 175/175 160s 915ms/step - accuracy: 0.5147 - loss: 0.6943 - val_accuracy: 0.5271 - val_loss: 0.6896 Epoch 16/20 175/175 160s 912ms/step - accuracy: 0.5227 - loss: 0.6909 - val_accuracy: 0.5586 - val_loss: 0.6888 Phase 1 model saved as health_classifier_phase1.keras

```
from google.colab import files
model.save('/content/drive/MyDrive/health_classifier_phase1.keras')
print("Phase 1 model saved!")
files.download('/content/drive/MyDrive/health_classifier_phase1.keras')
```

→ Phase 1 model saved!

```
# Phase 2 fine-tuning
```

```
# Unfreeze base model for fine-tuning
base_model.trainable = True

model.compile(
    optimizer=tf.keras.optimizers.Adam(1e-5),
    loss='categorical_crossentropy',
    metrics=['accuracy']
)

history_phase2 = model.fit(
    train_data,
    validation_data=val_data,
    epochs=10,
    callbacks=[earlystop, reduce_lr]
)

#save model
model.save("health_classifier_phase2_finetuned.keras")
```

print("Phase 2 model saved locally as health_classifier_phase2_finetuned.keras")

```
→ Epoch 1/10
    175/175
                                – 369s 1s/step – accuracy: 0.5968 – loss: 0.6800 – val_accuracy: 0.5004 – val_loss: 0.8658 –
    Epoch 2/10
                                - 162s 924ms/step - accuracy: 0.7270 - loss: 0.5415 - val_accuracy: 0.6493 - val_loss: 0.6220
    175/175
    Epoch 3/10
    175/175 -
                                - 162s 928ms/step - accuracy: 0.7819 - loss: 0.4610 - val_accuracy: 0.8150 - val_loss: 0.4264
    Epoch 4/10
    175/175 -
                                - 165s 940ms/step - accuracy: 0.8170 - loss: 0.4003 - val_accuracy: 0.8521 - val_loss: 0.3433
    Epoch 5/10
    175/175
                               – 160s 914ms/step – accuracy: 0.8399 – loss: 0.3679 – val_accuracy: 0.8564 – val_loss: 0.3364
    Epoch 6/10
    175/175
                                - 161s 921ms/step - accuracy: 0.8612 - loss: 0.3293 - val_accuracy: 0.8239 - val_loss: 0.4100
    Epoch 7/10
                               – 163s 932ms/step – accuracy: 0.8550 – loss: 0.3330 – val_accuracy: 0.8757 – val_loss: 0.2939
    175/175 -
    Epoch 8/10
    175/175 -
                                - 165s 943ms/step - accuracy: 0.8616 - loss: 0.3110 - val_accuracy: 0.8918 - val_loss: 0.2751
    Epoch 9/10
    175/175
                                - 161s 922ms/step - accuracy: 0.8767 - loss: 0.2909 - val_accuracy: 0.8961 - val_loss: 0.2650
    Epoch 10/10
    175/175
                                · 166s 949ms/step – accuracy: 0.8813 – loss: 0.2789 – val_accuracy: 0.8936 – val_loss: 0.2769
    Phase 2 model saved locally as health_classifier_phase2_finetuned.keras
```

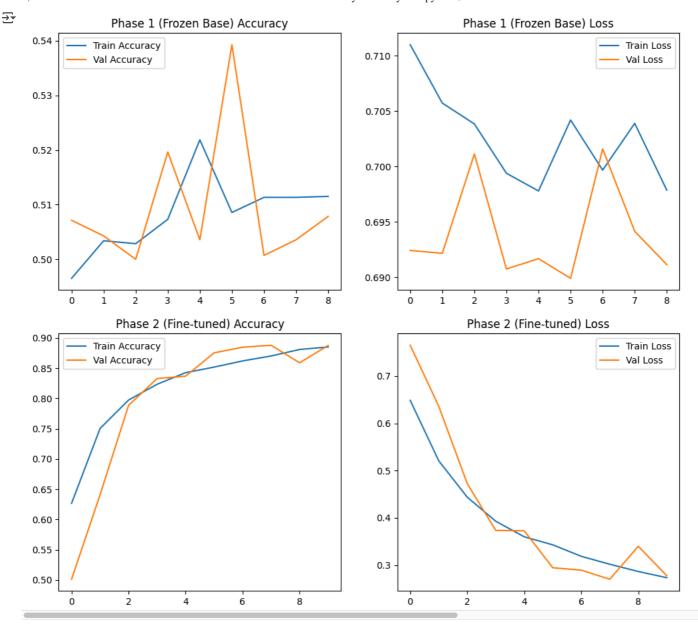
```
model.save('/content/drive/MyDrive/health_classifier_phase2.keras')
print("Phase 2 model saved!")
```

files.download('/content/drive/MyDrive/health_classifier_phase2.keras')

→ Phase 1 model saved!

Visualise accuracy and loss

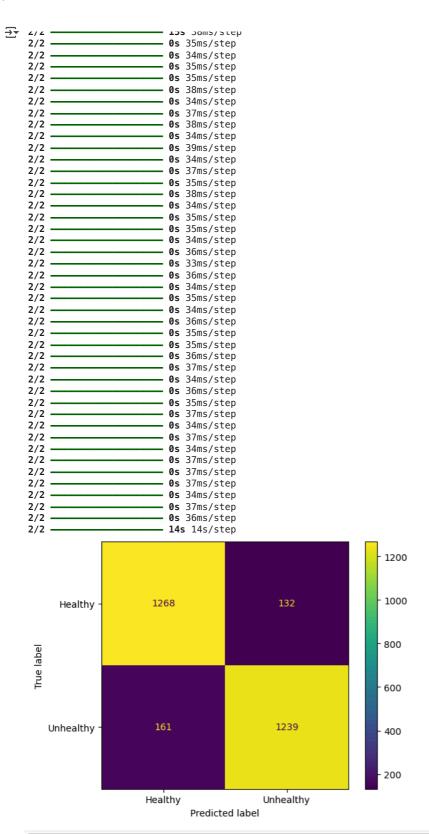
```
import matplotlib.pyplot as plt
def plot_training(history, title=""):
    plt.figure(figsize=(12,5))
    # Accuracy
    plt.subplot(1,2,1)
    plt.plot(history.history['accuracy'], label='Train Accuracy')
    plt.plot(history.history['val_accuracy'], label='Val Accuracy')
    plt.title(f"{title} Accuracy")
    plt.legend()
    # Loss
    plt.subplot(1,2,2)
    plt.plot(history.history['loss'], label='Train Loss')
    plt.plot(history.history['val_loss'], label='Val Loss')
    plt.title(f"{title} Loss")
    plt.legend()
    plt.show()
plot_training(history_phase1, "Phase 1 (Frozen Base)")
plot_training(history_phase2, "Phase 2 (Fine-tuned)")
```



model.save("final_health_classifier.keras")
print("Final model saved!")

WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save_model(model)`. This file Final model saved!

```
#Evaluate model:
oss, acc = model.evaluate(val_data)
print(f"Final Validation Accuracy: {acc*100:.2f}%")
    44/44
                               - 29s 657ms/step - accuracy: 0.8832 - loss: 0.2725
    Final Validation Accuracy: 88.61%
#confusion matrix
from \ sklearn.metrics \ import \ confusion\_matrix, \ ConfusionMatrixDisplay
import numpy as np
y_true = []
y_pred = []
for x_batch, y_batch in val_data:
    y_true.extend(np.argmax(y_batch, axis=1))
    preds = model.predict(x_batch)
    y_pred.extend(np.argmax(preds, axis=1))
    if len(y_true) >= val_data.samples:
        break
cm = confusion_matrix(y_true, y_pred)
disp = ConfusionMatrixDisplay(confusion_matrix=cm, display_labels=["Healthy", "Unhealthy"])
disp.plot()
```



#classification report

from sklearn.metrics import classification_report

 $\verb|print(classification_report(y_true, y_pred, target_names=["Healthy", "Unhealthy"])||$

_		precision	recall	f1-score	support
	Healthy	0.89	0.91	0.90	1400
	Unhealthy	0.90	0.89	0.89	1400
	accuracy			0.90	2800
	macro avg	0.90	0.90	0.90	2800
	weighted avg	0.90	0.90	0.90	2800