FINE TUNING THE PRETRAINED MODEL

```
import numpy as np
import pandas as pd
import tensorflow as tf
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.models import load model
from tensorflow.keras.applications import InceptionV3
from tensorflow.keras.layers import Dense, GlobalAveragePooling2D, BatchNormalization
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.callbacks import EarlyStopping, ModelCheckpoint
from sklearn.utils.class weight import compute class weight
# Step 1: Load the pre-trained model
model = load model('/content/drive/MyDrive/diabetic retinopathy binary model.h5')
    WARNING:absl:Compiled the loaded model, but the compiled metrics have yet to be built. `model.compile metrics` will be empty until you t
# Unfreeze the top layers of the pre-trained model for fine-tuning
# Assuming the first 249 layers belong to InceptionV3 and should be kept frozen
for layer in model.layers[:249]:
    layer.trainable = False # Freeze the first 249 layers
 #Unfreeze the remaining layers for fine-tuning
for layer in model.layers[249:]:
    layer.trainable = True # Unfreeze the top layers for fine-tuning
model.summary()
```

→ Model: "functional"

Layer (type)	Output Shape	Param #	Connected to
<pre>input_layer (InputLayer)</pre>	(None, 299, 299, 3)	0	-
conv2d (Conv2D)	(None, 149, 149, 32)	864	input_layer[0][0]
batch_normalization (BatchNormalization)	(None, 149, 149, 32)	96	conv2d[0][0]
activation (Activation)	(None, 149, 149, 32)	0	batch_normalization[0
conv2d_1 (Conv2D)	(None, 147, 147, 32)	9,216	activation[0][0]
batch_normalization_1 (BatchNormalization)	(None, 147, 147, 32)	96	conv2d_1[0][0]
activation_1 (Activation)	(None, 147, 147, 32)	0	batch_normalization_1
conv2d_2 (Conv2D)	(None, 147, 147, 64)	18,432	activation_1[0][0]
batch_normalization_2 (BatchNormalization)	(None, 147, 147, 64)	192	conv2d_2[0][0]
activation_2 (Activation)	(None, 147, 147, 64)	0	batch_normalization_2
max_pooling2d (MaxPooling2D)	(None, 73, 73, 64)	0	activation_2[0][0]
conv2d_3 (Conv2D)	(None, 73, 73, 80)	5,120	max_pooling2d[0][0]
batch_normalization_3 (BatchNormalization)	(None, 73, 73, 80)	240	conv2d_3[0][0]
activation_3 (Activation)	(None, 73, 73, 80)	0	batch_normalization_3
conv2d_4 (Conv2D)	(None, 71, 71, 192)	138,240	activation_3[0][0]
batch_normalization_4 (BatchNormalization)	(None, 71, 71, 192)	576	conv2d_4[0][0]
activation_4 (Activation)	(None, 71, 71, 192)	0	batch_normalization_4
<pre>max_pooling2d_1 (MaxPooling2D)</pre>	(None, 35, 35, 192)	0	activation_4[0][0]
conv2d_8 (Conv2D)	(None, 35, 35, 64)	12,288	max_pooling2d_1[0][0]
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batch_normalization_8 (BatchNormalization)	(None, 35, 35, 64)	192 	conv2d_8[0][0]
activation_8 (Activation)	(None, 35, 35, 64)	0	batch_normalization_8
conv2d_6 (Conv2D)	(None, 35, 35, 48)	9,216	max_pooling2d_1[0][0]
conv2d_9 (Conv2D)	(None, 35, 35, 96)	55,296	activation_8[0][0]
batch_normalization_6 (BatchNormalization)	(None, 35, 35, 48)	144	conv2d_6[0][0]
batch_normalization_9 (BatchNormalization)	(None, 35, 35, 96)	288	conv2d_9[0][0]
activation_6 (Activation)	(None, 35, 35, 48)	0	batch_normalization_6
activation_9 (Activation)	(None, 35, 35, 96)	0	batch_normalization_9
average_pooling2d (AveragePooling2D)	(None, 35, 35, 192)	0	max_pooling2d_1[0][0]
conv2d_5 (Conv2D)	(None, 35, 35, 64)	12,288	max_pooling2d_1[0][0]
conv2d_7 (Conv2D)	(None, 35, 35, 64)	76,800	activation_6[0][0]
conv2d_10 (Conv2D)	(None, 35, 35, 96)	82,944	activation_9[0][0]
conv2d_11 (Conv2D)	(None, 35, 35, 32)	6,144	average_pooling2d[0][…
batch_normalization_5 (BatchNormalization)	(None, 35, 35, 64)	192	conv2d_5[0][0]
batch_normalization_7 (BatchNormalization)	(None, 35, 35, 64)	192	conv2d_7[0][0]
batch_normalization_10 (BatchNormalization)	(None, 35, 35, 96)	288	conv2d_10[0][0]
batch_normalization_11 (BatchNormalization)	(None, 35, 35, 32)	96	conv2d_11[0][0]
activation_5 (Activation)	(None, 35, 35, 64)	0	batch_normalization_5
activation_7 (Activation)	(None, 35, 35, 64)	0	batch_normalization_7
activation_10 (Activation)	(None, 35, 35, 96)	0	batch_normalization_1
activation 11	(None 35 35 32)	а	hatch normalization 1

(Activation)	(10110, 55, 55, 52)	<u> </u>	500001_101.110122001_1
mixed0 (Concatenate)	(None, 35, 35, 256)	0	activation_5[0][0], activation_7[0][0], activation_10[0][0], activation_11[0][0]
conv2d_15 (Conv2D)	(None, 35, 35, 64)	16,384	mixed0[0][0]
batch_normalization_15 (BatchNormalization)	(None, 35, 35, 64)	192	conv2d_15[0][0]
activation_15 (Activation)	(None, 35, 35, 64)	0	batch_normalization_1
conv2d_13 (Conv2D)	(None, 35, 35, 48)	12,288	mixed0[0][0]
conv2d_16 (Conv2D)	(None, 35, 35, 96)	55,296	activation_15[0][0]
batch_normalization_13 (BatchNormalization)	(None, 35, 35, 48)	144	conv2d_13[0][0]
batch_normalization_16 (BatchNormalization)	(None, 35, 35, 96)	288	conv2d_16[0][0]
activation_13 (Activation)	(None, 35, 35, 48)	0	batch_normalization_1
activation_16 (Activation)	(None, 35, 35, 96)	0	batch_normalization_1.
average_pooling2d_1 (AveragePooling2D)	(None, 35, 35, 256)	0	mixed0[0][0]
conv2d_12 (Conv2D)	(None, 35, 35, 64)	16,384	mixed0[0][0]
conv2d_14 (Conv2D)	(None, 35, 35, 64)	76,800	activation_13[0][0]
conv2d_17 (Conv2D)	(None, 35, 35, 96)	82,944	activation_16[0][0]
conv2d_18 (Conv2D)	(None, 35, 35, 64)	16,384	average_pooling2d_1[0.
batch_normalization_12 (BatchNormalization)	(None, 35, 35, 64)	192	conv2d_12[0][0]
batch_normalization_14 (BatchNormalization)	(None, 35, 35, 64)	192	conv2d_14[0][0]
batch_normalization_17	(None, 35, 35, 96)	288	conv2d_17[0][0]

(BatchNormalization)		1	
batch_normalization_18 (BatchNormalization)	(None, 35, 35, 64)	192	conv2d_18[0][0]
activation_12 (Activation)	(None, 35, 35, 64)	0	batch_normalization_1
activation_14 (Activation)	(None, 35, 35, 64)	0	batch_normalization_1
activation_17 (Activation)	(None, 35, 35, 96)	0	batch_normalization_1
activation_18 (Activation)	(None, 35, 35, 64)	0	batch_normalization_1
mixed1 (Concatenate)	(None, 35, 35, 288)	0	activation_12[0][0], activation_14[0][0], activation_17[0][0], activation_18[0][0]
conv2d_22 (Conv2D)	(None, 35, 35, 64)	18,432	 mixed1[0][0]
batch_normalization_22 (BatchNormalization)	(None, 35, 35, 64)	192	conv2d_22[0][0]
activation_22 (Activation)	(None, 35, 35, 64)	0	batch_normalization_2
conv2d_20 (Conv2D)	(None, 35, 35, 48)	13,824	mixed1[0][0]
conv2d_23 (Conv2D)	(None, 35, 35, 96)	55,296	activation_22[0][0]
batch_normalization_20 (BatchNormalization)	(None, 35, 35, 48)	144	conv2d_20[0][0]
batch_normalization_23 (BatchNormalization)	(None, 35, 35, 96)	288	conv2d_23[0][0]
activation_20 (Activation)	(None, 35, 35, 48)	0	batch_normalization_2
activation_23 (Activation)	(None, 35, 35, 96)	0	batch_normalization_2
average_pooling2d_2 (AveragePooling2D)	(None, 35, 35, 288)	0	mixed1[0][0]
conv2d 19 (Conv2D)	(None. 35. 35. 64)	18.432	 mixed1[0][0]

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conv2d_21 (Conv2D)	(None, 35, 35, 64)	76,800	activation_20[0][0]
conv2d_24 (Conv2D)	(None, 35, 35, 96)	82,944	activation_23[0][0]
conv2d_25 (Conv2D)	(None, 35, 35, 64)	18,432	average_pooling2d_2[0
batch_normalization_19 (BatchNormalization)	(None, 35, 35, 64)	192	conv2d_19[0][0]
batch_normalization_21 (BatchNormalization)	(None, 35, 35, 64)	192	conv2d_21[0][0]
batch_normalization_24 (BatchNormalization)	(None, 35, 35, 96)	288	conv2d_24[0][0]
batch_normalization_25 (BatchNormalization)	(None, 35, 35, 64)	192	conv2d_25[0][0]
activation_19 (Activation)	(None, 35, 35, 64)	0	batch_normalization_1
activation_21 (Activation)	(None, 35, 35, 64)	0	batch_normalization_2
activation_24 (Activation)	(None, 35, 35, 96)	0	batch_normalization_2
activation_25 (Activation)	(None, 35, 35, 64)	0	batch_normalization_2
mixed2 (Concatenate)	(None, 35, 35, 288)	0	activation_19[0][0], activation_21[0][0], activation_24[0][0], activation_25[0][0]
conv2d_27 (Conv2D)	(None, 35, 35, 64)	18,432	mixed2[0][0]
batch_normalization_27 (BatchNormalization)	(None, 35, 35, 64)	192	conv2d_27[0][0]
activation_27 (Activation)	(None, 35, 35, 64)	0	batch_normalization_2
conv2d_28 (Conv2D)	(None, 35, 35, 96)	55,296	activation_27[0][0]
batch_normalization_28 (BatchNormalization)	(None, 35, 35, 96)	288	conv2d_28[0][0]
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activation_28 (Activation)	(None, 35, 35, 96)	0	batch_normalization_2…
conv2d_26 (Conv2D)	(None, 17, 17, 384)	995,328	mixed2[0][0]
conv2d_29 (Conv2D)	(None, 17, 17, 96)	82,944	activation_28[0][0]
batch_normalization_26 (BatchNormalization)	(None, 17, 17, 384)	1,152	conv2d_26[0][0]
batch_normalization_29 (BatchNormalization)	(None, 17, 17, 96)	288	conv2d_29[0][0]
activation_26 (Activation)	(None, 17, 17, 384)	0	batch_normalization_2
activation_29 (Activation)	(None, 17, 17, 96)	0	batch_normalization_2
max_pooling2d_2 (MaxPooling2D)	(None, 17, 17, 288)	0	mixed2[0][0]
mixed3 (Concatenate)	(None, 17, 17, 768)	0	activation_26[0][0], activation_29[0][0], max_pooling2d_2[0][0]
conv2d_34 (Conv2D)	(None, 17, 17, 128)	98,304	mixed3[0][0]
batch_normalization_34 (BatchNormalization)	(None, 17, 17, 128)	384	conv2d_34[0][0]
activation_34 (Activation)	(None, 17, 17, 128)	0	batch_normalization_3
conv2d_35 (Conv2D)	(None, 17, 17, 128)	114,688	activation_34[0][0]
batch_normalization_35 (BatchNormalization)	(None, 17, 17, 128)	384	conv2d_35[0][0]
activation_35 (Activation)	(None, 17, 17, 128)	0	batch_normalization_3
conv2d_31 (Conv2D)	(None, 17, 17, 128)	98,304	mixed3[0][0]
conv2d_36 (Conv2D)	(None, 17, 17, 128)	114,688	activation_35[0][0]
batch_normalization_31 (BatchNormalization)	(None, 17, 17, 128)	384	conv2d_31[0][0]
batch normalization 36	(None. 17. 17. 128)	384	conv2d 36[0][0]

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(BatchNormalization)			<u>-</u> <u>-</u> <u>-</u>
activation_31 (Activation)	(None, 17, 17, 128)	0	batch_normalization_3
activation_36 (Activation)	(None, 17, 17, 128)	0	batch_normalization_3
conv2d_32 (Conv2D)	(None, 17, 17, 128)	114,688	activation_31[0][0]
conv2d_37 (Conv2D)	(None, 17, 17, 128)	114,688	activation_36[0][0]
batch_normalization_32 (BatchNormalization)	(None, 17, 17, 128)	384	conv2d_32[0][0]
batch_normalization_37 (BatchNormalization)	(None, 17, 17, 128)	384	conv2d_37[0][0]
activation_32 (Activation)	(None, 17, 17, 128)	0	 batch_normalization_3
activation_37 (Activation)	(None, 17, 17, 128)	0	 batch_normalization_3
average_pooling2d_3 (AveragePooling2D)	(None, 17, 17, 768)	0	mixed3[0][0]
conv2d_30 (Conv2D)	(None, 17, 17, 192)	147,456	mixed3[0][0]
conv2d_33 (Conv2D)	(None, 17, 17, 192)	172,032	activation_32[0][0]
conv2d_38 (Conv2D)	(None, 17, 17, 192)	172,032	activation_37[0][0]
conv2d_39 (Conv2D)	(None, 17, 17, 192)	147,456	 average_pooling2d_3[0
<pre>batch_normalization_30 (BatchNormalization)</pre>	(None, 17, 17, 192)	576	conv2d_30[0][0]
batch_normalization_33 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_33[0][0]
batch_normalization_38 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_38[0][0]
batch_normalization_39 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_39[0][0]
activation_30 (Activation)	(None, 17, 17, 192)	0	batch_normalization_3

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activation_33 (Activation)	(None, 17, 17, 192)	0	 batch_normalization_3
activation_38 (Activation)	(None, 17, 17, 192)	0	batch_normalization_3
activation_39 (Activation)	(None, 17, 17, 192)	0	batch_normalization_3
mixed4 (Concatenate)	(None, 17, 17, 768)	0	activation_30[0][0], activation_33[0][0], activation_38[0][0], activation_39[0][0]
conv2d_44 (Conv2D)	(None, 17, 17, 160)	122,880	 mixed4[0][0]
batch_normalization_44 (BatchNormalization)	(None, 17, 17, 160)	480	conv2d_44[0][0]
activation_44 (Activation)	(None, 17, 17, 160)	0	batch_normalization_4
conv2d_45 (Conv2D)	(None, 17, 17, 160)	179,200	 activation_44[0][0]
batch_normalization_45 (BatchNormalization)	(None, 17, 17, 160)	480	conv2d_45[0][0]
activation_45 (Activation)	(None, 17, 17, 160)	0	 batch_normalization_4
conv2d_41 (Conv2D)	(None, 17, 17, 160)	122,880	 mixed4[0][0]
conv2d_46 (Conv2D)	(None, 17, 17, 160)	179,200	activation_45[0][0]
batch_normalization_41 (BatchNormalization)	(None, 17, 17, 160)	480	conv2d_41[0][0]
batch_normalization_46 (BatchNormalization)	(None, 17, 17, 160)	480	conv2d_46[0][0]
activation_41 (Activation)	(None, 17, 17, 160)	0	batch_normalization_4
activation_46 (Activation)	(None, 17, 17, 160)	0	batch_normalization_4
conv2d_42 (Conv2D)	(None, 17, 17, 160)	179,200	activation_41[0][0]
conv2d 47 (Conv2D)	(None, 17, 17, 160)	179,200	activation 46[0][0]

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batch_normalization_42 (BatchNormalization)	(None, 17, 17, 160)	480	conv2d_42[0][0]
batch_normalization_47 (BatchNormalization)	(None, 17, 17, 160)	480	conv2d_47[0][0]
activation_42 (Activation)	(None, 17, 17, 160)	0	batch_normalization_4
activation_47 (Activation)	(None, 17, 17, 160)	0	batch_normalization_4
average_pooling2d_4 (AveragePooling2D)	(None, 17, 17, 768)	0	mixed4[0][0]
conv2d_40 (Conv2D)	(None, 17, 17, 192)	147,456	mixed4[0][0]
conv2d_43 (Conv2D)	(None, 17, 17, 192)	215,040	activation_42[0][0]
conv2d_48 (Conv2D)	(None, 17, 17, 192)	215,040	activation_47[0][0]
conv2d_49 (Conv2D)	(None, 17, 17, 192)	147,456	average_pooling2d_4[0
batch_normalization_40 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_40[0][0]
batch_normalization_43 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_43[0][0]
batch_normalization_48 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_48[0][0]
batch_normalization_49 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_49[0][0]
activation_40 (Activation)	(None, 17, 17, 192)	0	batch_normalization_4
activation_43 (Activation)	(None, 17, 17, 192)	0	batch_normalization_4
activation_48 (Activation)	(None, 17, 17, 192)	0	batch_normalization_4
activation_49 (Activation)	(None, 17, 17, 192)	0	batch_normalization_4
mixed5 (Concatenate)	(None, 17, 17, 768)	0	activation_40[0][0],

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			activation_43[0][0], activation_48[0][0], activation_49[0][0]
conv2d_54 (Conv2D)	(None, 17, 17, 160)	122,880	mixed5[0][0]
batch_normalization_54 (BatchNormalization)	(None, 17, 17, 160)	480	conv2d_54[0][0]
activation_54 (Activation)	(None, 17, 17, 160)	0	batch_normalization_5
conv2d_55 (Conv2D)	(None, 17, 17, 160)	179,200	activation_54[0][0]
batch_normalization_55 (BatchNormalization)	(None, 17, 17, 160)	480	conv2d_55[0][0]
activation_55 (Activation)	(None, 17, 17, 160)	0	batch_normalization_5
conv2d_51 (Conv2D)	(None, 17, 17, 160)	122,880	mixed5[0][0]
conv2d_56 (Conv2D)	(None, 17, 17, 160)	179,200	activation_55[0][0]
batch_normalization_51 (BatchNormalization)	(None, 17, 17, 160)	480	conv2d_51[0][0]
batch_normalization_56 (BatchNormalization)	(None, 17, 17, 160)	480	conv2d_56[0][0]
activation_51 (Activation)	(None, 17, 17, 160)	0	batch_normalization_5
activation_56 (Activation)	(None, 17, 17, 160)	0	batch_normalization_5
conv2d_52 (Conv2D)	(None, 17, 17, 160)	179,200	activation_51[0][0]
conv2d_57 (Conv2D)	(None, 17, 17, 160)	179,200	activation_56[0][0]
<pre>batch_normalization_52 (BatchNormalization)</pre>	(None, 17, 17, 160)	480	conv2d_52[0][0]
<pre>batch_normalization_57 (BatchNormalization)</pre>	(None, 17, 17, 160)	480	conv2d_57[0][0]
activation_52 (Activation)	(None, 17, 17, 160)	0	batch_normalization_5
activation 57	(None, 17, 17, 160)	0	 batch normalization 5

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(Activation)			
average_pooling2d_5 (AveragePooling2D)	(None, 17, 17, 768)	0	mixed5[0][0]
conv2d_50 (Conv2D)	(None, 17, 17, 192)	147,456	mixed5[0][0]
conv2d_53 (Conv2D)	(None, 17, 17, 192)	215,040	activation_52[0][0]
conv2d_58 (Conv2D)	(None, 17, 17, 192)	215,040	activation_57[0][0]
conv2d_59 (Conv2D)	(None, 17, 17, 192)	147,456	average_pooling2d_5[0
batch_normalization_50 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_50[0][0]
batch_normalization_53 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_53[0][0]
batch_normalization_58 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_58[0][0]
batch_normalization_59 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_59[0][0]
activation_50 (Activation)	(None, 17, 17, 192)	0	batch_normalization_5
activation_53 (Activation)	(None, 17, 17, 192)	0	batch_normalization_5
activation_58 (Activation)	(None, 17, 17, 192)	0	batch_normalization_5
activation_59 (Activation)	(None, 17, 17, 192)	0	batch_normalization_5
mixed6 (Concatenate)	(None, 17, 17, 768)	0	activation_50[0][0], activation_53[0][0], activation_58[0][0], activation_59[0][0]
conv2d_64 (Conv2D)	(None, 17, 17, 192)	147,456	mixed6[0][0]
batch_normalization_64 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_64[0][0]
activation_64 (Activation)	(None, 17, 17, 192)	0	batch_normalization_6

conv2d_65 (Conv2D)	(None, 17, 17, 192)	258,048	activation_64[0][0]
batch_normalization_65 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_65[0][0]
activation_65 (Activation)	(None, 17, 17, 192)	0	 batch_normalization_6
conv2d_61 (Conv2D)	(None, 17, 17, 192)	147,456	 mixed6[0][0]
conv2d_66 (Conv2D)	(None, 17, 17, 192)	258,048	activation_65[0][0]
batch_normalization_61 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_61[0][0]
batch_normalization_66 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_66[0][0]
activation_61 (Activation)	(None, 17, 17, 192)	0	batch_normalization_6
activation_66 (Activation)	(None, 17, 17, 192)	0	batch_normalization_6
conv2d_62 (Conv2D)	(None, 17, 17, 192)	258,048	activation_61[0][0]
conv2d_67 (Conv2D)	(None, 17, 17, 192)	258,048	activation_66[0][0]
batch_normalization_62 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_62[0][0]
batch_normalization_67 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_67[0][0]
activation_62 (Activation)	(None, 17, 17, 192)	0	batch_normalization_6
activation_67 (Activation)	(None, 17, 17, 192)	0	batch_normalization_6
average_pooling2d_6 (AveragePooling2D)	(None, 17, 17, 768)	0	mixed6[0][0]
conv2d_60 (Conv2D)	(None, 17, 17, 192)	147,456	mixed6[0][0]
conv2d_63 (Conv2D)	(None, 17, 17, 192)	258,048	activation_62[0][0]
conv2d_68 (Conv2D)	(None, 17, 17, 192)	258,048	activation_67[0][0]

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conv2d_69 (Conv2D)	(None, 17, 17, 192)	147,456	average_pooling2d_6[0
batch_normalization_60 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_60[0][0]
batch_normalization_63 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_63[0][0]
batch_normalization_68 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_68[0][0]
batch_normalization_69 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_69[0][0]
activation_60 (Activation)	(None, 17, 17, 192)	0	batch_normalization_6
activation_63 (Activation)	(None, 17, 17, 192)	0	batch_normalization_6
activation_68 (Activation)	(None, 17, 17, 192)	0	batch_normalization_6
activation_69 (Activation)	(None, 17, 17, 192)	0	batch_normalization_6
mixed7 (Concatenate)	(None, 17, 17, 768)	0	activation_60[0][0], activation_63[0][0], activation_68[0][0], activation_69[0][0]
conv2d_72 (Conv2D)	(None, 17, 17, 192)	147,456	mixed7[0][0]
batch_normalization_72 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_72[0][0]
activation_72 (Activation)	(None, 17, 17, 192)	0	batch_normalization_7
conv2d_73 (Conv2D)	(None, 17, 17, 192)	258,048	activation_72[0][0]
batch_normalization_73 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_73[0][0]
activation_73 (Activation)	(None, 17, 17, 192)	0	batch_normalization_7
conv2d_70 (Conv2D)	(None, 17, 17, 192)	147,456	mixed7[0][0]
conv2d 74 (Conv2D)	(None 17 17 102)	250 040	20tivation 72[0][0]

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CONVZU_/4 (CONVZU)	(NUITE, 1/, 1/, 192)	400,000	
batch_normalization_70 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_70[0][0]
batch_normalization_74 (BatchNormalization)	(None, 17, 17, 192)	576	conv2d_74[0][0]
activation_70 (Activation)	(None, 17, 17, 192)	0	batch_normalization_7
activation_74 (Activation)	(None, 17, 17, 192)	0	batch_normalization_7
conv2d_71 (Conv2D)	(None, 8, 8, 320)	552,960	activation_70[0][0]
conv2d_75 (Conv2D)	(None, 8, 8, 192)	331,776	activation_74[0][0]
batch_normalization_71 (BatchNormalization)	(None, 8, 8, 320)	960	conv2d_71[0][0]
batch_normalization_75 (BatchNormalization)	(None, 8, 8, 192)	576	conv2d_75[0][0]
activation_71 (Activation)	(None, 8, 8, 320)	0	batch_normalization_7
activation_75 (Activation)	(None, 8, 8, 192)	0	batch_normalization_7
<pre>max_pooling2d_3 (MaxPooling2D)</pre>	(None, 8, 8, 768)	0	mixed7[0][0]
mixed8 (Concatenate)	(None, 8, 8, 1280)	0	activation_71[0][0], activation_75[0][0], max_pooling2d_3[0][0]
conv2d_80 (Conv2D)	(None, 8, 8, 448)	573,440	 mixed8[0][0]
batch_normalization_80 (BatchNormalization)	(None, 8, 8, 448)	1,344	conv2d_80[0][0]
activation_80 (Activation)	(None, 8, 8, 448)	0	batch_normalization_8
conv2d_77 (Conv2D)	(None, 8, 8, 384)	491,520	mixed8[0][0]
conv2d_81 (Conv2D)	(None, 8, 8, 384)	1,548,288	activation_80[0][0]
batch_normalization_77	(None, 8, 8, 384)	1,152	conv2d_77[0][0]

(BatchNormalization)			
batch_normalization_81 (BatchNormalization)	(None, 8, 8, 384)	1,152	conv2d_81[0][0]
activation_77 (Activation)	(None, 8, 8, 384)	0	batch_normalization_7
activation_81 (Activation)	(None, 8, 8, 384)	0	batch_normalization_8
conv2d_78 (Conv2D)	(None, 8, 8, 384)	442,368	activation_77[0][0]
conv2d_79 (Conv2D)	(None, 8, 8, 384)	442,368	activation_77[0][0]
conv2d_82 (Conv2D)	(None, 8, 8, 384)	442,368	activation_81[0][0]
conv2d_83 (Conv2D)	(None, 8, 8, 384)	442,368	activation_81[0][0]
average_pooling2d_7 (AveragePooling2D)	(None, 8, 8, 1280)	0	mixed8[0][0]
conv2d_76 (Conv2D)	(None, 8, 8, 320)	409,600	mixed8[0][0]
batch_normalization_78 (BatchNormalization)	(None, 8, 8, 384)	1,152	conv2d_78[0][0]
batch_normalization_79 (BatchNormalization)	(None, 8, 8, 384)	1,152	conv2d_79[0][0]
batch_normalization_82 (BatchNormalization)	(None, 8, 8, 384)	1,152	conv2d_82[0][0]
batch_normalization_83 (BatchNormalization)	(None, 8, 8, 384)	1,152	conv2d_83[0][0]
conv2d_84 (Conv2D)	(None, 8, 8, 192)	245,760	average_pooling2d_7[0
batch_normalization_76 (BatchNormalization)	(None, 8, 8, 320)	960	conv2d_76[0][0]
activation_78 (Activation)	(None, 8, 8, 384)	0	batch_normalization_7
activation_79 (Activation)	(None, 8, 8, 384)	0	batch_normalization_7
activation_82 (Activation)	(None, 8, 8, 384)	0	batch_normalization_8

			1.7
activation_83 (Activation)	(None, 8, 8, 384)	0	 batch_normalization_8
batch_normalization_84 (BatchNormalization)	(None, 8, 8, 192)	576	conv2d_84[0][0]
activation_76 (Activation)	(None, 8, 8, 320)	0	batch_normalization_7
mixed9_0 (Concatenate)	(None, 8, 8, 768)	0	activation_78[0][0], activation_79[0][0]
concatenate (Concatenate)	(None, 8, 8, 768)	0	activation_82[0][0], activation_83[0][0]
activation_84 (Activation)	(None, 8, 8, 192)	0	batch_normalization_8
mixed9 (Concatenate)	(None, 8, 8, 2048)	0	activation_76[0][0], mixed9_0[0][0], concatenate[0][0], activation_84[0][0]
conv2d_89 (Conv2D)	(None, 8, 8, 448)	917,504	mixed9[0][0]
batch_normalization_89 (BatchNormalization)	(None, 8, 8, 448)	1,344	conv2d_89[0][0]
activation_89 (Activation)	(None, 8, 8, 448)	0	batch_normalization_8
conv2d_86 (Conv2D)	(None, 8, 8, 384)	786,432	mixed9[0][0]
conv2d_90 (Conv2D)	(None, 8, 8, 384)	1,548,288	activation_89[0][0]
batch_normalization_86 (BatchNormalization)	(None, 8, 8, 384)	1,152	conv2d_86[0][0]
batch_normalization_90 (BatchNormalization)	(None, 8, 8, 384)	1,152	conv2d_90[0][0]
activation_86 (Activation)	(None, 8, 8, 384)	0	batch_normalization_8
activation_90 (Activation)	(None, 8, 8, 384)	0	batch_normalization_9
conv2d_87 (Conv2D)	(None, 8, 8, 384)	442,368	activation_86[0][0]
	 	 	[

conv2d_88 (Conv2D)	(None, 8, 8, 384)	442,368	activation_86[0][0]
conv2d_91 (Conv2D)	(None, 8, 8, 384)	442,368	activation_90[0][0]
conv2d_92 (Conv2D)	(None, 8, 8, 384)	442,368	activation_90[0][0]
average_pooling2d_8 (AveragePooling2D)	(None, 8, 8, 2048)	0	mixed9[0][0]
conv2d_85 (Conv2D)	(None, 8, 8, 320)	655,360	mixed9[0][0]
batch_normalization_87 (BatchNormalization)	(None, 8, 8, 384)	1,152	conv2d_87[0][0]
batch_normalization_88 (BatchNormalization)	(None, 8, 8, 384)	1,152	conv2d_88[0][0]
batch_normalization_91 (BatchNormalization)	(None, 8, 8, 384)	1,152	conv2d_91[0][0]
batch_normalization_92 (BatchNormalization)	(None, 8, 8, 384)	1,152	conv2d_92[0][0]

```
# Step 2: Recompile the model with a lower learning rate
model.compile(optimizer=Adam(learning_rate=1e-5), loss='binary_crossentropy', metrics=['accuracy'])
# Load CSV file containing image labels (DR or Not DR)
df labels = pd.read csv('/content/drive/MyDrive/labels mbrset.csv') # Replace with your CSV file path
# Convert DR labels to integers for class weight computation
df_{abels}['dr_{diagnosis_int'}] = df_{abels}['final_icdr'].apply(lambda x: 0 if x == 0 else 1)
# Define paths and parameters
IMG SIZE = (299, 299)
BATCH SIZE = 32
TRAIN PATH = '/content/drive/MyDrive/images'
# ImageDataGenerator for training and validation
train_datagen = ImageDataGenerator(
    rescale=1./255,
    validation split=0.2, # Use 20% of data for validation
    rotation_range=20,
    width shift range=0.2,
    height shift range=0.2,
    shear range=0.2,
    zoom range=0.2,
    horizontal flip=True
# Load training and validation data
train_generator = train_datagen.flow_from_dataframe(
    dataframe=df_labels,
    directory=TRAIN PATH,
    x_col='file', # Image filename column
    y_col='dr_diagnosis_int', # Single binary output (DR or Not DR)
    target size=IMG SIZE,
    batch size=BATCH SIZE,
    class_mode='raw',
    subset='training',
    shuffle=True
)
```

```
valid generator = train datagen.flow from dataframe(
    dataframe=df labels,
    directory=TRAIN PATH,
    x_col='file',
    y_col='dr_diagnosis_int',
    target_size=IMG_SIZE,
    batch size=BATCH SIZE,
    class mode='raw',
    subset='validation',
    shuffle=True
    Found 4132 validated image filenames.
     Found 1032 validated image filenames.
# Step 3: Handle class imbalance by computing class weights
class_weights = compute_class_weight('balanced', classes=np.unique(df_labels['dr_diagnosis_int']), y=df_labels['dr_diagnosis_int'])
class_weight_dict = dict(enumerate(class_weights))
# Step 4: Train the model with fine-tuning
checkpoint = ModelCheckpoint('best finetuned model.keras', monitor='val loss', save best only=True, mode='min')
early stopping = EarlyStopping(monitor='val loss', patience=5, restore best weights=True)
history = model.fit(
    train generator,
    epochs=10, # Reduce the number of epochs to avoid overfitting during fine-tuning
    validation data=valid generator,
    class weight=class weight dict,
    callbacks=[checkpoint, early stopping],
    steps per epoch=train generator.samples // BATCH SIZE,
    validation steps=valid generator.samples // BATCH SIZE
    Epoch 1/10
     /usr/local/lib/python3.10/dist-packages/keras/src/trainers/data_adapters/py_dataset_adapter.py:121: UserWarning: Your `PyDataset` class
       self._warn_if_super_not_called()
     129/129 -
                                — 2157s 16s/step - accuracy: 0.6940 - loss: 0.6021 - val accuracy: 0.7480 - val loss: 0.5398
     Epoch 2/10
                               — 22:40 11s/step - accuracy: 0.5625 - loss: 0.8582/usr/lib/python3.10/contextlib.py:153: UserWarning: Your in
       self.gen.throw(typ, value, traceback)
     129/129 -
                                — 13s 15ms/step - accuracy: 0.5625 - loss: 0.8582 - val_accuracy: 0.6250 - val_loss: 0.6068
     Epoch 3/10
```

```
— 2012s 15s/step - accuracy: 0.7163 - loss: 0.5552 - val accuracy: 0.7188 - val loss: 0.5790
     129/129 -
     Epoch 4/10
     129/129
                                 - 16s 31ms/step - accuracy: 0.6875 - loss: 0.4989 - val accuracy: 0.7500 - val loss: 0.4132
     Epoch 5/10
                                - 1966s 15s/step - accuracy: 0.7500 - loss: 0.5243 - val accuracy: 0.7158 - val loss: 0.5658
     129/129
     Epoch 6/10
                                - 63s 419ms/step - accuracy: 0.7812 - loss: 0.6065 - val accuracy: 0.6250 - val loss: 0.7653
     129/129 -
     Epoch 7/10
     129/129 -
                                 - 1997s 15s/step - accuracy: 0.7452 - loss: 0.5217 - val accuracy: 0.7168 - val loss: 0.5648
     Epoch 8/10
     129/129 -
                                 - 14s 15ms/step - accuracy: 0.8125 - loss: 0.4402 - val accuracy: 0.7500 - val loss: 0.6014
     Epoch 9/10
     129/129 -
                                - 1953s 15s/step - accuracy: 0.7615 - loss: 0.4905 - val accuracy: 0.7178 - val loss: 0.5621
# Save the fine-tuned model
model.save('/content/drive/MyDrive/diabetic retinopathy finetuned model.h5')
    WARNING:absl:You are saving your model as an HDF5 file via `model.save()` or `keras.saving.save model(model)`. This file format is consi
# Optional: Evaluate the model
val predictions = model.predict(valid generator)
val predictions binary = (val predictions > 0.5).astype(int)
    33/33 -
                              -- 355s 11s/step
# Performance evaluation
from sklearn.metrics import classification report, confusion matrix, roc auc score
true labels = valid generator.labels # True labels from validation set
print(classification report(true labels, val predictions binary, target names=['Not DR', 'DR']))
cm = confusion matrix(true labels, val predictions binary)
import seaborn as sns
import matplotlib.pyplot as plt
sns.heatmap(cm, annot=True, fmt='d', cmap='Blues', xticklabels=['Not DR', 'DR'], yticklabels=['Not DR', 'DR'])
plt.title('Confusion Matrix')
plt.show()
roc auc = roc auc score(true labels, val predictions)
```

10/10/24, 11:45 PM print(T kuc Auc Score: {roc_auc..4T})

₹	precision	recall	f1-score	support
Not DR	0.73	0.65	0.69	748
DR	0.29	0.38	0.33	284
accuracy			0.58	1032
macro avg	0.51	0.52	0.51	1032
weighted avg	0.61	0.58	0.59	1032

