```
[]
                x = GlobalAveragePooling2D()(base vgg.output)
                x = Dense(512, activation="relu")(x)
                 x = Dense(64, activation="relu")(x)
                output = Dense(10, activation="softmax")(x) # 10 classes
                model_vgg = Model(inputs=base_vgg.input, outputs=output)
                opt = SGD(learning_rate=0.0001)
                model_vgg.compile(optimizer=opt, loss="categorical_crossentropy", metrics=["accuracy"])
                model vgg.summarv()
                history_vgg = model_vgg.fit(
                       train gen.
                       validation_data=val_gen
                   (GlobalAveragePooling2D)
          ₹
                    dense_3 (Dense)
                                                                        (None, 512)
                                                                                                                        262,656
                   dense 4 (Dense)
                                                                        (None, 64)
                                                                                                                          32,832
                   dense 5 (Dense)
                                                                        (None, 10)
                                                                                                                              650
                Total params: 15,010,826 (57.26 MB)
Trainable params: 296,138 (1.13 MB)
Non-trainable params: 14,714,888 (56.13 MB)
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                   self._warn_if_super_not_called()
                 Epoch 1/50
                 59/59
                                                         - 49s 802ms/step - accuracy: 0.1044 - loss: 2.3517 - val accuracy: 0.1050 - val loss: 2.3528
                 Epoch 2/50
                                                        - 31s 528ms/step - accuracy: 0.1005 - loss: 2.3509 - val accuracy: 0.1100 - val loss: 2.3404
                59/59 -
                 Epoch 3/50
                                                        - 32s 547ms/step - accuracy: 0.1113 - loss: 2.3301 - val_accuracy: 0.1300 - val loss: 2.3299
                59/59 -
                 Epoch 4/50
                59/59 -
                                                         - 31s 527ms/step - accuracy: 0.1114 - loss: 2.3067 - val accuracy: 0.1300 - val loss: 2.3214
                Epoch 5/50
59/59
                                                          31s 518ms/step - accuracy: 0.1400 - loss: 2.2977 - val_accuracy: 0.1400 - val_loss: 2.3142
                 Epoch 6/50
                 59/59
                                                          30s 507ms/step - accuracy: 0.1444 - loss: 2.3024 - val_accuracy: 0.1600 - val_loss: 2.3083
                Epoch 7/50
                 59/59
                                                          31s 532ms/step - accuracy: 0.1648 - loss: 2.2905 - val_accuracy: 0.1650 - val_loss: 2.3034
                Epoch 8/50
                 59/59
                                                          30s 500ms/step - accuracy: 0.1864 - loss: 2.2779 - val_accuracy: 0.1750 - val_loss: 2.2992
                Epoch 9/50
                59/59
                                                        - 31s 529ms/step - accuracy: 0.1983 - loss: 2.2644 - val_accuracy: 0.1800 - val_loss: 2.2957
                Epoch 10/50
                 59/59
                                                         30s 511ms/step - accuracy: 0.1837 - loss: 2.2747 - val accuracy: 0.1750 - val loss: 2.2928
                Epoch 11/50
                59/59
                                                        - 31s 530ms/step - accuracy: 0.1971 - loss: 2.2665 - val_accuracy: 0.1700 - val_loss: 2.2903
                 Epoch 12/50
                                                        - 32s 547ms/step - accuracy: 0.1958 - loss: 2.2623 - val accuracy: 0.1800 - val loss: 2.2879
                59/59 -
                 Epoch 13/50
                                                         - 30s 514ms/step - accuracy: 0.2032 - loss: 2.2575 - val accuracy: 0.1800 - val loss: 2.2859
                59/59 -
                 . -
Epoch 14/50
                                                          34s 571ms/step - accuracy: 0.2139 - loss: 2.2502 - val accuracy: 0.1850 - val loss: 2.2841
                 59/59 -
                 Epoch 15/50
                 59/59
                                                          30s 510ms/step - accuracy: 0.2087 - loss: 2.2528 - val_accuracy: 0.1700 - val_loss: 2.2822
                 Enoch 16/50
                 59/59
                                                          31s 527ms/step - accuracy: 0.2329 - loss: 2.2490 - val_accuracy: 0.1800 - val_loss: 2.2806
                Epoch 17/50
                 59/59
                                                          34s 583ms/step - accuracy: 0.2411 - loss: 2.2422 - val_accuracy: 0.1800 - val_loss: 2.2794
                 Fnoch 18/50
                 59/59
                                                          30s 510ms/step - accuracy: 0.2618 - loss: 2.2371 - val_accuracy: 0.1900 - val_loss: 2.2781
                Epoch 19/50
                 59/59
                                                          30s 507ms/step - accuracy: 0.2565 - loss: 2.2395 - val_accuracy: 0.1900 - val_loss: 2.2769
                Epoch 20/50
                                                          32s 545ms/step - accuracy: 0.2713 - loss: 2.2352 - val accuracy: 0.1950 - val loss: 2.2757
                59/59
                 Epoch 21/50
                 59/59 -
                                                         - 31s 522ms/step - accuracy: 0.2924 - loss: 2.2340 - val accuracy: 0.2050 - val loss: 2.2746
                 Epoch 22/50
                                                         33s 556ms/step - accuracy: 0.2824 - loss: 2.2286 - val accuracy: 0.1900 - val loss: 2.2737
                59/59 •
                 Epoch 23/50
[]
                import matplotlib.pyplot as plt
                 plt.plot(history_vgg.history['accuracy'], label="train accuracy")
                plt.plot(history_vgg.history['val_accuracy'], label="validation accuracy")
plt.xlabel("Epoch")
                 plt.ylabel("Accuracy"
                 nlt.legend()
                 plt.title("Train vs Validation Accuracy")
                 plt.show()
          ₹
                                                         Train vs Validation Accuracy

    train accuracy

                       0.40
                                        validation accuracy
                       0.35
                       0.30
                       0.25
                       0.20
                       0.15
                       0.10
```

