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import tensorflow as tf
from tensorflow.keras.preprocessing.image import ImageDataGenerator
import os
import matplotlib.pyplot as plt
import numpy as np
from tensorflow.keras.preprocessing import
image dataset from directory
from tensorflow.keras.layers import Dense, Flatten, Conv2D,
MaxPooling2D, Dropout, GlobalAveragePooling2D,
Input, BatchNormalization
from tensorflow.keras.preprocessing.image import ImageDataGenerator
from tensorflow.keras.applications import MobileNetV2
from tensorflow.keras.models import Sequential
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.utils import to categorical
from tensorflow.keras.models import load model
import warnings
warnings.filterwarnings('ignore')
from tensorflow.keras.preprocessing.image import img to array,load img
from tensorflow.keras.preprocessing import image
from tensorflow.keras.regularizers import 12
from tensorflow.keras.callbacks import EarlyStopping
!mkdir -p ~/.kaggle
!cp kaggle.json ~/.kaggle/
!chmod 600 /root/.kaggle/kaggle.json
!kaggle datasets download -d samuelcortinhas/cats-and-dogs-image-
classification
Dataset URL: https://www.kaggle.com/datasets/samuelcortinhas/cats-and-
dogs-image-classification
License(s): CCO-1.0
Downloading cats-and-dogs-image-classification.zip to /content
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inflating: train/dogs/dog 375.jpg
inflating: train/dogs/dog 376.jpg
inflating: train/dogs/dog 378.jpg
inflating: train/dogs/dog 381.jpg
inflating: train/dogs/dog 389.jpg
inflating: train/dogs/dog 39.jpg
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inflating: train/dogs/dog 393.jpg
inflating: train/dogs/dog 4.jpg
inflating: train/dogs/dog 401.jpg
inflating: train/dogs/dog 402.jpg
inflating: train/dogs/dog 403.jpg
inflating: train/dogs/dog 404.jpg
inflating: train/dogs/dog 405.jpg
inflating: train/dogs/dog 407.jpg
inflating: train/dogs/dog 408.jpg
inflating: train/dogs/dog 409.jpg
inflating: train/dogs/dog 412.jpg
inflating: train/dogs/dog 414.jpg
inflating: train/dogs/dog 417.jpg
inflating: train/dogs/dog 418.jpg
inflating: train/dogs/dog 419.jpg
inflating: train/dogs/dog 423.jpg
inflating: train/dogs/dog 427.jpg
inflating: train/dogs/dog 428.jpg
inflating: train/dogs/dog 429.jpg
inflating: train/dogs/dog 432.jpg
inflating: train/dogs/dog 433.jpg
inflating: train/dogs/dog 435.jpg
inflating: train/dogs/dog_436.jpg
inflating: train/dogs/dog 437.jpg
inflating: train/dogs/dog 438.jpg
inflating: train/dogs/dog 440.jpg
inflating: train/dogs/dog 441.jpg
```

```
inflating: train/dogs/dog 444.jpg
inflating: train/dogs/dog 445.jpg
inflating: train/dogs/dog 446.jpg
inflating: train/dogs/dog 447.jpg
inflating: train/dogs/dog 448.jpg
inflating: train/dogs/dog 45.jpg
inflating: train/dogs/dog 450.jpg
inflating: train/dogs/dog 451.jpg
inflating: train/dogs/dog 452.jpg
inflating: train/dogs/dog 458.jpg
inflating: train/dogs/dog 466.jpg
inflating: train/dogs/dog 468.jpg
inflating: train/dogs/dog 47.jpg
inflating: train/dogs/dog 471.jpg
inflating: train/dogs/dog 473.jpg
inflating: train/dogs/dog_474.jpg
inflating: train/dogs/dog 475.jpg
inflating: train/dogs/dog 478.jpg
inflating: train/dogs/dog 479.jpg
inflating: train/dogs/dog 480.jpg
inflating: train/dogs/dog 483.jpg
inflating: train/dogs/dog 484.jpg
inflating: train/dogs/dog 485.jpg
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inflating: train/dogs/dog 489.jpg
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inflating: train/dogs/dog 492.jpg
inflating: train/dogs/dog 494.jpg
inflating: train/dogs/dog 496.jpg
inflating: train/dogs/dog 50.jpg
inflating: train/dogs/dog 501.jpg
inflating: train/dogs/dog 502.jpg
inflating: train/dogs/dog 504.jpg
inflating: train/dogs/dog 505.jpg
inflating: train/dogs/dog 509.jpg
inflating: train/dogs/dog 51.jpg
inflating: train/dogs/dog 512.jpg
inflating: train/dogs/dog_513.jpg
inflating: train/dogs/dog_514.jpg
inflating: train/dogs/dog 523.jpg
inflating: train/dogs/dog 525.jpg
inflating: train/dogs/dog 529.jpg
inflating: train/dogs/dog_532.jpg
inflating: train/dogs/dog 533.jpg
inflating: train/dogs/dog 535.jpg
inflating: train/dogs/dog 54.jpg
inflating: train/dogs/dog 540.jpg
inflating: train/dogs/dog 541.jpg
inflating: train/dogs/dog 543.jpg
```

```
inflating: train/dogs/dog_545.jpg
inflating: train/dogs/dog 546.jpg
inflating: train/dogs/dog 549.jpg
inflating: train/dogs/dog 550.jpg
inflating: train/dogs/dog 552.jpg
inflating: train/dogs/dog_553.jpg
inflating: train/dogs/dog 554.jpg
inflating: train/dogs/dog 556.jpg
inflating: train/dogs/dog 558.jpg
inflating: train/dogs/dog 559.jpg
inflating: train/dogs/dog 56.jpg
inflating: train/dogs/dog 560.jpg
inflating: train/dogs/dog 564.jpg
inflating: train/dogs/dog 566.jpg
inflating: train/dogs/dog_57.jpg
inflating: train/dogs/dog_570.jpg
inflating: train/dogs/dog 571.jpg
inflating: train/dogs/dog_572.jpg
inflating: train/dogs/dog 576.jpg
inflating: train/dogs/dog 579.jpg
inflating: train/dogs/dog 580.jpg
inflating: train/dogs/dog 582.jpg
inflating: train/dogs/dog 583.jpg
inflating: train/dogs/dog 589.jpg
inflating: train/dogs/dog 591.jpg
inflating: train/dogs/dog 596.jpg
inflating: train/dogs/dog 62.jpg
inflating: train/dogs/dog 67.jpg
inflating: train/dogs/dog 69.jpg
inflating: train/dogs/dog 7.jpg
inflating: train/dogs/dog 70.jpg
inflating: train/dogs/dog 72.jpg
inflating: train/dogs/dog 74.jpg
inflating: train/dogs/dog 76.jpg
inflating: train/dogs/dog 77.jpg
inflating: train/dogs/dog 78.jpg
inflating: train/dogs/dog 8.jpg
inflating: train/dogs/dog 80.jpg
inflating: train/dogs/dog 82.jpg
inflating: train/dogs/dog 84.jpg
inflating: train/dogs/dog 85.jpg
inflating: train/dogs/dog 86.jpg
inflating: train/dogs/dog 9.jpg
inflating: train/dogs/dog 90.jpg
inflating: train/dogs/dog_92.jpg
inflating: train/dogs/dog 93.jpg
inflating: train/dogs/dog 97.jpg
inflating: train/dogs/dog 98.jpg
inflating: train/dogs/dog 99.jpg
```

```
train='/content/train'
test='/content/test'
train datagen = ImageDataGenerator(
    rescale=1./255,
    horizontal flip=True,
    rotation range=30,
    zoom range=0.2,
    validation split=0.2)
test datagen = ImageDataGenerator(
    rescale=1./255,
    horizontal flip=True,
    rotation range=30,
    zoom range=0.2,)
train gen = train datagen.flow from directory(
                                         train,
                                         class mode='categorical',
                                         batch size = 32,
                                         target size = (224, 224),
                                         shuffle = True,
                                         seed = 42,
                                         subset = "training")
Found 447 images belonging to 2 classes.
val gen = train datagen.flow from directory(
                                         train,
                                         class mode ="categorical",
                                         batch size = 32,
                                         target size = (224,224),
                                         shuffle = False,
                                         seed = 42,
                                         subset = "validation")
Found 110 images belonging to 2 classes.
test set = test datagen.flow from directory(
                                         class mode ="categorical",
                                         batch_size = 32,
                                         target size = (224,224),
                                         shuffle = False,
                                         seed = 42)
Found 140 images belonging to 2 classes.
base1 = MobileNetV2(weights='imagenet', include top=False,
input shape=(224, 224, 3))
```

```
Downloading data from https://storage.googleapis.com/tensorflow/keras-
applications/mobilenet v2/
mobilenet_v2_weights_tf_dim_ordering_tf_kernels_1.0_224_no_top.h5
9406464/9406464 -
                             ---- Os Ous/step
m1 = Sequential()
m1.add(Input(shape=(224, 224, 3)))
m1.add(base1) # Add MobileNetV2 as the base
m1.add(GlobalAveragePooling2D())
m1.add(Dense(1024, activation='relu'))
m1.add(Dropout(0.5))
m1.add(Dense(512, activation='relu'))
m1.add(Dropout(0.5))
m1.add(Dense(2, activation='softmax')) # 2 output units for binary
classification
'''for layer in m1.layers[-20:]:
    layer.trainable = False
base1.trainable = False
m1.compile(optimizer=Adam(learning rate=0.0001),
             loss='categorical crossentropy',
             metrics=['accuracy'])
m1.summary()
Model: "sequential"
Layer (type)
                                       Output Shape
Param #
mobilenetv2 1.00 224 (Functional)
                                      (None, 7, 7, 1280)
2,257,984
 global average pooling2d
                                       (None, 1280)
  (GlobalAveragePooling2D)
 dense (Dense)
                                       (None, 1024)
1,311,744
 dropout (Dropout)
                                       (None, 1024)
```

```
dense 1 (Dense)
                                      (None, 512)
524,800
 dropout 1 (Dropout)
                                      (None, 512)
0
 dense 2 (Dense)
                                      (None, 2)
1,026
Total params: 4,095,554 (15.62 MB)
Trainable params: 1,837,570 (7.01 MB)
Non-trainable params: 2,257,984 (8.61 MB)
history = m1.fit(train gen, validation data=val gen, epochs=10)
Epoch 1/10
                  38s 2s/step - accuracy: 0.5733 - loss:
14/14 —
0.8644 - val_accuracy: 0.9091 - val_loss: 0.2575
Epoch 2/10
14/14 —
                     —— 10s 483ms/step - accuracy: 0.8600 - loss:
0.3318 - val_accuracy: 0.8909 - val_loss: 0.2018
Epoch 3/10
14/14 -
                  ———— 9s 421ms/step - accuracy: 0.9024 - loss:
0.2305 - val accuracy: 0.9000 - val loss: 0.1833
Epoch 4/10
              8s 327ms/step - accuracy: 0.9272 - loss:
14/14 ——
0.1891 - val accuracy: 0.9091 - val loss: 0.1992
Epoch 5/10
                     —— 11s 329ms/step - accuracy: 0.9066 - loss:
14/14 —
0.2175 - val_accuracy: 0.9364 - val_loss: 0.1299
Epoch 6/10
                     —— 12s 425ms/step - accuracy: 0.9192 - loss:
14/14 –
0.2087 - val_accuracy: 0.9182 - val_loss: 0.1546
Epoch 7/10
                      9s 420ms/step - accuracy: 0.9478 - loss:
0.1485 - val accuracy: 0.9364 - val loss: 0.1348
Epoch 8/10
                     —— 9s 427ms/step - accuracy: 0.9704 - loss:
14/14 -
0.0994 - val accuracy: 0.9364 - val loss: 0.1522
Epoch 9/10
          9s 335ms/step - accuracy: 0.9539 - loss:
14/14 —
0.1058 - val_accuracy: 0.9091 - val_loss: 0.1939
```

```
Epoch 10/10
              9s 352ms/step - accuracy: 0.9426 - loss:
14/14 -
0.1415 - val accuracy: 0.9364 - val loss: 0.1579
m1.evaluate(test set)
[0.13037477433681488, 0.9357143044471741]
# Function to load and preprocess the image
def load_and_prep_image(img_path, img_size=(224, 224)):
   img = load_img(img_path, target_size=img_size) # Load and resize
   img = img to array(img) # Convert to array
   img = np.expand_dims(img, axis=0) # Add batch dimension
   img = img / 255.0 # Normalize the image
   return imq
# Function to plot the image and prediction
def plot image with prediction(img path, model):
   img = load and prep image(img path) # Preprocess the image
   pred = model.predict(img) # Predict the class
   class_names = ['Cat', 'Dog'] # Adjust class names as per your
dataset
   predicted_class = np.argmax(pred, axis=1)[0] # Get the predicted
class (0 or 1)
   predicted label = class names[predicted class] # Get the
corresponding label
   # Display the image and prediction
   img to display = load img(img path, target size=(224, 224)) #
Load the image for display
   plt.imshow(img to display)
   plt.title(f"Predicted Label: {predicted label}")
   plt.axis("off")
   plt.show()
# Example of using the function
# Provide the path to an image in your test dataset
test img path = "/content/test/cats/cat 119.jpg"
plot image with prediction(test img path, m1)
             Os 50ms/step
1/1 ——
```

Predicted Label: Cat



```
m1.save('cat and dog.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 considered legacy. We recommend using instead the native Keras
format, e.g. `model.save('my model.keras')` or
`keras.saving.save model(model, 'my model.keras')`.
model=Sequential()
model.add(Conv2D(64,(3,3),activation='relu',input shape=(224,224,3)))
model.add(MaxPooling2D((2,2)))
model.add(Conv2D(128,(3,3),activation='relu'))
model.add(MaxPooling2D((2,2)))
model.add(Conv2D(128,(3,3),activation='relu'))
model.add(MaxPooling2D((2,2)))
model.add(Conv2D(512,(3,3),activation='relu'))
model.add(MaxPooling2D((2,2)))
model.add(Flatten())
model.add(Dense(1024,activation='relu'))
model.add(Dense(2,activation='softmax'))
```

```
model.compile(optimizer=Adam(learning rate=0.001),
             loss='categorical crossentropy',
             metrics=['accuracy'])
model.summary()
Model: "sequential 7"
Layer (type)
                                     Output Shape
Param # |
 conv2d 24 (Conv2D)
                                     (None, 222, 222, 64)
1,792
 max pooling2d 24 (MaxPooling2D)
                                     (None, 111, 111, 64)
0 |
conv2d 25 (Conv2D)
                                      (None, 109, 109, 128)
73,856
 max pooling2d 25 (MaxPooling2D)
                                     (None, 54, 54, 128)
 conv2d_26 (Conv2D)
                                      (None, 52, 52, 128)
147,584
 max_pooling2d_26 (MaxPooling2D)
                                     (None, 26, 26, 128)
conv2d 27 (Conv2D)
                                      (None, 24, 24, 512)
590,336 T
 max pooling2d 27 (MaxPooling2D) | (None, 12, 12, 512)
 flatten_6 (Flatten)
                                     (None, 73728)
0
```

```
dense 15 (Dense)
                                    (None, 1024)
75,498,496
dense 16 (Dense)
                                    (None, 2)
2,050
Total params: 76,314,114 (291.12 MB)
Trainable params: 76,314,114 (291.12 MB)
Non-trainable params: 0 (0.00 B)
history = model.fit(
   train_gen,
   epochs=100,
   validation data=val gen,
)
Epoch 1/100
                  ———— 16s 726ms/step - accuracy: 0.4999 - loss:
14/14 —
2.0910 - val accuracy: 0.5000 - val loss: 0.7230
0.6932 - val accuracy: 0.5091 - val loss: 0.6930
Epoch 3/100
             9s 384ms/step - accuracy: 0.5185 - loss:
14/14 -----
0.6932 - val accuracy: 0.5545 - val loss: 0.6927
Epoch 4/100
                 _____ 10s 394ms/step - accuracy: 0.5127 - loss:
14/14 —
0.6929 - val accuracy: 0.5182 - val loss: 0.6910
Epoch 5/100
                    ——— 11s 414ms/step - accuracy: 0.5170 - loss:
14/14 —
0.6920 - val accuracy: 0.5545 - val loss: 0.6865
Epoch 6/100
                    —— 11s 523ms/step - accuracy: 0.5386 - loss:
14/14 -
0.6898 - val accuracy: 0.5727 - val loss: 0.6780
Epoch 7/100
                 _____ 10s 528ms/step - accuracy: 0.5272 - loss:
14/14 —
0.6881 - val accuracy: 0.5273 - val loss: 0.6826
Epoch 8/100
            ______ 19s 392ms/step - accuracy: 0.5478 - loss:
14/14 -----
0.6900 - val accuracy: 0.5727 - val_loss: 0.6616
Epoch 9/100
               _____ 12s 466ms/step - accuracy: 0.6065 - loss:
14/14 —
0.6598 - val accuracy: 0.6455 - val loss: 0.6520
Epoch 10/100
14/14 -
                     —— 19s 445ms/step - accuracy: 0.6831 - loss:
```

```
0.6258 - val accuracy: 0.6545 - val_loss: 0.6500
Epoch 11/100
                                   9s 390ms/step - accuracy: 0.6933 - loss:
14/14 ———
0.6131 - val accuracy: 0.6364 - val loss: 0.6496
Epoch 12/100
                                     ———— 10s 390ms/step - accuracy: 0.6753 - loss:
0.6199 - val accuracy: 0.6364 - val loss: 0.5876
Epoch 13/100
                                        ———— 10s 430ms/step - accuracy: 0.6806 - loss:
14/14 ---
0.5943 - val accuracy: 0.6455 - val loss: 0.6254
Epoch 14/100 10s 512ms/step - accuracy: 0.6674 - loss:
0.6211 - val accuracy: 0.6273 - val loss: 0.6365
Epoch 15/100 10/14 10/14 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16 10/16
0.5844 - val accuracy: 0.6636 - val_loss: 0.6402
Epoch 16/100 9s 394ms/step - accuracy: 0.7021 - loss:
0.5824 - val accuracy: 0.6545 - val loss: 0.5962
Epoch 17/100
0.5866 - val accuracy: 0.6636 - val loss: 0.6117
Epoch 18/100
                                       _____ 20s 506ms/step - accuracy: 0.7538 - loss:
0.5407 - val_accuracy: 0.6545 - val_loss: 0.6034
Epoch 19/100
                                    ———— 9s 436ms/step - accuracy: 0.7280 - loss:
14/14 -
0.5623 - val accuracy: 0.6727 - val loss: 0.6504
Epoch 20/100 9s 388ms/step - accuracy: 0.6674 - loss:
0.5923 - val accuracy: 0.6364 - val loss: 0.6156
Epoch 21/100 14/14 ______ 11s 377ms/step - accuracy: 0.6990 - loss:
0.5579 - val accuracy: 0.7000 - val loss: 0.6289
Epoch 22/100 14/14 ______ 10s 484ms/step - accuracy: 0.7196 - loss:
0.5602 - val accuracy: 0.6545 - val loss: 0.5997
Epoch 23/100
14/14 ————
                             ______ 10s 475ms/step - accuracy: 0.7386 - loss:
0.5266 - val accuracy: 0.6091 - val loss: 0.7577
Epoch 24/100
                                      9s 387ms/step - accuracy: 0.7053 - loss:
14/14 ——
0.5625 - val_accuracy: 0.6545 - val_loss: 0.6314
Epoch 25/100
                                       ———— 10s 398ms/step - accuracy: 0.7563 - loss:
14/14 —
0.5185 - val_accuracy: 0.6818 - val_loss: 0.5912
0.4639 - val accuracy: 0.6455 - val loss: 0.6030
```

```
0.4794 - val accuracy: 0.6455 - val loss: 0.6514
Epoch 28/100 9s 407ms/step - accuracy: 0.7238 - loss:
0.5229 - val accuracy: 0.6818 - val loss: 0.6509
Epoch 29/100
              ______ 10s 393ms/step - accuracy: 0.7978 - loss:
14/14 ———
0.4360 - val accuracy: 0.6455 - val loss: 0.7280
Epoch 30/100
                _____ 10s 458ms/step - accuracy: 0.7777 - loss:
14/14 ———
0.5005 - val_accuracy: 0.6636 - val_loss: 0.5979
Epoch 31/100
                 ———— 10s 483ms/step - accuracy: 0.7762 - loss:
14/14 ———
0.4800 - val accuracy: 0.7000 - val loss: 0.6959
Epoch 32/100 10s 469ms/step - accuracy: 0.7402 - loss:
0.5133 - val_accuracy: 0.7091 - val_loss: 0.6098
Epoch 33/100 9s 383ms/step - accuracy: 0.7960 - loss:
0.4222 - val accuracy: 0.7091 - val loss: 0.6435
Epoch 34/100 10s 390ms/step - accuracy: 0.7574 - loss:
0.4826 - val accuracy: 0.7091 - val loss: 0.5740
Epoch 35/100 10s 464ms/step - accuracy: 0.7591 - loss:
0.4401 - val accuracy: 0.7000 - val_loss: 0.5816
Epoch 36/100
                ———— 10s 476ms/step - accuracy: 0.8164 - loss:
14/14 ———
0.3964 - val_accuracy: 0.7182 - val_loss: 0.5944
Epoch 37/100
                _____ 10s 466ms/step - accuracy: 0.8426 - loss:
14/14 ——
0.3614 - val_accuracy: 0.7182 - val loss: 0.6599
Epoch 38/100 11s 466ms/step - accuracy: 0.8115 - loss:
0.3867 - val accuracy: 0.7364 - val loss: 0.5440
Epoch 39/100 20s 451ms/step - accuracy: 0.8069 - loss:
0.3866 - val accuracy: 0.7091 - val loss: 0.6049
Epoch 40/100 14/14 10s 459ms/step - accuracy: 0.8401 - loss:
0.3491 - val accuracy: 0.7364 - val loss: 0.5851
Epoch 41/100 10s 411ms/step - accuracy: 0.8263 - loss:
0.3652 - val accuracy: 0.7000 - val loss: 0.7032
Epoch 42/100
               11s 470ms/step - accuracy: 0.8088 - loss:
0.3877 - val accuracy: 0.6909 - val loss: 0.6360
Epoch 43/100
```

```
0.3925 - val accuracy: 0.6364 - val loss: 0.6618
Epoch 44/100
                ———— 10s 398ms/step - accuracy: 0.8420 - loss:
14/14 ----
0.3571 - val accuracy: 0.7000 - val loss: 0.7266
Epoch 45/100 10s 465ms/step - accuracy: 0.8163 - loss:
0.4058 - val accuracy: 0.7000 - val loss: 0.6249
Epoch 46/100 14/14 10s 471ms/step - accuracy: 0.8088 - loss:
0.3633 - val accuracy: 0.6545 - val loss: 0.6302
Epoch 47/100
              10s 482ms/step - accuracy: 0.8277 - loss:
14/14 ———
0.3274 - val accuracy: 0.7455 - val loss: 0.5782
Epoch 48/100
              _____ 11s 464ms/step - accuracy: 0.8167 - loss:
14/14 -----
0.3754 - val_accuracy: 0.7182 - val_loss: 0.6916
Epoch 49/100
                 20s 473ms/step - accuracy: 0.8576 - loss:
0.2937 - val accuracy: 0.7364 - val loss: 0.5516
Epoch 50/100
                ———— 10s 451ms/step - accuracy: 0.8399 - loss:
14/14 ----
0.3145 - val accuracy: 0.7273 - val loss: 0.6361
Epoch 51/100 11s 464ms/step - accuracy: 0.8430 - loss:
0.3096 - val accuracy: 0.7364 - val loss: 0.5407
Epoch 52/100 9s 396ms/step - accuracy: 0.8912 - loss:
0.2648 - val accuracy: 0.6545 - val loss: 0.7377
Epoch 53/100 14/14 10s 391ms/step - accuracy: 0.9002 - loss:
0.2276 - val accuracy: 0.6818 - val loss: 0.6872
Epoch 54/100
              _____ 10s 487ms/step - accuracy: 0.8987 - loss:
14/14 ———
0.2280 - val accuracy: 0.6091 - val loss: 0.8626
Epoch 55/100
                ———— 10s 493ms/step - accuracy: 0.8979 - loss:
0.2622 - val accuracy: 0.6364 - val loss: 0.6914
Epoch 56/100
               ______ 20s 401ms/step - accuracy: 0.8418 - loss:
14/14 —
0.3104 - val accuracy: 0.6727 - val loss: 0.8149
0.2193 - val accuracy: 0.7091 - val loss: 0.7994
0.2505 - val accuracy: 0.7000 - val loss: 0.9132
Epoch 59/100
14/14 -
           _____ 10s 431ms/step - accuracy: 0.9034 - loss:
```

```
0.2579 - val accuracy: 0.6727 - val loss: 0.8235
Epoch 60/100
                 _____ 10s 389ms/step - accuracy: 0.8699 - loss:
14/14 ———
0.3120 - val accuracy: 0.6727 - val loss: 0.7928
Epoch 61/100
                 ———— 11s 467ms/step - accuracy: 0.8431 - loss:
0.3233 - val accuracy: 0.7091 - val loss: 0.6546
Epoch 62/100
                   ——— 10s 438ms/step - accuracy: 0.8925 - loss:
14/14 ---
0.2472 - val accuracy: 0.7364 - val loss: 0.6831
Epoch 63/100 9s 443ms/step - accuracy: 0.8923 - loss:
0.2463 - val accuracy: 0.7091 - val loss: 0.7355
Epoch 64/100 9s 382ms/step - accuracy: 0.8904 - loss:
0.2312 - val accuracy: 0.6818 - val loss: 0.8603
Epoch 65/100 14/14 10s 385ms/step - accuracy: 0.8989 - loss:
0.2131 - val accuracy: 0.6455 - val loss: 0.9930
Epoch 66/100
             ______ 10s 429ms/step - accuracy: 0.9168 - loss:
14/14 -----
0.2007 - val accuracy: 0.7182 - val loss: 0.7831
Epoch 67/100
                  ——— 9s 436ms/step - accuracy: 0.9332 - loss:
0.1904 - val accuracy: 0.7455 - val loss: 0.8285
Epoch 68/100
                 _____ 10s 399ms/step - accuracy: 0.9300 - loss:
14/14 –
0.1647 - val accuracy: 0.7091 - val loss: 0.9757
Epoch 69/100 9s 388ms/step - accuracy: 0.9560 - loss:
0.1426 - val accuracy: 0.6727 - val loss: 1.2775
Epoch 70/100 14/14 10s 452ms/step - accuracy: 0.8668 - loss:
0.3844 - val accuracy: 0.6727 - val loss: 1.1506
Epoch 71/100 14/14 11s 501ms/step - accuracy: 0.9322 - loss:
0.1896 - val accuracy: 0.6727 - val loss: 0.9762
Epoch 72/100
14/14
               ______ 10s 468ms/step - accuracy: 0.9326 - loss:
0.1565 - val accuracy: 0.6545 - val loss: 1.2432
Epoch 73/100
                  ——— 9s 388ms/step - accuracy: 0.9330 - loss:
14/14 ——
0.1983 - val_accuracy: 0.6636 - val_loss: 1.0931
Epoch 74/100
                    ——— 11s 403ms/step - accuracy: 0.9178 - loss:
14/14 —
0.1673 - val_accuracy: 0.7545 - val_loss: 0.7239
0.1549 - val accuracy: 0.7455 - val loss: 0.8322
```

```
0.1094 - val accuracy: 0.7364 - val loss: 0.7794
Epoch 77/100

9s 399ms/step - accuracy: 0.9386 - loss:
0.1522 - val accuracy: 0.7364 - val loss: 1.2355
Epoch 78/100
             ______ 11s 473ms/step - accuracy: 0.9571 - loss:
14/14 ———
0.1296 - val accuracy: 0.7182 - val loss: 1.0408
Epoch 79/100
               11s 468ms/step - accuracy: 0.9227 - loss:
14/14 -----
0.1696 - val_accuracy: 0.6273 - val_loss: 1.0524
Epoch 80/100
                _____ 20s 477ms/step - accuracy: 0.9211 - loss:
14/14 ——
0.1734 - val accuracy: 0.7727 - val loss: 0.7476
Epoch 81/100 9s 455ms/step - accuracy: 0.9594 - loss:
0.1346 - val_accuracy: 0.7273 - val_loss: 1.0627
Epoch 82/100 10/14 10s 394ms/step - accuracy: 0.9583 - loss:
0.1473 - val accuracy: 0.7364 - val loss: 0.8902
0.1096 - val accuracy: 0.7455 - val loss: 0.7104
0.1104 - val_accuracy: 0.7273 - val_loss: 1.0158
Epoch 85/100
               _____ 10s 473ms/step - accuracy: 0.9783 - loss:
14/14 ———
0.1066 - val_accuracy: 0.7000 - val_loss: 1.0136
Epoch 86/100
               9s 396ms/step - accuracy: 0.9608 - loss:
14/14 -
0.1001 - val_accuracy: 0.6636 - val_loss: 1.1873
Epoch 87/100 10s 392ms/step - accuracy: 0.9484 - loss:
0.1377 - val accuracy: 0.7273 - val loss: 1.0400
Epoch 88/100 10/14 10/14 10/162 10/162 10/162
0.0864 - val accuracy: 0.7636 - val loss: 1.0463
Epoch 89/100 14/14 10s 447ms/step - accuracy: 0.9609 - loss:
0.0923 - val accuracy: 0.7455 - val loss: 0.9022
Epoch 90/100 14/14 10s 417ms/step - accuracy: 0.9684 - loss:
0.0847 - val accuracy: 0.7455 - val loss: 1.1936
Epoch 91/100
              ______ 10s 393ms/step - accuracy: 0.9858 - loss:
0.0736 - val accuracy: 0.7273 - val loss: 1.2792
Epoch 92/100
```

```
———— 10s 403ms/step - accuracy: 0.9794 - loss:
0.0506 - val accuracy: 0.7182 - val loss: 1.4677
Epoch 93/100
                     ——— 11s 552ms/step - accuracy: 0.9863 - loss:
14/14 ---
0.0469 - val accuracy: 0.7364 - val loss: 1.2311
Epoch 94/100
               ______ 10s 467ms/step - accuracy: 0.9595 - loss:
14/14 ---
0.0816 - val accuracy: 0.7182 - val loss: 0.9116
Epoch 95/100 14/14 10s 492ms/step - accuracy: 0.9834 - loss:
0.0612 - val accuracy: 0.6818 - val loss: 1.0956
Epoch 96/100
                11s 469ms/step - accuracy: 0.9723 - loss:
14/14 -----
0.1259 - val accuracy: 0.7727 - val loss: 0.9579
Epoch 97/100
                 _____ 10s 383ms/step - accuracy: 0.9613 - loss:
14/14 ———
0.1324 - val accuracy: 0.7636 - val loss: 1.0059
Epoch 98/100
                     —— 10s 480ms/step - accuracy: 0.9687 - loss:
0.0725 - val accuracy: 0.7545 - val loss: 1.1011
Epoch 99/100
                   _____ 10s 454ms/step - accuracy: 0.9791 - loss:
14/14 -
0.0544 - val accuracy: 0.7091 - val loss: 1.1072
Epoch 100/100 9s 397ms/step - accuracy: 0.9871 - loss:
0.0443 - val accuracy: 0.7545 - val loss: 1.0947
model.evaluate(test set)
5/5 ——
              ______ 2s 473ms/step - accuracy: 0.6938 - loss:
1.8862
[2.0231473445892334, 0.6857143044471741]
# Function to load and preprocess the image
def load and prep image(img path, img size=(224, 224)):
   img = load img(img path, target size=img size) # Load and resize
   img = img to array(img) # Convert to array
   img = np.expand dims(img, axis=0) # Add batch dimension
    img = img / 255.0 # Normalize the image
    return img
# Function to plot the image and prediction
def plot image with prediction(img path, model):
   img = load and prep image(img path) # Preprocess the image
   pred = model.predict(img) # Predict the class
   class_names = ['Cat', 'Dog'] # Adjust class names as per your
dataset
   predicted class = np.argmax(pred, axis=1)[0] # Get the predicted
```

```
class (0 or 1)
    predicted_label = class_names[predicted_class] # Get the
corresponding label
    # Display the image and prediction
    img_to_display = load_img(img_path, target_size=(224, 224)) #
Load the image for display
    plt.imshow(img_to_display)
    plt.title(f"Predicted Label: {predicted_label}")
    plt.axis("off")
    plt.show()
# Example of using the function
# Provide the path to an image in your test dataset
test img path = "/content/test/dogs/dog 302.jpg"
plot_image_with_prediction(test_img_path, model)
1/1 -
                        - 0s 17ms/step
```

Predicted Label: Dog

