```
In [ ]: # If running in a Jupyter notebook, use this magic command to install TensorFlow
        # !pip install tensorflow
        # Step 1: Import necessary libraries
        import tensorflow as tf
        from tensorflow.keras.preprocessing.image import ImageDataGenerator
        from tensorflow.keras.preprocessing import image
        import numpy as np
        import os
        from sklearn.model selection import train test split
        import shutil
        # Step 2: Create train and validation directories within the current directory
        original_data_dir_pizza = 'pizza'
        original_data_dir_not_pizza = 'not_pizza'
        base dir = '.'
        train_dir = os.path.join(base_dir, 'train')
        validation_dir = os.path.join(base_dir, 'validation')
        train_pizza_dir = os.path.join(train_dir, 'pizza')
        validation pizza dir = os.path.join(validation dir, 'pizza')
        train_not_pizza_dir = os.path.join(train_dir, 'not_pizza')
        validation_not_pizza_dir = os.path.join(validation_dir, 'not_pizza')
        # Create directories
        os.makedirs(train_pizza_dir, exist_ok=True)
        os.makedirs(validation pizza dir, exist ok=True)
        os.makedirs(train_not_pizza_dir, exist_ok=True)
        os.makedirs(validation_not_pizza_dir, exist_ok=True)
        # Step 3: Split the data into training and validation sets
        pizza_images = os.listdir(original_data_dir_pizza)
        not pizza images = os.listdir(original data dir not pizza)
        train_pizza, val_pizza = train_test_split(pizza_images, test_size=0.2, random_state
        train_not_pizza, val_not_pizza = train_test_split(not_pizza_images, test_size=0.2,
        # Copy images to train and validation directories
        def copy_images(image_list, source_dir, target_dir):
            for image in image_list:
                shutil.copy(os.path.join(source_dir, image), target_dir)
        copy_images(train_pizza, original_data_dir_pizza, train_pizza_dir)
        copy_images(val_pizza, original_data_dir_pizza, validation_pizza_dir)
        copy_images(train_not_pizza, original_data_dir_not_pizza, train_not_pizza_dir)
        copy_images(val_not_pizza, original_data_dir_not_pizza, validation_not_pizza_dir)
        # Step 4: Data preprocessing
        # Image Data Generator for training set with augmentation
        train datagen = ImageDataGenerator(
            rescale=1./255,
            shear range=0.2,
            zoom_range=0.2,
```

```
horizontal_flip=True
)
# Image Data Generator for validation set without augmentation
validation_datagen = ImageDataGenerator(rescale=1./255)
# Training set
training_set = train_datagen.flow_from_directory(
   train dir,
   target_size=(64, 64),
   batch_size=32,
   class_mode='binary'
# Validation set
validation_set = validation_datagen.flow_from_directory(
   validation_dir,
   target_size=(64, 64),
   batch_size=32,
   class_mode='binary'
)
# Step 5: Build the CNN model
cnn = tf.keras.models.Sequential()
# Convolution
cnn.add(tf.keras.layers.Conv2D(filters=32, kernel_size=3, activation='relu', input_
# Pooling
cnn.add(tf.keras.layers.MaxPooling2D(pool_size=2, strides=2))
# Adding a second convolutional layer
cnn.add(tf.keras.layers.Conv2D(filters=32, kernel_size=3, activation='relu'))
cnn.add(tf.keras.layers.MaxPooling2D(pool_size=2, strides=2))
# Flattening
cnn.add(tf.keras.layers.Flatten())
# Full connection
cnn.add(tf.keras.layers.Dense(units=128, activation='relu'))
cnn.add(tf.keras.layers.Dense(units=1, activation='sigmoid'))
# Step 6: Compile the CNN
cnn.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])
# Step 7: Train the CNN
cnn.fit(x=training_set, validation_data=validation_set, epochs=25)
# Step 8: Make predictions on new images
def predict_image(image_path):
   test_image = image.load_img(image_path, target_size=(64, 64))
   test_image = image.img_to_array(test_image)
   test_image = np.expand_dims(test_image, axis=0)
   result = cnn.predict(test_image)
   if result[0][0] == 1:
        prediction = 'not pizza'
   else:
```

```
prediction = 'pizza'
return prediction

# Example usage
image1_path = 'some_image.jpg' # Replace with your image path
print(f'The image is {predict_image(image1_path)}')
```

Found 1572 images belonging to 2 classes. Found 394 images belonging to 2 classes.

C:\Users\joelr\AppData\Local\Packages\PythonSoftwareFoundation.Python.3.11_qbz5n2kfr a8p0\LocalCache\local-packages\Python311\site-packages\keras\src\layers\convolutiona 1\base_conv.py:107: UserWarning: Do not pass an `input_shape`/`input_dim` argument t o a layer. When using Sequential models, prefer using an `Input(shape)` object as th e first layer in the model instead.

super().__init__(activity_regularizer=activity_regularizer, **kwargs)
Epoch 1/25

C:\Users\joelr\AppData\Local\Packages\PythonSoftwareFoundation.Python.3.11_qbz5n2kfr a8p0\LocalCache\local-packages\Python311\site-packages\keras\src\trainers\data_adapt ers\py_dataset_adapter.py:121: UserWarning: Your `PyDataset` class should call `supe r().__init__(**kwargs)` in its constructor. `**kwargs` can include `workers`, `use_m ultiprocessing`, `max_queue_size`. Do not pass these arguments to `fit()`, as they w ill be ignored.

self._warn_if_super_not_called()

```
9s 142ms/step - accuracy: 0.5306 - loss: 0.7376 - val_acc
uracy: 0.5990 - val_loss: 0.6763
Epoch 2/25
50/50 -
                ------ 3s 46ms/step - accuracy: 0.6314 - loss: 0.6456 - val_accu
racy: 0.6853 - val_loss: 0.6245
Epoch 3/25
50/50 -----
                  ----- 3s 45ms/step - accuracy: 0.6782 - loss: 0.6084 - val_accu
racy: 0.6421 - val_loss: 0.6117
Epoch 4/25
50/50 -----
               ______ 3s 45ms/step - accuracy: 0.7345 - loss: 0.5622 - val_accu
racy: 0.6624 - val_loss: 0.6494
Epoch 5/25
50/50 -----
             racy: 0.6980 - val loss: 0.6219
Epoch 6/25
                     — 3s 48ms/step - accuracy: 0.7452 - loss: 0.5427 - val accu
racy: 0.7234 - val_loss: 0.5460
Epoch 7/25
                     — 3s 46ms/step - accuracy: 0.7533 - loss: 0.5042 - val accu
racy: 0.7716 - val_loss: 0.5299
Epoch 8/25
                   ---- 3s 46ms/step - accuracy: 0.7891 - loss: 0.4827 - val accu
50/50 -
racy: 0.7640 - val_loss: 0.5038
Epoch 9/25
50/50 -
                   ---- 3s 46ms/step - accuracy: 0.7701 - loss: 0.4811 - val_accu
racy: 0.7513 - val loss: 0.5170
racy: 0.7513 - val_loss: 0.5339
Epoch 11/25
                 ------ 3s 48ms/step - accuracy: 0.7979 - loss: 0.4368 - val accu
racy: 0.7589 - val loss: 0.5278
Epoch 12/25
                     — 3s 49ms/step - accuracy: 0.8116 - loss: 0.4298 - val accu
racy: 0.7462 - val_loss: 0.5161
Epoch 13/25
50/50 -
                  ----- 3s 48ms/step - accuracy: 0.8221 - loss: 0.4145 - val accu
racy: 0.7487 - val loss: 0.5243
Epoch 14/25
50/50 -
                 3s 49ms/step - accuracy: 0.8087 - loss: 0.4021 - val_accu
racy: 0.7716 - val_loss: 0.5313
Epoch 15/25
                   ---- 3s 49ms/step - accuracy: 0.8318 - loss: 0.3864 - val_accu
50/50 -----
racy: 0.7843 - val_loss: 0.5114
Epoch 16/25
               3s 49ms/step - accuracy: 0.8373 - loss: 0.3737 - val_accu
50/50 -----
racy: 0.7766 - val_loss: 0.5192
Epoch 17/25
                racy: 0.7766 - val loss: 0.5013
Epoch 18/25
               3s 48ms/step - accuracy: 0.8509 - loss: 0.3534 - val_accu
racy: 0.7716 - val_loss: 0.4879
Epoch 19/25
                     - 3s 49ms/step - accuracy: 0.8692 - loss: 0.3460 - val_accu
racy: 0.7792 - val loss: 0.4976
```

```
Epoch 20/25
            3s 47ms/step - accuracy: 0.8574 - loss: 0.3291 - val_accu
50/50 -----
racy: 0.7690 - val_loss: 0.5707
Epoch 21/25
50/50 -----
              3s 47ms/step - accuracy: 0.8533 - loss: 0.3310 - val_accu
racy: 0.7741 - val_loss: 0.6037
Epoch 22/25
               3s 48ms/step - accuracy: 0.8739 - loss: 0.3215 - val_accu
50/50 -----
racy: 0.7944 - val_loss: 0.5209
Epoch 23/25
                  3s 47ms/step - accuracy: 0.8584 - loss: 0.3266 - val_accu
50/50 ----
racy: 0.7817 - val_loss: 0.5011
Epoch 24/25
50/50 -----
                    --- 3s 48ms/step - accuracy: 0.8637 - loss: 0.3025 - val_accu
racy: 0.7843 - val_loss: 0.5194
Epoch 25/25
                3s 48ms/step - accuracy: 0.8673 - loss: 0.3059 - val_accu
50/50 -----
racy: 0.7843 - val_loss: 0.4945
1/1 0s 37ms/step
The image is not pizza
```

In []: