```
In [1]: import tensorflow as tf
         from tensorflow.keras.models import Sequential
         from tensorflow.keras.layers import Conv2D, MaxPooling2D, Flatten, Dense
         from tensorflow.keras.preprocessing.image import ImageDataGenerator
         import matplotlib.pyplot as plt
         import numpy as np
In [5]: # Paths to the directories
         train dir = 'train'
         val dir = 'val'
         test_dir ='test'
In [6]: # 1. Data Preprocessing and Augmentation
         train datagen = ImageDataGenerator(rescale=1./255, zoom range=0.2, horizontal flip=
         val datagen = ImageDataGenerator(rescale=1./255)
         test datagen = ImageDataGenerator(rescale=1./255)
In [7]: # Loading images from 'train' directory
         train generator = train datagen.flow from directory(
             train dir,
             target size=(150, 150), # Resize images to 150x150
             batch_size=32,
             class_mode='binary' # Binary classification (pneumonia vs normal)
       Found 5216 images belonging to 2 classes.
In [8]: # Loading images from 'val' directory
         val_generator = val_datagen.flow_from_directory(
             val dir,
             target size=(150, 150),
             batch_size=32,
             class_mode='binary'
       Found 16 images belonging to 2 classes.
In [9]: # Loading images from 'test' directory
         test_generator = test_datagen.flow_from_directory(
             test_dir,
             target_size=(150, 150),
             batch_size=32,
             class mode='binary'
       Found 624 images belonging to 2 classes.
In [10]: # 2. Build the CNN model
         model = Sequential([
             Conv2D(32, (3, 3), activation='relu', input shape=(150, 150, 3)),
             MaxPooling2D(pool_size=(2, 2)),
             Conv2D(64, (3, 3), activation='relu'),
             MaxPooling2D(pool_size=(2, 2)),
```

```
Conv2D(128, (3, 3), activation='relu'),
   MaxPooling2D(pool_size=(2, 2)),

Flatten(),
   Dense(128, activation='relu'),
   Dense(1, activation='sigmoid') # Binary classification (pneumonia vs normal)
])
```

C:\Users\bonde\anaconda3\Lib\site-packages\keras\src\layers\convolutional\base\_conv. py:107: UserWarning: Do not pass an `input\_shape`/`input\_dim` argument to a layer. W hen using Sequential models, prefer using an `Input(shape)` object as the first laye r in the model instead.

super().\_\_init\_\_(activity\_regularizer=activity\_regularizer, \*\*kwargs)

```
In [11]: # 3. Compile the model
model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])
```

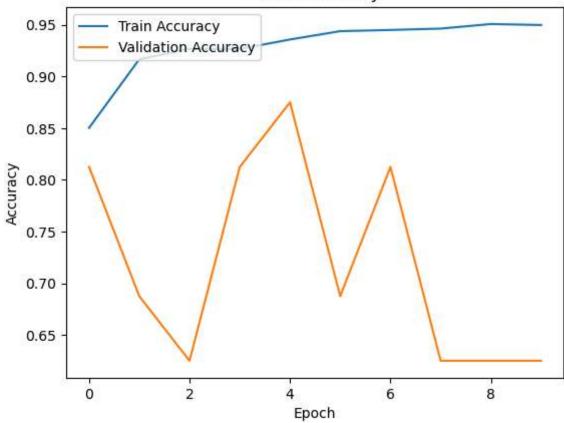
## Epoch 1/10

C:\Users\bonde\anaconda3\Lib\site-packages\keras\src\trainers\data\_adapters\py\_datas et\_adapter.py:122: UserWarning: Your `PyDataset` class should call `super().\_\_init\_\_ (\*\*kwargs)` in its constructor. `\*\*kwargs` can include `workers`, `use\_multiprocessi ng`, `max\_queue\_size`. Do not pass these arguments to `fit()`, as they will be ignor ed.

self.\_warn\_if\_super\_not\_called()

```
106s 612ms/step - accuracy: 0.7878 - loss: 0.4909 - val
       _accuracy: 0.8125 - val loss: 0.4107
       Epoch 2/10
                          97s 574ms/step - accuracy: 0.9130 - loss: 0.2123 - val_
       163/163 -
       accuracy: 0.6875 - val_loss: 0.7915
       Epoch 3/10
                       97s 574ms/step - accuracy: 0.9304 - loss: 0.1769 - val_
       163/163 —
       accuracy: 0.6250 - val_loss: 0.8974
       Epoch 4/10
                       97s 576ms/step - accuracy: 0.9242 - loss: 0.1762 - val
       163/163 -
       accuracy: 0.8125 - val_loss: 0.4453
       Epoch 5/10
       163/163 — 97s 573ms/step - accuracy: 0.9376 - loss: 0.1580 - val
       accuracy: 0.8750 - val loss: 0.5180
       Epoch 6/10
                          127s 759ms/step - accuracy: 0.9427 - loss: 0.1511 - val
       163/163 ---
       accuracy: 0.6875 - val loss: 0.5395
       Epoch 7/10
                              100s 593ms/step - accuracy: 0.9376 - loss: 0.1481 - val
       163/163 -
       _accuracy: 0.8125 - val_loss: 0.2721
       Epoch 8/10
       163/163 —
                           103s 613ms/step - accuracy: 0.9444 - loss: 0.1431 - val
       _accuracy: 0.6250 - val_loss: 1.5458
       Epoch 9/10
                          108s 646ms/step - accuracy: 0.9512 - loss: 0.1316 - val
       163/163 -
       accuracy: 0.6250 - val loss: 1.1715
       Epoch 10/10
                   127s 751ms/step - accuracy: 0.9491 - loss: 0.1308 - val
       163/163 ———
       _accuracy: 0.6250 - val_loss: 0.9418
In [13]: # 5. Evaluate the model on the test set
        test loss, test acc = model.evaluate(test generator)
         print(f"Test Accuracy: {test_acc:.3f}")
        # 6. Plot training & validation accuracy
         plt.plot(history.history['accuracy'], label='Train Accuracy')
         plt.plot(history.history['val accuracy'], label='Validation Accuracy')
         plt.title('Model accuracy')
         plt.ylabel('Accuracy')
         plt.xlabel('Epoch')
        plt.legend(loc='upper left')
        plt.show()
       20/20 -
                           8s 390ms/step - accuracy: 0.8440 - loss: 0.5190
       Test Accuracy: 0.864
```

## Model accuracy



```
In [14]: # Make predictions on the test data
predictions = model.predict(test_generator)

# Get the class labels for binary classification (NORMAL = 0, PNEUMONIA = 1)
class_labels = list(test_generator.class_indices.keys())
```

## **20/20 7s** 332ms/step

```
# Function to plot images with their predictions
In [17]:
         def show_predictions(generator, num_images=10):
             plt.figure(figsize=(15, 15))
             for i in range(num images):
                 # Get a batch of test images and labels
                 img_batch, label_batch = next(generator) # Use next() to get images and La
                 img = img_batch[0] # Get the first image in the batch
                 label = label_batch[0] # Get the first label in the batch
                 # Predict on the single image
                 prediction = model.predict(np.expand dims(img, axis=0)) # Make prediction
                 predicted_class = (prediction > 0.5).astype("int32")[0][0]
                 # Plot the image and show prediction
                 plt.subplot(5, 5, i + 1)
                 plt.imshow(img)
                 plt.title(f'Pred: {class_labels[predicted_class]}\nTrue: {class_labels[int(
                 plt.axis('off')
```

plt.show()

```
In [18]: # Show 10 predictions
             show_predictions(test_generator, num_images=10)
                                           - 0s 127ms/step
          1/1 -
                                           - 0s 34ms/step
          1/1
                                             0s 31ms/step
                                             0s 18ms/step
          1/1 -
          1/1 -
                                           • 0s 18ms/step
          1/1
                                             0s 15ms/step
          1/1 -
                                             0s 16ms/step
          1/1 -
                                           - 0s 47ms/step
          1/1
                                           • 0s 16ms/step
          1/1 -
                                           - 0s 16ms/step
             Pred: PNEUMONIA
True: PNEUMONIA
                                     Pred: PNEUMONIA
True: PNEUMONIA
                                                              Pred: NORMAL
True: NORMAL
                                                                                    Pred: PNEUMONIA
True: PNEUMONIA
                                                                                                            Pred: PNEUMONIA
True: PNEUMONIA
```

Pred: PNEUMONIA True: PNEUMONIA Pred: NORMAL True: NORMAL Pred: PNEUMONIA True: PNEUMONIA

In [ ]:

Pred: PNEUMONIA True: NORMAL Pred: PNEUMONIA True: PNEUMONIA