

# COSC 4337

## Keras\_First\_CNN

```
[1]: import tensorflow as tf
from tensorflow import keras
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Activation, Dense, Flatten,
↳BatchNormalization, Conv2D, MaxPool2D
from tensorflow.keras.optimizers import Adam
from tensorflow.keras.metrics import categorical_crossentropy
from sklearn.metrics import confusion_matrix
from tensorflow.keras.preprocessing.image import ImageDataGenerator
import numpy as np
import itertools
import os
import random
import matplotlib.pyplot as plt
%matplotlib inline
```

```
[2]: train_path = "chest_xray/train"
test_path = "chest_xray/test"
valid_path = "chest_xray/val"
```

```
[3]: # Creating train, test and valid batches from the respective directories
train_batches = ImageDataGenerator(preprocessing_function=tf.keras.applications.
↳vgg16.preprocess_input).flow_from_directory(directory=train_path,
↳target_size=(224,224), classes=['pneumonia', 'normal'], batch_size=10)
valid_batches = ImageDataGenerator(preprocessing_function=tf.keras.applications.
↳vgg16.preprocess_input).flow_from_directory(directory=valid_path,
↳target_size=(224,224), classes=['pneumonia', 'normal'], batch_size=10)
test_batches = ImageDataGenerator(preprocessing_function=tf.keras.applications.
↳vgg16.preprocess_input).flow_from_directory(directory=test_path,
↳target_size=(224,224), classes=['pneumonia', 'normal'], batch_size=10,
↳shuffle=False)
```

Found 5216 images belonging to 2 classes.

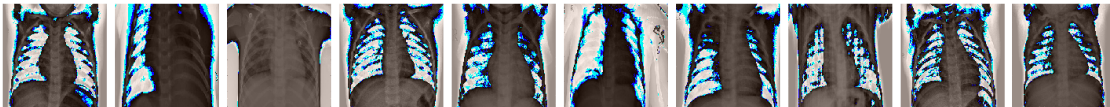
Found 16 images belonging to 2 classes.

Found 624 images belonging to 2 classes.

```
[4]: # plot images in the form of a 1 by 10 grid and resize img to 20x20
def plotImages(images_arr):
```

```
fig, axes = plt.subplots(1, 10, figsize=(20,20))
axes = axes.flatten()
for img, ax in zip( images_arr, axes):
    ax.imshow(img.astype(np.uint8))
    ax.axis('off')
plt.tight_layout()
plt.show()
```

```
[5]: imgs, labels = next(train_batches)
plotImages(imgs)
print(labels)
```



```
[[0. 1.]
 [1. 0.]
 [1. 0.]
 [0. 1.]
 [1. 0.]
 [1. 0.]
 [1. 0.]
 [1. 0.]
 [1. 0.]
 [0. 1.]
 [0. 1.]]
```

```
[6]: model = Sequential([
    Conv2D(filters=32, kernel_size=(3, 3), activation='relu', padding = 'same',
    ↪input_shape=(224,224,3)),
    MaxPool2D(pool_size=(2, 2), strides=2),
    Conv2D(filters=64, kernel_size=(3, 3), activation='relu', padding = 'same'),
    MaxPool2D(pool_size=(2, 2), strides=2),
    Flatten(),
    Dense(units=2, activation='softmax')
])
```

WARNING:tensorflow:From C:\Users\RizkN\.conda\envs\tf1\lib\site-packages\tensorflow\_core\python\ops\resource\_variable\_ops.py:1630: calling BaseResourceVariable.\_\_init\_\_ (from tensorflow.python.ops.resource\_variable\_ops) with constraint is deprecated and will be removed in a future version.  
Instructions for updating:  
If using Keras pass \*\_constraint arguments to layers.

```
[7]: model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
conv2d (Conv2D)	(None, 224, 224, 32)	896
max_pooling2d (MaxPooling2D)	(None, 112, 112, 32)	0
conv2d_1 (Conv2D)	(None, 112, 112, 64)	18496
max_pooling2d_1 (MaxPooling2D)	(None, 56, 56, 64)	0
flatten (Flatten)	(None, 200704)	0
dense (Dense)	(None, 2)	401410
Total params: 420,802		
Trainable params: 420,802		
Non-trainable params: 0		

```
[8]: model.compile(optimizer=Adam(learning_rate=0.0001),  
    loss='categorical_crossentropy', metrics=['accuracy'])
```

```
[9]: # recall that we set batch_size = 10 during preprocessing  
    batch_size = 10  
  
    model.fit(  
        x = train_batches,  
        steps_per_epoch=train_batches.samples // batch_size,  
        epochs=10,  
        validation_data=valid_batches,  
        validation_steps=valid_batches.samples // batch_size,  
        verbose=2)
```

```
Epoch 1/10  
Epoch 1/10  
521/521 - 244s - loss: 2.4432 - acc: 0.9291 - val_loss: 1.5285 - val_acc: 0.9000  
Epoch 2/10  
Epoch 1/10  
521/521 - 222s - loss: 0.2381 - acc: 0.9721 - val_loss: 0.6962 - val_acc: 0.9000  
Epoch 3/10  
Epoch 1/10  
521/521 - 226s - loss: 0.0602 - acc: 0.9889 - val_loss: 1.6920 - val_acc: 0.8000  
Epoch 4/10  
Epoch 1/10  
521/521 - 224s - loss: 0.0223 - acc: 0.9940 - val_loss: 0.1357 - val_acc: 0.9000  
Epoch 5/10
```

```

Epoch 1/10
521/521 - 229s - loss: 0.0042 - acc: 0.9983 - val_loss: 0.0070 - val_acc: 1.0000
Epoch 6/10
Epoch 1/10
521/521 - 224s - loss: 0.0038 - acc: 0.9987 - val_loss: 0.6665 - val_acc: 0.8000
Epoch 7/10
Epoch 1/10
521/521 - 223s - loss: 1.1300e-04 - acc: 1.0000 - val_loss: 0.6575 - val_acc:
0.8000
Epoch 8/10
Epoch 1/10
521/521 - 227s - loss: 2.1484e-05 - acc: 1.0000 - val_loss: 0.6330 - val_acc:
0.8000
Epoch 9/10
Epoch 1/10
521/521 - 220s - loss: 1.6232e-05 - acc: 1.0000 - val_loss: 0.6668 - val_acc:
0.8000
Epoch 10/10
Epoch 1/10
521/521 - 220s - loss: 1.2365e-05 - acc: 1.0000 - val_loss: 0.6070 - val_acc:
0.8000

```

```
[9]: <tensorflow.python.keras.callbacks.History at 0x1fdacb41208>
```

```
[10]: # making predictions
predictions = model.predict(x = test_batches, verbose=0)
```

```
[13]: def plot_confusion_matrix(cm, classes,
                                normalize=False,
                                title='Confusion matrix',
                                cmap=plt.cm.Blues):
    """
    This function prints and plots the confusion matrix.
    Normalization can be applied by setting `normalize=True`.
    """
    plt.imshow(cm, interpolation='nearest', cmap=cmap)
    plt.title(title)
    plt.colorbar()
    tick_marks = np.arange(len(classes))
    plt.xticks(tick_marks, classes, rotation=45)
    plt.yticks(tick_marks, classes)
    if normalize:
        cm = cm.astype('float') / cm.sum(axis=1)[:, np.newaxis]
        print("Normalized confusion matrix")
    else:
        print('Confusion matrix, without normalization')
    print(cm)
    thresh = cm.max() / 2.
```

```

for i, j in itertools.product(range(cm.shape[0]), range(cm.shape[1])):
    plt.text(j, i, cm[i, j],
             horizontalalignment="center",
             color="white" if cm[i, j] > thresh else "black")
plt.tight_layout()
plt.ylabel('True label')
plt.xlabel('Predicted label')

```

```

[14]: # confusion matrix
cm = confusion_matrix(y_true=test_batches.classes, y_pred=np.
    ↪argmax(predictions, axis=-1))
plot_confusion_matrix(cm, classes = ['Pneumonia', 'Normal'])

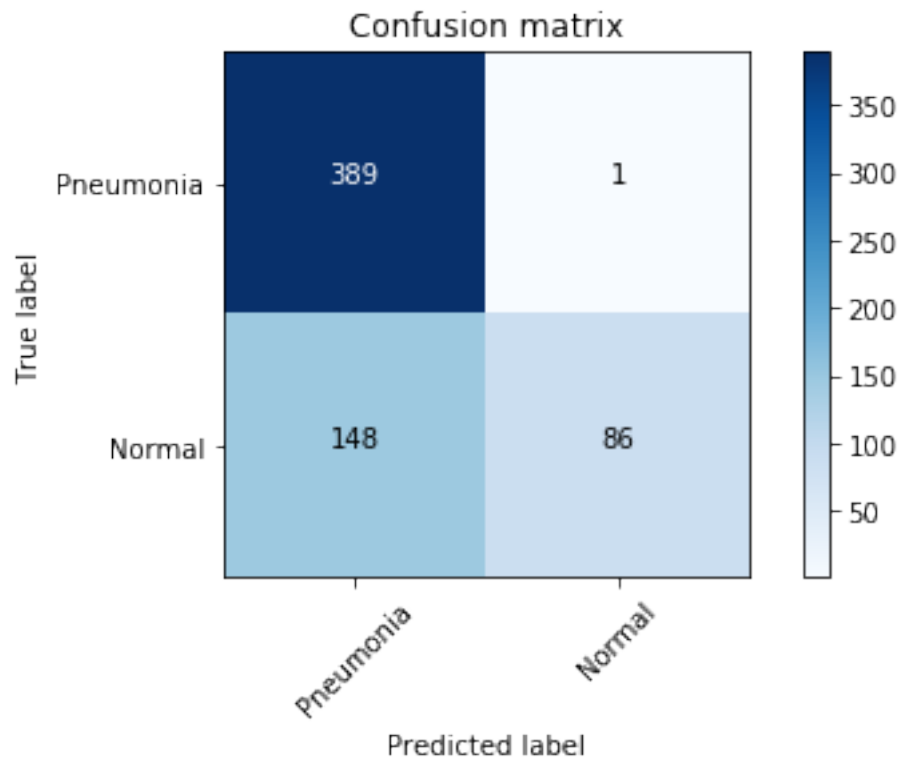
```

Confusion matrix, without normalization

```

[[389   1]
 [148  86]]

```



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[ ]:
```