

```
# This Python 3 environment comes with many helpful analytics libraries installed
# It is defined by the kaggle/python Docker image: https://github.com/kaggle/docker-python
# For example, here's several helpful packages to load

import numpy as np # linear algebra
import pandas as pd # data processing, CSV file I/O (e.g. pd.read_csv)

# Input data files are available in the read-only "../input/" directory
# For example, running this (by clicking run or pressing Shift+Enter) will list all files under the input directory

import os
for dirname, _, filenames in os.walk('/kaggle/input'):
    for filename in filenames:
        print(os.path.join(dirname, filename))

# You can write up to 20GB to the current directory (/kaggle/working/) that gets preserved as output when you creat
# You can also write temporary files to /kaggle/temp/, but they won't be saved outside of the current session
```


 [Show hidden output](#)

```
import numpy as np
import pandas as pd
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras.preprocessing.image import ImageDataGenerator, img_to_array, load_img
import matplotlib.pyplot as plt
import seaborn as sns
import glob
import cv2
from keras import Sequential
from keras.layers import Flatten,Dense,BatchNormalization,Dropout,LeakyReLU,GlobalAveragePooling2D
from keras.optimizers import Adam,RMSprop
from keras.losses import BinaryCrossentropy
from keras.applications import ResNet50V2
from keras.callbacks import EarlyStopping
```

```
train_path = "/kaggle/input/chest-xray-pneumonia/chest_xray/train/"
val_path = "/kaggle/input/chest-xray-pneumonia/chest_xray/val/"
test_path = "/kaggle/input/chest-xray-pneumonia/chest_xray/test/"
```

```
normal = glob.glob(train_path+"NORMAL/*.jpeg")
pneumonia = glob.glob(train_path+"PNEUMONIA/*.jpeg")
```

```
number_of_normal_img = len(normal)
number_of_pneumonia_img = len(pneumonia)
print("number of normal images:", number_of_normal_img)
print("number of pneumonia images:", number_of_pneumonia_img)
```

 number of normal images: 1341  
number of pneumonia images: 3875

```
train_val_generator = ImageDataGenerator(rescale=1./255,horizontal_flip=True,zoom_range=0.3)
test_generator = ImageDataGenerator(rescale=1./255)
```

```
train = train_val_generator.flow_from_directory(train_path,
                                                batch_size=128,
                                                target_size=(220,220),
                                                color_mode="rgb",
                                                class_mode="binary",
                                                shuffle=True,
```

```
seed=42,
subset="training")
```

➡ Found 5216 images belonging to 2 classes.

```
val = train_val_generator.flow_from_directory(val_path,
                                             batch_size=4,
                                             target_size=(220,220),
                                             color_mode="rgb",
                                             class_mode="binary",
                                             shuffle=True,
                                             seed=42)
```

➡ Found 16 images belonging to 2 classes.

```
test = test_generator.flow_from_directory(test_path,
                                          batch_size=32,
                                          target_size=(220,220),
                                          color_mode="rgb",
                                          class_mode="binary")
```

➡ Found 624 images belonging to 2 classes.

```
resnet50 = ResNet50V2(weights = "imagenet", input_shape = (220,220,3), include_top = False)
```

➡ Downloading data from [https://storage.googleapis.com/tensorflow/keras-applications/resnet/resnet50v2\\_weights\\_94668760/94668760](https://storage.googleapis.com/tensorflow/keras-applications/resnet/resnet50v2_weights_94668760/94668760) 0s 0us/step



```
for layer in resnet50.layers:
    layer.trainable = False
```

```
model = Sequential()
```

```
model.add(resnet50)
```

```
for layer in resnet50.layers:
    layer.trainable = False
```

```
model.add(Flatten())
```

```
model.add(Dense(units = 128, activation = "relu"))
model.add(Dropout(0.5))
```

```
model.add(Dense(units = 1, activation = "sigmoid"))
```

```
model.compile(optimizer = "adam", loss = "binary_crossentropy", metrics = ["accuracy"])
```

```
model.summary()
```

Model: "sequential"

Layer (type)	Output Shape	Param #
resnet50v2 (Functional)	(None, 7, 7, 2048)	23,564,800
flatten (Flatten)	(None, 100352)	0
dense (Dense)	(None, 128)	12,845,184
dropout (Dropout)	(None, 128)	0
dense_1 (Dense)	(None, 1)	129

Total params: 36,410,113 (138.89 MB)

Trainable params: 12,845,313 (49.00 MB)

Non-trainable params: 23,564,800 (89.89 MB)

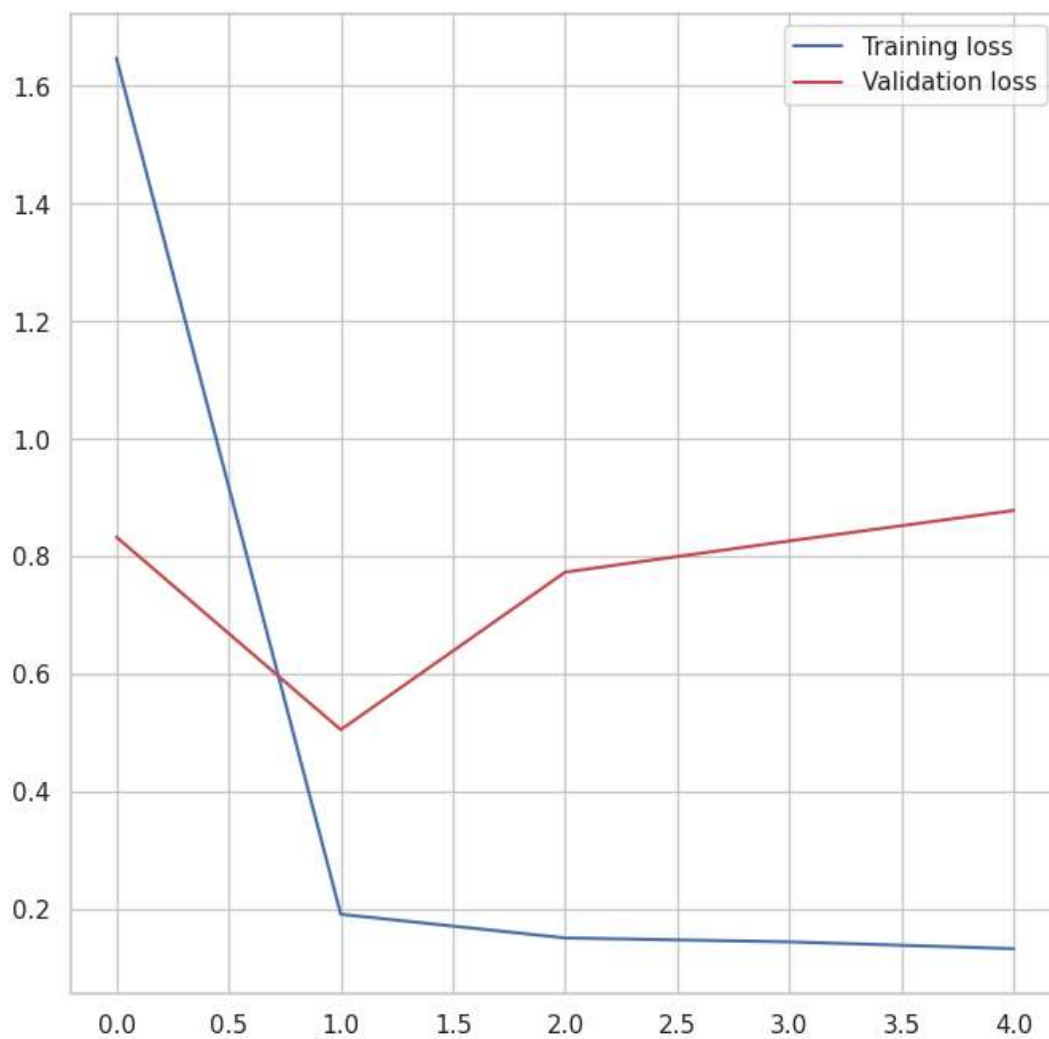
```
hist = model.fit(train, validation_data=val, epochs=5)
```

Epoch 1/5  
 /usr/local/lib/python3.10/dist-packages/keras/src/trainers/data\_adapters/py\_dataset\_adapter.py:122: UserWarning  
 self.\_warn\_if\_super\_not\_called()  
 41/41 ————— 559s 13s/step - accuracy: 0.7895 - loss: 3.6661 - val\_accuracy: 0.7500 - val\_loss:  
 Epoch 2/5  
 41/41 ————— 555s 13s/step - accuracy: 0.9219 - loss: 0.1998 - val\_accuracy: 0.7500 - val\_loss:  
 Epoch 3/5  
 41/41 ————— 545s 13s/step - accuracy: 0.9316 - loss: 0.1619 - val\_accuracy: 0.8750 - val\_loss:  
 Epoch 4/5  
 41/41 ————— 542s 13s/step - accuracy: 0.9452 - loss: 0.1447 - val\_accuracy: 0.7500 - val\_loss:  
 Epoch 5/5  
 41/41 ————— 544s 13s/step - accuracy: 0.9512 - loss: 0.1229 - val\_accuracy: 0.7500 - val\_loss:

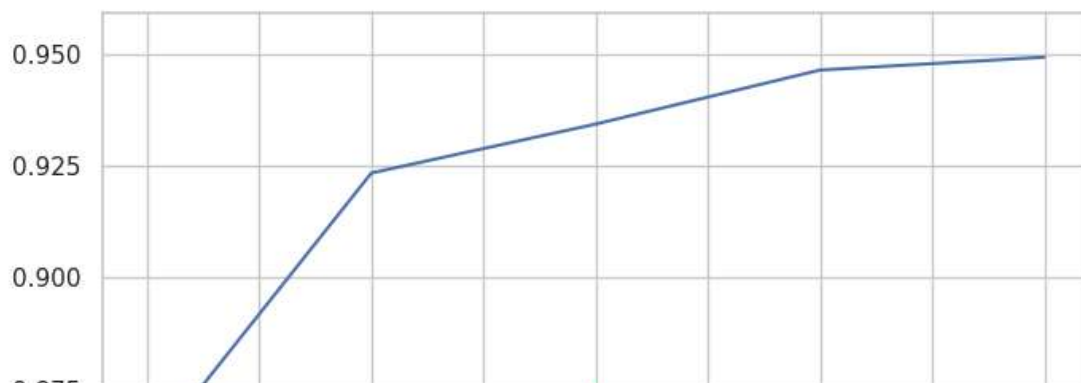
```
plt.figure(figsize=(8,8))
plt.plot(hist.history['loss'], color='b', label="Training loss")
plt.plot(hist.history['val_loss'], color='r', label="Validation loss")
plt.legend()
plt.show()

plt.figure()

plt.figure(figsize=(8,8))
plt.plot(hist.history['accuracy'], color='b', label="Training accuracy")
plt.plot(hist.history['val_accuracy'], color='r', label="Validation accuracy")
plt.legend(loc = "lower right")
plt.show()
```



<Figure size 640x480 with 0 Axes>



```
test_results = model.evaluate(test)
```



20/20 — 55s 3s/step - accuracy: 0.9380 - loss: 0.2407

