


```
inflating: chest_xray/___MACOSX/chest_xray/test/NORMAL/._IM-0046-0001.jpeg
inflating: chest_xray/___MACOSX/chest_xray/test/NORMAL/._IM-0049-0001.jpeg
inflating: chest_xray/___MACOSX/chest_xray/test/NORMAL/._IM-0050-0001.jpeg
inflating: chest_xray/___MACOSX/chest_xray/test/NORMAL/._IM-0059-0001.jpeg
inflating: chest_xray/___MACOSX/chest_xray/test/NORMAL/._IM-0061-0001.jpeg
inflating: chest_xray/___MACOSX/chest_xray/test/NORMAL/._IM-0063-0001.jpeg
inflating: chest_xray/___MACOSX/chest_xray/test/NORMAL/._IM-0065-0001.jpeg
inflating: chest_xray/___MACOSX/chest_xray/test/NORMAL/._IM-0067-0001.jpeg
inflating: chest_xray/___MACOSX/chest_xray/test/NORMAL/._IM-0069-0001.jpeg
inflating: chest_xray/___MACOSX/chest_xray/test/NORMAL/._IM-0070-0001.jpeg
inflating: chest_xray/___MACOSX/chest_xray/test/NORMAL/._IM-0071-0001.jpeg
inflating: chest_xray/___MACOSX/chest_xray/test/NORMAL/._IM-0073-0001.jpeg
inflating: chest_xray/___MACOSX/chest_xray/test/NORMAL/._IM-0075-0001.jpeg
inflating: chest_xray/___MACOSX/chest_xray/test/NORMAL/._IM-0077-0001.jpeg
inflating: chest_xray/___MACOSX/chest_xray/test/NORMAL/._IM-0079-0001.jpeg
inflating: chest_xray/___MACOSX/chest_xray/test/NORMAL/._IM-0081-0001.jpeg
inflating: chest_xray/___MACOSX/chest_xray/test/NORMAL/._IM-0083-0001.jpeg
```

```
import numpy as np
import tensorflow as tf
from tensorflow.keras.layers import Input, Conv2D, MaxPooling2D, UpSampling2D, Flatten,
from tensorflow.keras.models import Model, Sequential
from sklearn.model_selection import train_test_split
from tensorflow.keras.utils import to_categorical
from tensorflow import keras
```

```
img_width, img_height = 64, 64
batchsize = 64
num_of_class = 2
```

```
train = keras.utils.image_dataset_from_directory(
    directory='/Users/spikezaidspeigel/Desktop/dl lab/chest_xray/chest_xray/train',
    labels='inferred',
    label_mode='categorical',
    batch_size=batchsize,
    image_size=(img_width, img_height))
```

```
test = keras. utils.image_dataset_from_directory(
    directory='/Users/spikezaidspeigel/Desktop/dl lab/chest_xray/chest_xray/test',
    labels='inferred',
    label_mode='categorical',
    batch_size=batchsize,
    image_size=(img_width, img_height))
```

```
Found 5216 files belonging to 2 classes.
Found 624 files belonging to 2 classes.
```

```
x_train = []
y_train = []
x_test = []
y_test = []
```

```
for feature, label in train:
    x_train.append(feature.numpy())
    y_train.append(label.numpy())
```

```
for feature, label in test:
```

```

x_test.append(feature.numpy())
y_test.append(label.numpy())

x_train = np.concatenate(x_train, axis=0)
x_test = np.concatenate(x_test, axis=0)
y_train = np.concatenate(y_train, axis=0)
y_test = np.concatenate(y_test, axis=0)

x_train=x_train/256
x_test=x_test/256

print(x_train.shape)
print(x_test.shape)

(5216, 64, 64, 3)
(624, 64, 64, 3)

input_shape = img_width, img_height, 3
input_img = Input(shape=input_shape)
x = Conv2D(32, (8, 8), activation='relu', padding='same')(input_img)
x = MaxPooling2D((2, 2), padding='same')(x)

x = Conv2D(16, (8, 8), activation='relu', padding='same')(x)
x = MaxPooling2D((2, 2), padding='same')(x)

x = Conv2D(8, (8, 8), activation='relu', padding='same')(x)
encoded = MaxPooling2D((2, 2), padding='same')(x)

x = Conv2D(8, (8, 8), activation='relu', padding='same')(encoded)
x = UpSampling2D((2, 2))(x)

x = Conv2D(16, (8, 8), activation='relu', padding='same')(x)
x = UpSampling2D((2, 2))(x)

x = Conv2D(32, (8, 8), activation='relu', padding='same')(x)
x = UpSampling2D((2, 2))(x)

decoded = Conv2D(3, (8, 8), activation='sigmoid', padding='same')(x)

autoencoder = Model(input_img, decoded)
autoencoder.compile(optimizer='adam', loss='mean_squared_error')

```

NOTE for everyone: we're using x_test as both the input and the target for validation during training. This is because the autoencoder aims to reconstruct the input images, so during validation, you want to compare the reconstructed images to the original input images to calculate the loss.

```

autoencoder.fit(x_train, x_train, epochs=5, batch_size=64, shuffle=True, validation_data=

Epoch 1/5
82/82 [=====] - 273s 3s/step - loss: 0.0214 - val_loss: 0.
Epoch 2/5
82/82 [=====] - 263s 3s/step - loss: 0.0102 - val_loss: 0.
Epoch 3/5

```

```
82/82 [=====] - 241s 3s/step - loss: 0.0079 - val_loss: 0.  
Epoch 4/5  
82/82 [=====] - 280s 3s/step - loss: 0.0068 - val_loss: 0.  
Epoch 5/5  
82/82 [=====] - 247s 3s/step - loss: 0.0063 - val_loss: 0.  
<keras.callbacks.History at 0x7f8ead18c5b0>
```

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