

max_pooling2d_2 (MaxPooling2D (None, 26, 26, 128))	0
flatten (Flatten)	0
dense (Dense)	(None, 86528)11075712
dense_1 (Dense)	(None, 2)258

Total params: 11,261,442
Trainable params: 11,261,442
Non-trainable params: 0

```
In [23]: print(np.size(x))
print(np.shape(x))
print(np.size(y))
print(np.shape(y))

29202432
(194, 224, 224, 3)
38
(194, 2)
```

```
In [24]: #this stores the fitting information
#put in our own data for the images and labelling stuff

history = model.fit(x, y, epochs = 10, validation_data = (xt, yt), callbacks = [es])

#dont forget to keep training until your accuracy becomes bad

Epoch 1/10
7/7 ===== - 19s 3s/step - loss: 0.7603 - accuracy: 0.5463 -
val_loss: 0.6945 - val_accuracy: 0.5000
Epoch 2/10
7/7 ===== - 13s 2s/step - loss: 0.7005 - accuracy: 0.5379 -
val_loss: 0.6847 - val_accuracy: 0.5000
Epoch 3/10
7/7 ===== - 14s 2s/step - loss: 0.6844 - accuracy: 0.5163 -
val_loss: 0.6784 - val_accuracy: 0.5909
Epoch 4/10
7/7 ===== - 14s 2s/step - loss: 0.6737 - accuracy: 0.6050 -
val_loss: 0.6769 - val_accuracy: 0.6364
Epoch 5/10
7/7 ===== - 14s 2s/step - loss: 0.6700 - accuracy: 0.6041 -
val_loss: 0.6708 - val_accuracy: 0.6364
Epoch 6/10
7/7 ===== - 13s 2s/step - loss: 0.6632 - accuracy: 0.6111 -
val_loss: 0.6727 - val_accuracy: 0.5909
Epoch 7/10
7/7 ===== - 14s 2s/step - loss: 0.6718 - accuracy: 0.6098 -
val_loss: 0.6675 - val_accuracy: 0.7273
Epoch 8/10
7/7 ===== - 13s 2s/step - loss: 0.6511 - accuracy: 0.6498 -
val_loss: 0.6688 - val_accuracy: 0.6818
Epoch 9/10
7/7 ===== - 13s 2s/step - loss: 0.6641 - accuracy: 0.5980 -
val_loss: 0.6722 - val_accuracy: 0.7273
Epoch 10/10
7/7 ===== - 14s 2s/step - loss: 0.6652 - accuracy: 0.5865 -
val_loss: 0.6655 - val_accuracy: 0.6364
```

```
In [25]: # Plot the change in accuracy and validation accuracy as a function of epochs

plt.plot(history.history['accuracy'], label='accuracy')
plt.plot(history.history['val_accuracy'], label='val_accuracy')
plt.xlabel('Epoch')
plt.ylabel('Accuracy')
plt.ylim(0, 1)
plt.legend(loc='lower right')
```



```
In [26]: performance = model.evaluate(X_test, y_test, verbose=1)

2/2 [=====] - 1s 453ms/step - loss: 0.6503 - accuracy: 0.6667
```

```
In [27]: y_pred = model.predict(X_test)
print(y_pred)

[[0.5788577 0.42114228]
 [0.97771764 0.02228238]
 [0.59596943 0.4049305 ]
 [0.50533015 0.49466982]
 [0.68056893 0.31943107]
 [0.5151514 0.48484862]
 [0.5313143 0.46868566]
 [0.5827789 0.4172211 ]
 [0.58487195 0.4151281 ]
 [0.5960251 0.4035748 ]
 [0.60610163 0.3938983 ]
 [0.5258783 0.47412172]
 [0.536712 0.46328804]
 [0.5292513 0.4707487 ]
 [0.5387855 0.46121445]
 [0.5465227 0.45347735]
 [0.33974645 0.6602535 ]
 [0.5355231 0.46447688]
 [0.58430686 0.41569328]
 [0.57782865 0.42217132]
 [0.7352031 0.26479688]
 [0.6021158 0.39788422]
 [0.57931334 0.4208667 ]
 [0.49606153 0.50393844]
 [0.58762383 0.4123762 ]
 [0.6887743 0.31122577]
 [0.54571277 0.4542873 ]
 [0.45948133 0.54051864]
 [0.6057914 0.39420864]
 [0.4465654 0.55343455]
 [0.60244924 0.39755073]
 [0.57046854 0.4295314 ]
 [0.6301627 0.3698373 ]
 [0.4920884 0.5079116 ]
 [0.4035052 0.5964948 ]
 [0.56033206 0.43966797]
 [0.5877249 0.41227505]
 [0.56644195 0.43355805]
 [0.40571567 0.59428436]
 [0.48935604 0.5106439 ]
 [0.52726674 0.47273326]
 [0.3934633 0.60653675]
 [0.48530552 0.5146594 ]
 [0.4542345 0.5457655 ]
 [0.60464644 0.39535362]
 [0.5187622 0.48121766]
 [0.5898149 0.41018507]
 [0.6575892 0.34241083]
 [0.5853153 0.41468465]
 [0.5554536 0.44454637]
 [0.59884906 0.40115097]
 [0.5957013 0.40429872]
 [0.5611944 0.43880558]
 [0.6112777 0.3887223 ]]
```