

In [1]:

```
import cv2
import matplotlib.pyplot as plt
import numpy
from matplotlib import pyplot
from keras.models import Sequential
from keras.layers import Conv2D
from keras.layers import MaxPooling2D
from keras.layers import Dense
from keras.layers import Flatten
from keras.optimizers import SGD
from keras.preprocessing.image import ImageDataGenerator
import warnings
warnings.filterwarnings('ignore')
```

In [2]:

```
model = Sequential()
```

In [3]:

```
#add first convolution layer with the input shape
model.add(Conv2D(32, (3, 3),
                 activation='relu',
                 kernel_initializer='he_uniform',
padding='same', input_shape=(150, 150, 3)))
#add a Maxpooling layer
model.add(MaxPooling2D((2, 2)))
#add another convolution layer with more hidden units
model.add(Conv2D(64, (3, 3), activation='relu', kernel_initializer='he_uniform',
padding='same'))
model.add(MaxPooling2D((2, 2)))
#add another convolution layer with more hidden units
model.add(Conv2D(128, (3, 3), activation='relu', kernel_initializer='he_uniform',
padding='same'))
model.add(MaxPooling2D((2, 2)))
```

In [4]:

```
model.add(Flatten())
model.add(Dense(128,activation='relu',kernel_initializer='he_uniform'))
model.add(Dense(1, activation='sigmoid'))
```

In [5]:

```
model.compile(optimizer=SGD(lr=0.01),loss='binary_crossentropy', metrics=['accuracy'])
```

In [6]:

```
# create data generators
train_datagen=ImageDataGenerator(rescale=1.0/255.0,
width_shift_range=0.1,height_shift_range=0.1, horizontal_flip=True)
test_datagen = ImageDataGenerator(rescale=1.0/255.0)
# prepare iterators
train_it=train_datagen.flow_from_directory('training_set',class_mode='binary', batch_size=64)
test_it=test_datagen.flow_from_directory('test_set',class_mode='binary', batch_size=64, tar
```

Found 160 images belonging to 2 classes.

Found 33 images belonging to 2 classes.

In [7]:

```
history=model.fit_generator(train_it,steps_per_epoch=len(train_it),validation_data=test_it,
```

Epoch 1/20

3/3 [=====] - 3s 805ms/step - loss: 1.8597 - accuracy: 0.5188 - val_loss: 0.7426 - val_accuracy: 0.4848

Epoch 2/20

3/3 [=====] - 2s 642ms/step - loss: 0.7058 - accuracy: 0.5625 - val_loss: 0.6942 - val_accuracy: 0.4848

Epoch 3/20

3/3 [=====] - 2s 921ms/step - loss: 0.6862 - accuracy: 0.5625 - val_loss: 0.6965 - val_accuracy: 0.4848

Epoch 4/20

3/3 [=====] - 2s 903ms/step - loss: 0.6854 - accuracy: 0.6250 - val_loss: 0.6875 - val_accuracy: 0.4848

Epoch 5/20

3/3 [=====] - 2s 919ms/step - loss: 0.6724 - accuracy: 0.5625 - val_loss: 0.6758 - val_accuracy: 0.5455

Epoch 6/20

3/3 [=====] - 2s 655ms/step - loss: 0.6659 - accuracy: 0.6313 - val_loss: 0.7124 - val_accuracy: 0.4848

Epoch 7/20

3/3 [=====] - 2s 894ms/step - loss: 0.6579 - accuracy: 0.6687 - val_loss: 0.6696 - val_accuracy: 0.5455

Epoch 8/20

3/3 [=====] - 3s 992ms/step - loss: 0.6453 - accuracy: 0.6500 - val_loss: 0.6547 - val_accuracy: 0.6061

Epoch 9/20

3/3 [=====] - 2s 936ms/step - loss: 0.6770 - accuracy: 0.6938 - val_loss: 0.6573 - val_accuracy: 0.5455

Epoch 10/20

3/3 [=====] - 2s 681ms/step - loss: 0.6335 - accuracy: 0.6938 - val_loss: 0.6500 - val_accuracy: 0.5758

Epoch 11/20

3/3 [=====] - 2s 657ms/step - loss: 0.6236 - accuracy: 0.7312 - val_loss: 0.6416 - val_accuracy: 0.6061

Epoch 12/20

3/3 [=====] - 2s 678ms/step - loss: 0.6337 - accuracy: 0.6812 - val_loss: 0.6733 - val_accuracy: 0.5152

Epoch 13/20

3/3 [=====] - 2s 899ms/step - loss: 0.6477 - accuracy: 0.6562 - val_loss: 0.6459 - val_accuracy: 0.6061

Epoch 14/20

3/3 [=====] - 2s 735ms/step - loss: 0.6375 - accuracy: 0.6562 - val_loss: 0.6125 - val_accuracy: 0.7879

Epoch 15/20

3/3 [=====] - 2s 955ms/step - loss: 0.6104 - accuracy: 0.6938 - val_loss: 0.6423 - val_accuracy: 0.5455

Epoch 16/20

3/3 [=====] - 2s 941ms/step - loss: 0.6238 - accuracy: 0.6500 - val_loss: 0.6347 - val_accuracy: 0.5455

Epoch 17/20

3/3 [=====] - 2s 917ms/step - loss: 0.6002 - accuracy: 0.7375 - val_loss: 0.7454 - val_accuracy: 0.4848

Epoch 18/20

3/3 [=====] - 2s 666ms/step - loss: 0.6028 - accuracy: 0.6625 - val_loss: 0.5894 - val_accuracy: 0.6667

Epoch 19/20

3/3 [=====] - 2s 699ms/step - loss: 0.6421 - accuracy:

cy: 0.6687 - val_loss: 0.6183 - val_accuracy: 0.6061

Epoch 20/20

3/3 [=====] - 2s 976ms/step - loss: 0.5956 - accuracy: 0.5152

cy: 0.7188 - val_loss: 0.6998 - val_accuracy: 0.5152

In [8]:

```
model.save('monkeypox-vs-chickenpox.h5')
```

In [9]:

```
_,acc=model.evaluate_generator(test_it, steps=len(test_it), verbose=0)
print('> %.3f' % (acc * 100.0))
```

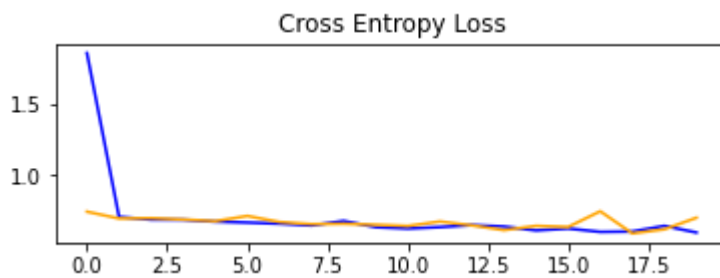
> 51.515

In [10]:

```
from matplotlib import pyplot
pyplot.subplot(211)
pyplot.title('Cross Entropy Loss')
pyplot.plot(history.history['loss'], color='blue', label='train')
pyplot.plot(history.history['val_loss'], color='orange', label='test')
```

Out[10]:

[<matplotlib.lines.Line2D at 0x1ed2f2ae560>]

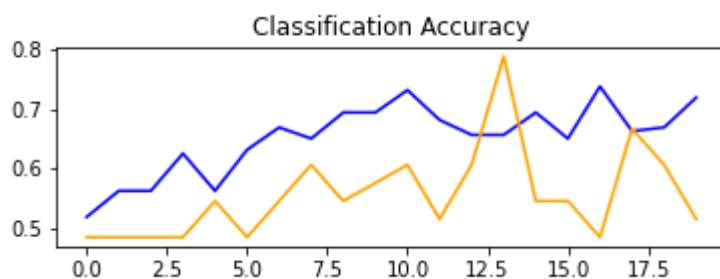


In [11]:

```
pyplot.subplot(212)
pyplot.title('Classification Accuracy')
pyplot.plot(history.history['accuracy'], color='blue', label='train')
pyplot.plot(history.history['val_accuracy'], color='orange', label='test')
```

Out[11]:

[<matplotlib.lines.Line2D at 0x1ed2f36cfa0>]



In [12]:

```
import sys
filename = sys.argv[0].split('/')[ -1]
pyplot.savefig(filename + '_plot.png')
pyplot.close()
```

In [13]:

```
import keras
import tensorflow as tf
from tensorflow.keras.utils import load_img, img_to_array
```

In [14]:

```
import numpy as np
from keras.preprocessing import image
test_image = keras.utils.load_img('image.png', target_size = (150, 150))
test_image = tf.keras.preprocessing.image.img_to_array(test_image)
test_image = np.expand_dims(test_image, axis = 0)
result = model.predict(test_image)
train_it.class_indices
if result[0][0] == 1:
    prediction = 'Monkeypox'
    print(prediction)
else:
    prediction = 'Chickenpox'
    print(prediction)
```

1/1 [=====] - 0s 65ms/step
Monkeypox