Artificial Intelligence Term Project (Horses or Humans)

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Dataset information:

- Horses or Humans is a dataset of 300×300 images, created by Laurence Moroney, that is licensed CC-By-2.0 for anybody to use in learning or testing computer vision algorithms.
- The set contains 500 rendered images of various species of horse in various poses in various locations. It also contains 527 rendered images of humans in various poses and locations
- number of traning images = 1027
- image size 300x300 width x height.
- Number of classes 2 labels.



Load Dataset:

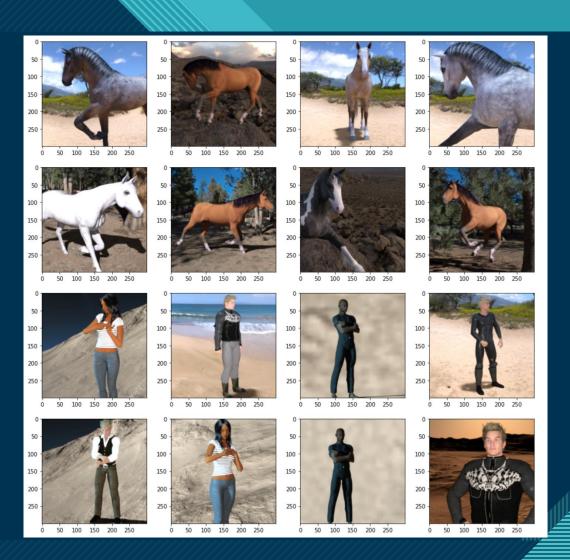
- Loading the dataset from google URL (zip file).
- Using OS library give us accses to the zipfile to unzip the file or data.

```
[43] !wget --no-check-certificate \
         https://storage.googleapis.com/laurencemoroney-blog.appspot.com/horse-or-human.zip \
         -0 /tmp/horse-or-human.zip
    --2020-12-04 23:15:04-- https://storage.googleapis.com/laurencemoroney-blog.appspot.com/horse-or-human.zip
    Resolving storage.googleapis.com (storage.googleapis.com)... 172.217.7.176, 142.250.73.208, 142.250.73.240, ...
    Connecting to storage.googleapis.com (storage.googleapis.com) | 172.217.7.176 | :443... connected.
    HTTP request sent, awaiting response... 200 OK
    Length: 149574867 (143M) [application/zip]
    Saving to: '/tmp/horse-or-human.zip'
    /tmp/horse-or-human 100%[===========] 142.65M 295MB/s
                                                                        in 0.5s
    2020-12-04 23:15:05 (295 MB/s) - '/tmp/horse-or-human.zip' saved [149574867/149574867]
[13] import os
     import zipfile
    local zip = '/tmp/horse-or-human.zip'
     zip ref = zipfile.ZipFile(local zip, 'r')
    zip ref.extractall('/tmp/horse-or-human')
     zip ref.close()
```



Display a batch of 8 horses and 8 humans pictures:

```
[28] %matplotlib inline
     import matplotlib.pyplot as plt
     import matplotlib.image as mpimg
     # Parameters for our graph; we'll output images in a 4x4 configuration
     ncols = 4
     # Index for iterating over images
     pic index = 0
     # Set up matplotlib fig, and size it to fit 4x4 pics
     fig = plt.gcf()
     fig.set size inches(ncols * 4, nrows * 4)
     pic index += 8
     next horse pix = [os.path.join(train horse dir, fname)
                     for fname in train horse names[pic index-8:pic index]]
     next human pix = [os.path.join(train human dir, fname)
                     for fname in train human names[pic index-8:pic index]]
     for i, img path in enumerate(next horse pix+next human pix):
       # Set up subplot; subplot indices start at 1
       sp = plt.subplot(nrows, ncols, i + 1)
       # Don't show axes (or gridlines)
       sp.axis('on')
       img = mpimg.imread(img path)
       plt.imshow(img)
     plt.show()
```





Model:

Building a Small Model from Scratch :1.import tensorflow.

2.add convolutional layers

```
import tensorflow as tf
model = tf.keras.models.Sequential([
    # Note the input shape is the desired size of the image 300x300 with 3 bytes color
    # This is the first convolution
    tf.keras.layers.Conv2D(16, (3,3), activation='relu', input shape=(300, 300, 3)),
    tf.keras.layers.MaxPooling2D(2, 2),
    # The second convolution
    tf.keras.layers.Conv2D(32, (3,3), activation='relu'),
    tf.keras.layers.MaxPooling2D(2,2),
    # The third convolution
    tf.keras.layers.Conv2D(64, (3,3), activation='relu'),
   tf.keras.layers.MaxPooling2D(2,2),
    # The fourth convolution
    tf.keras.layers.Conv2D(64, (3,3), activation='relu'),
    tf.keras.layers.MaxPooling2D(2,2),
    # The fifth convolution
    tf.keras.layers.Conv2D(64, (3,3), activation='relu'),
    tf.keras.layers.MaxPooling2D(2,2),
    # Flatten the results to feed into a DNN
    tf.keras.layers.Flatten(),
    # 512 neuron hidden layer
    tf.keras.layers.Dense(512, activation='relu'),
    # Only 1 output neuron. It will contain a value from 0-1 where 0 for 1 class ('horses') and 1 for the other ('humans'
    tf.keras.layers.Dense(1, activation='sigmoid')
```

The model.summary() method call prints a summary of the NN

```
model.summary()
Model: "sequential"
Layer (type)
                              Output Shape
                                                         Param #
                              (None, 298, 298, 16)
conv2d (Conv2D)
                                                         448
max pooling2d (MaxPooling2D) (None, 149, 149, 16)
conv2d 1 (Conv2D)
                              (None, 147, 147, 32)
                                                         4640
max pooling2d 1 (MaxPooling2 (None, 73, 73, 32)
conv2d 2 (Conv2D)
                              (None, 71, 71, 64)
                                                         18496
max pooling2d 2 (MaxPooling2 (None, 35, 35, 64)
conv2d 3 (Conv2D)
                              (None, 33, 33, 64)
                                                         36928
max_pooling2d_3 (MaxPooling2 (None, 16, 16, 64)
conv2d 4 (Conv2D)
                              (None, 14, 14, 64)
                                                         36928
max pooling2d 4 (MaxPooling2 (None, 7, 7, 64)
flatten (Flatten)
                              (None, 3136)
dense (Dense)
                              (None, 512)
                                                         1606144
dense 1 (Dense)
                                                         513
Total params: 1,704,097
Trainable params: 1,704,097
Non-trainable params: 0
```



Training(Epoch, Batch_size, Activation_function)

Note:

- Batch_size: 128
- Activation_function : ReLU
- ReLU stands for rectified linear unit, and is a type of activation function.
- We have 1027 images in the dataset and 2 classes

```
from tensorflow.keras.preprocessing.image import ImageDataGenerator
# All images will be rescaled by 1./255
train datagen = ImageDataGenerator(rescale=1/255)
# Flow training images in batches of 128 using train datagen generator
train generator = train datagen.flow from directory(
        '/tmp/horse-or-human/', # This is the source directory for training images
        target size=(300, 300),
        batch size=128,
        # Since we use binary crossentropy loss, we need binary labels
        class mode='binary')
```

Found 1027 images belonging to 2 classes.



Training(Epoch, Batch_size, Activation_function)

Note:

• Steps_per_Epoch = (1027 / 128) = 8

verbose = 1, which includes both progress bar and one line per epoch

```
history = model.fit(
    train generator,
    steps per epoch=8,
    epochs=15,
    verbose=1)
Epoch 1/15
8/8 [=========]
                            - 5s 677ms/step - loss: 0.7410 - accuracy: 0.5829
Epoch 2/15
8/8 [=========]
                            - 6s 764ms/step - loss: 0.7678 - accuracy: 0.6630
Epoch 3/15
8/8 [=========]
                            - 5s 674ms/step - loss: 0.3936 - accuracy: 0.8610
Epoch 4/15
8/8 [==========]
                             5s 671ms/step - loss: 0.2927 - accuracy: 0.8710
Epoch 5/15
6s 747ms/step - loss: 0.3728 - accuracy: 0.8574
Epoch 6/15
8/8 [==========]
                             6s 753ms/step - loss: 0.1532 - accuracy: 0.9492
Epoch 7/15
8/8 [==========]
                             5s 664ms/step - loss: 0.1652 - accuracy: 0.9388
Epoch 8/15
5s 670ms/step - loss: 0.0538 - accuracy: 0.9833
Epoch 9/15
5s 663ms/step - loss: 1.7270 - accuracy: 0.8287
Epoch 10/15
5s 665ms/step - loss: 0.2149 - accuracy: 0.9210
Epoch 11/15
                            - 5s 659ms/step - loss: 0.0695 - accuracy: 0.9711
Epoch 12/15
8/8 [=========]
                            - 6s 754ms/step - loss: 0.0241 - accuracy: 0.9944
Epoch 13/15
                            - 5s 652ms/step - loss: 0.0287 - accuracy: 0.9933
Epoch 14/15
8/8 [=============== ] - 5s 668ms/step - loss: 0.0245 - accuracy: 0.9889
Epoch 15/15
```



Result:

You can click on Choose
 File, after selecting the file
 the test will be performed
 and give the result

```
import numpy as np
from google.colab import files
from keras.preprocessing import image
uploaded = files.upload()
for fn in uploaded.keys():
  # predicting images
  path = '/content/' + fn
  img = image.load img(path, target size=(300, 300))
  x = image.img to array(img)
  x = np.expand dims(x, axis=0)
  images = np.vstack([x])
  classes = model.predict(images, batch size=10)
  print(classes[0])
  if classes[0]>0.5:
    print(fn + " is a human")
  else:
    print(fn + " is a horse")
```

Choose Files No file chosen

Cancel upload



Result:

After selecting an image from the horses file

```
import numpy as np
from google.colab import files
from keras.preprocessing import image
uploaded = files.upload()
for fn in uploaded.keys():
  # predicting images
  path = '/content/' + fn
  img = image.load img(path, target size=(300, 300))
  x = image.img to array(img)
  x = np.expand dims(x, axis=0)
  images = np.vstack([x])
  classes = model.predict(images, batch size=10)
  print(classes[0])
  if classes[0]>0.5:
    print(fn + " is a human")
    print(fn + " is a horse")
```

Choose Files horse01-0.png

horse01-0.png(image/png) - 151700 bytes, last modified: 10/22/2019 - 100% done
 Saving horse01-0.png to horse01-0.png
 [0.]
 horse01-0.png is a horse

After selecting an image from the human file

```
import numpy as np
from google.colab import files
from keras.preprocessing import image
uploaded = files.upload()
for fn in uploaded.keys():
  # predicting images
  path = '/content/' + fn
  img = image.load img(path, target size=(300, 300))
  x = image.img to array(img)
  x = np.expand dims(x, axis=0)
  images = np.vstack([x])
  classes = model.predict(images, batch size=10)
  print(classes[0])
  if classes[0]>0.5:
    print(fn + " is a human")
  else:
    print(fn + " is a horse")
```

Choose Files human01-00.png

human01-00.png(image/png) - 160725 bytes, last modified: 10/22/2019 - 100% done
Saving human01-00.png to human01-00.png
[1.]
human01-00.png is a human



Conclusion

• Finally In this project, we selected a dataset (Human_Horses) that is in Tenserflow, and built a neural network that uses more than one layer, and train the model and achieving high accuracy with minimum lost, and give the data for the training model and predicted the results.

3 Dec 2020