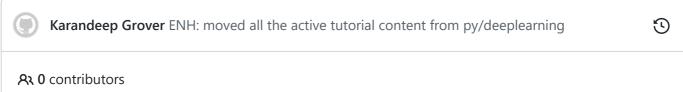


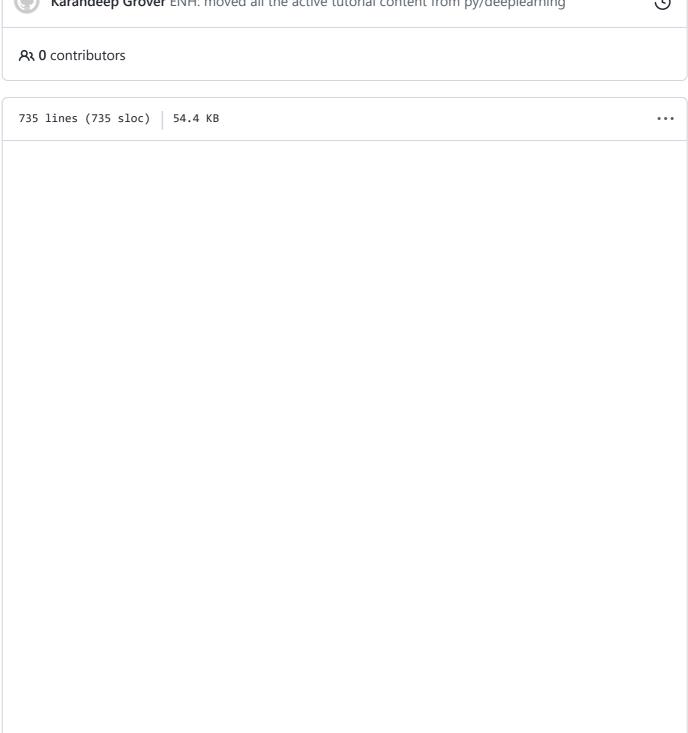
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gpu_performance_test_small_image_classification.ipynb





Small Image Classification Using Simple Aritifical Neural Network: GPU Benchmarking

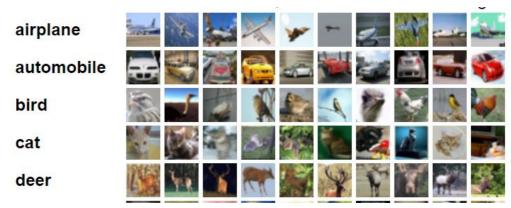
```
import tensorflow as tf
from tensorflow import keras
import matplotlib.pyplot as plt
import numpy as np
# Version Information
# tensorflow 2.2.0 , Cudnn7.6.5 and Cuda 10.1 , python 3.8
```

This command shows list of physical devices available for tensorflow. You can see GPU listed here. If you have NVIDIA GPU you need to install CUDA toolkit and cuDNN as per instruction on this webpage. Without proper installation you will not see GPU in list of devices

https://shawnhymel.com/1961/how-to-install-tensorflow-with-gpu-support-on-windows/

Load the dataset

Our dataset contains 60000 small training images that belongs to one of the below 10 classes



In [8]:

In [9]:

In [10]:



Data Visualization

Out[10]: (50000, 1)

y_train.shape

```
In [11]:
          def plot_sample(index):
              plt.figure(figsize = (10,1))
              plt.imshow(X_train[index])
```

```
In [12]:
          plot_sample(0)
```



In [13]: plot_sample(1)



In [14]: plot_sample(2)



In [15]: classes = ["airnlane"."automobile"."hird"."cat"."deer"."dog"."frog"."hor

```
In [16]:
          plot_sample(3)
In [17]:
          classes[y_train[3][0]]
Out[17]: 'deer'
In [18]:
          y_train[:3]
Out[18]: array([[6],
                 [9]], dtype=uint8)
In [19]:
          y_test.shape
Out[19]: (10000, 1)
In [20]:
          X_train.shape
Out[20]: (50000, 32, 32, 3)
         Preprocessing: Scale images
In [15]:
          X_train_scaled = X_train / 255
          X_test_scaled = X_test / 255
In [16]:
          y_train_categorical = keras.utils.to_categorical(
              y_train, num_classes=10, dtype='float32'
          y_test_categorical = keras.utils.to_categorical(
              y_test, num_classes=10, dtype='float32'
In [17]:
          y train[0:5]
Out[17]: array([[6],
                 [9],
                 [9],
                 [4],
                 [1]], dtype=uint8)
In [18]:
          y_train_categorical[0:5]
Out[18]: array([[0., 0., 0., 0., 0., 0., 1., 0., 0., 0.],
                 [0., 0., 0., 0., 0., 0., 0., 0., 0., 1.],
```

```
[0., 0., 0., 0., 0., 0., 0., 0., 0., 1.],
[0., 0., 0., 0., 1., 0., 0., 0., 0., 0.]
[0., 1., 0., 0., 0., 0., 0., 0., 0.]], dtype=float32)
```

Model building and training

```
In [19]:
          model = keras.Sequential([
                  keras.layers.Flatten(input_shape=(32,32,3)),
                  keras.layers.Dense(3000, activation='relu'),
                  keras.layers.Dense(1000, activation='relu'),
                  keras.layers.Dense(10, activation='sigmoid')
              1)
          model.compile(optimizer='SGD',
                        loss='categorical_crossentropy',
                        metrics=['accuracy'])
          model.fit(X_train_scaled, y_train_categorical, epochs=1)
         1563/1563 [=============== ] - 3s 2ms/step - loss: 1.8642 -
         accuracy: 0.3328
Out[19]:
```

Let's make some predictions

```
In [ ]:
          np.argmax(model.predict(X test scaled)[0])
In [ ]:
          y_test[0]
In [21]:
          def get_model():
              model = keras.Sequential([
                      keras.layers.Flatten(input_shape=(32,32,3)),
                      keras.layers.Dense(3000, activation='relu'),
                      keras.layers.Dense(1000, activation='relu'),
                      keras.layers.Dense(10, activation='sigmoid')
                  1)
              model.compile(optimizer='SGD',
                            loss='categorical crossentropy',
                            metrics=['accuracy'])
              return model
```

Measure training time on a CPU

```
In [22]:
         %%timeit -n1 -r1
         with tf.device('/CPU:0'):
             cpu_model = get_model()
             cpu_model.fit(X_train_scaled, y_train_categorical, epochs=1)
         1563/1563 [============== ] - 44s 28ms/step - loss: 1.8660
         - accuracy: 0.3301
         44.5 s \pm 0 ns per loop (mean \pm std. dev. of 1 run, 1 loop each)
```

Lets measure training time on a GPU (I've NVIDIA

Titan RTX)

```
In [23]:
        %%timeit -n1 -r1
        with tf.device('/GPU:0'):
           cpu model = get model()
           cpu_model.fit(X_train_scaled, y_train_categorical, epochs=1)
       1563/1563 [============== ] - 3s 2ms/step - loss: 1.8581 -
       accuracy: 0.3354
       3.6 s ± 0 ns per loop (mean ± std. dev. of 1 run, 1 loop each)
       Lets run same test for 10 epocs
In [25]:
       %%timeit -n1 -r1
        with tf.device('/CPU:0'):
           cpu_model = get_model()
           cpu_model.fit(X_train_scaled, y_train_categorical, epochs=10)
       Epoch 1/10
       - accuracy: 0.3328
       Epoch 2/10
       - accuracy: 0.4130
       Epoch 3/10
       1563/1563 [============== ] - 44s 28ms/step - loss: 1.5713
       - accuracy: 0.4486
       Epoch 4/10
       - accuracy: 0.4663
       Epoch 5/10
       1563/1563 [============== ] - 44s 28ms/step - loss: 1.4596
       - accuracy: 0.4837
       Epoch 6/10
       1563/1563 [============== ] - 44s 28ms/step - loss: 1.4157
       - accuracy: 0.5000
       Epoch 7/10
       1563/1563 [============== ] - 44s 28ms/step - loss: 1.3778
       - accuracy: 0.5155
       Epoch 8/10
       - accuracy: 0.5262
       Epoch 9/10
       1563/1563 [============== ] - 44s 28ms/step - loss: 1.3088
       - accuracy: 0.5392
       Epoch 10/10
       1563/1563 [============== ] - 44s 28ms/step - loss: 1.2786
       - accuracy: 0.5497
       7min 26s ± 0 ns per loop (mean ± std. dev. of 1 run, 1 loop each)
In [24]:
       %%timeit -n1 -r1
        with tf.device('/GPU:0'):
           cpu_model = get_model()
           cpu_model.fit(X_train_scaled, y_train_categorical, epochs=10)
       Epoch 1/10
       1563/1563 [============== ] - 3s 2ms/step - loss: 1.8624 -
       accuracy: 0.3322
       Epoch 2/10
```

```
accuracy: 0.4146
Epoch 3/10
accuracy: 0.4443
Epoch 4/10
1563/1563 [=============== ] - 3s 2ms/step - loss: 1.5068 -
accuracy: 0.4681
Epoch 5/10
accuracy: 0.4863
Epoch 6/10
1563/1563 [============== ] - 3s 2ms/step - loss: 1.4139 -
accuracy: 0.5009
Epoch 7/10
accuracy: 0.5147
```