

CS289A_HW01_MNIST

January 30, 2017

```
In [16]: %load_ext autoreload
```

The autoreload extension is already loaded. To reload it, use:
%reload_ext autoreload

```
In [17]: %autoreload 2
```

```
In [18]: import numpy as np
import HW01_utils as utils
import trainfunctions as tf
```

```
In [46]: _LOCAL_PATH = r"C:\Users\Mitch\Documents\Cal\2 - 2017 Spring\COMPSCI 289A
_DATA_PATH = "Data\hw01_data"

_DATA_DIR = _LOCAL_PATH + "\\\" + _DATA_PATH
trainpath = r"mnist\train.mat"
```

```
In [20]: valsetsize = 10000
samples = np.array([100, 200, 500, 1000, 2000, 5000, 10000])
hyperparams = np.logspace(-8,4,num=20)
```

```
In [21]: # Load MNIST training data
mnist = utils.loaddata(trainpath,_DATA_DIR,'trainX')

# Shuffle data before splitting
np.random.shuffle(mnist)
```

```
In [22]: trainset,valset = utils.partition(valsetsize,mnist)
trainsetarrays,trainsetlabels = utils.separatelabels(trainset)
valsetarrays,valsetlabels = utils.separatelabels(valset)
```

```
In [23]: Accs = np.empty((len(samples),len(hyperparams)))
i = 0 # sample index counter
for nsamples in samples:
    print(nsamples,'samples')
    j = 0 # hyperparameter index counter
    for hp in hyperparams:
        acc = tf.TrainAndScoreNsamples(trainsetarrays[:nsamples],trainsetlabels[:nsamples],hp)
```

```

print('\tC =',hp, '\tAccuracy:',acc)
Accs[i,j] = acc
j+=1
i+=1
print(Accs)

```

100 samples

```

C = 1e-08 Accuracy: 0.1119
C = 4.28133239872e-08 Accuracy: 0.2301
C = 1.83298071083e-07 Accuracy: 0.6438
C = 7.84759970351e-07 Accuracy: 0.7169
C = 3.35981828628e-06 Accuracy: 0.7133
C = 1.43844988829e-05 Accuracy: 0.7133
C = 6.15848211066e-05 Accuracy: 0.7133
C = 0.000263665089873 Accuracy: 0.7133
C = 0.00112883789168 Accuracy: 0.7133
C = 0.00483293023857 Accuracy: 0.7133
C = 0.0206913808111 Accuracy: 0.7133
C = 0.088586679041 Accuracy: 0.7133
C = 0.379269019073 Accuracy: 0.7133
C = 1.62377673919 Accuracy: 0.7133
C = 6.95192796178 Accuracy: 0.7133
C = 29.7635144163 Accuracy: 0.7133
C = 127.42749857 Accuracy: 0.7133
C = 545.559478117 Accuracy: 0.7133
C = 2335.72146909 Accuracy: 0.7133
C = 10000.0 Accuracy: 0.7133

```

200 samples

```

C = 1e-08 Accuracy: 0.0963
C = 4.28133239872e-08 Accuracy: 0.425
C = 1.83298071083e-07 Accuracy: 0.7747
C = 7.84759970351e-07 Accuracy: 0.8005
C = 3.35981828628e-06 Accuracy: 0.7947
C = 1.43844988829e-05 Accuracy: 0.7947
C = 6.15848211066e-05 Accuracy: 0.7947
C = 0.000263665089873 Accuracy: 0.7947
C = 0.00112883789168 Accuracy: 0.7947
C = 0.00483293023857 Accuracy: 0.7947
C = 0.0206913808111 Accuracy: 0.7947
C = 0.088586679041 Accuracy: 0.7947
C = 0.379269019073 Accuracy: 0.7947
C = 1.62377673919 Accuracy: 0.7947
C = 6.95192796178 Accuracy: 0.7947
C = 29.7635144163 Accuracy: 0.7947
C = 127.42749857 Accuracy: 0.7947
C = 545.559478117 Accuracy: 0.7947
C = 2335.72146909 Accuracy: 0.7947
C = 10000.0 Accuracy: 0.7947

```

500 samples

C = 1e-08	Accuracy: 0.3171
C = 4.28133239872e-08	Accuracy: 0.7527
C = 1.83298071083e-07	Accuracy: 0.865
C = 7.84759970351e-07	Accuracy: 0.8635
C = 3.35981828628e-06	Accuracy: 0.8552
C = 1.43844988829e-05	Accuracy: 0.8548
C = 6.15848211066e-05	Accuracy: 0.8548
C = 0.000263665089873	Accuracy: 0.8548
C = 0.00112883789168	Accuracy: 0.8548
C = 0.00483293023857	Accuracy: 0.8548
C = 0.0206913808111	Accuracy: 0.8548
C = 0.088586679041	Accuracy: 0.8548
C = 0.379269019073	Accuracy: 0.8548
C = 1.62377673919	Accuracy: 0.8548
C = 6.95192796178	Accuracy: 0.8548
C = 29.7635144163	Accuracy: 0.8548
C = 127.42749857	Accuracy: 0.8548
C = 545.559478117	Accuracy: 0.8548
C = 2335.72146909	Accuracy: 0.8548
C = 10000.0	Accuracy: 0.8548

1000 samples

C = 1e-08	Accuracy: 0.5782
C = 4.28133239872e-08	Accuracy: 0.85
C = 1.83298071083e-07	Accuracy: 0.8867
C = 7.84759970351e-07	Accuracy: 0.8909
C = 3.35981828628e-06	Accuracy: 0.8815
C = 1.43844988829e-05	Accuracy: 0.881
C = 6.15848211066e-05	Accuracy: 0.881
C = 0.000263665089873	Accuracy: 0.881
C = 0.00112883789168	Accuracy: 0.881
C = 0.00483293023857	Accuracy: 0.881
C = 0.0206913808111	Accuracy: 0.881
C = 0.088586679041	Accuracy: 0.881
C = 0.379269019073	Accuracy: 0.881
C = 1.62377673919	Accuracy: 0.881
C = 6.95192796178	Accuracy: 0.881
C = 29.7635144163	Accuracy: 0.881
C = 127.42749857	Accuracy: 0.881
C = 545.559478117	Accuracy: 0.881
C = 2335.72146909	Accuracy: 0.881
C = 10000.0	Accuracy: 0.881

2000 samples

C = 1e-08	Accuracy: 0.7981
C = 4.28133239872e-08	Accuracy: 0.8824
C = 1.83298071083e-07	Accuracy: 0.9043
C = 7.84759970351e-07	Accuracy: 0.9086
C = 3.35981828628e-06	Accuracy: 0.895

C = 1.43844988829e-05	Accuracy: 0.8926
C = 6.15848211066e-05	Accuracy: 0.8926
C = 0.000263665089873	Accuracy: 0.8926
C = 0.00112883789168	Accuracy: 0.8926
C = 0.00483293023857	Accuracy: 0.8926
C = 0.0206913808111	Accuracy: 0.8926
C = 0.088586679041	Accuracy: 0.8926
C = 0.379269019073	Accuracy: 0.8926
C = 1.62377673919	Accuracy: 0.8926
C = 6.95192796178	Accuracy: 0.8926
C = 29.7635144163	Accuracy: 0.8926
C = 127.42749857	Accuracy: 0.8926
C = 545.559478117	Accuracy: 0.8926
C = 2335.72146909	Accuracy: 0.8926
C = 10000.0	Accuracy: 0.8926

5000 samples

C = 1e-08	Accuracy: 0.8721
C = 4.28133239872e-08	Accuracy: 0.9038
C = 1.83298071083e-07	Accuracy: 0.9183
C = 7.84759970351e-07	Accuracy: 0.9202
C = 3.35981828628e-06	Accuracy: 0.9135
C = 1.43844988829e-05	Accuracy: 0.9041
C = 6.15848211066e-05	Accuracy: 0.903
C = 0.000263665089873	Accuracy: 0.903
C = 0.00112883789168	Accuracy: 0.903
C = 0.00483293023857	Accuracy: 0.903
C = 0.0206913808111	Accuracy: 0.903
C = 0.088586679041	Accuracy: 0.903
C = 0.379269019073	Accuracy: 0.903
C = 1.62377673919	Accuracy: 0.903
C = 6.95192796178	Accuracy: 0.903
C = 29.7635144163	Accuracy: 0.903
C = 127.42749857	Accuracy: 0.903
C = 545.559478117	Accuracy: 0.903
C = 2335.72146909	Accuracy: 0.903
C = 10000.0	Accuracy: 0.903

10000 samples

C = 1e-08	Accuracy: 0.8935
C = 4.28133239872e-08	Accuracy: 0.9147
C = 1.83298071083e-07	Accuracy: 0.9255
C = 7.84759970351e-07	Accuracy: 0.9298
C = 3.35981828628e-06	Accuracy: 0.9229
C = 1.43844988829e-05	Accuracy: 0.9118
C = 6.15848211066e-05	Accuracy: 0.9071
C = 0.000263665089873	Accuracy: 0.906
C = 0.00112883789168	Accuracy: 0.906
C = 0.00483293023857	Accuracy: 0.906
C = 0.0206913808111	Accuracy: 0.906

```

C = 0.088586679041          Accuracy: 0.906
C = 0.379269019073          Accuracy: 0.906
C = 1.62377673919           Accuracy: 0.906
C = 6.95192796178           Accuracy: 0.906
C = 29.7635144163           Accuracy: 0.906
C = 127.42749857            Accuracy: 0.906
C = 545.559478117           Accuracy: 0.906
C = 2335.72146909           Accuracy: 0.906
C = 10000.0                  Accuracy: 0.906
[[ 0.1119  0.2301  0.6438  0.7169  0.7133  0.7133  0.7133  0.7133  0.7133
   0.7133  0.7133  0.7133  0.7133  0.7133  0.7133  0.7133  0.7133  0.7133
   0.7133  0.7133]
 [ 0.0963  0.425   0.7747  0.8005  0.7947  0.7947  0.7947  0.7947  0.7947
   0.7947  0.7947  0.7947  0.7947  0.7947  0.7947  0.7947  0.7947  0.7947
   0.7947  0.7947]
 [ 0.3171  0.7527  0.865   0.8635  0.8552  0.8548  0.8548  0.8548  0.8548
   0.8548  0.8548  0.8548  0.8548  0.8548  0.8548  0.8548  0.8548  0.8548
   0.8548  0.8548]
 [ 0.5782  0.85     0.8867  0.8909  0.8815  0.881   0.881   0.881   0.881
   0.881   0.881   0.881   0.881   0.881   0.881   0.881   0.881   0.881
   0.881   0.881 ]
 [ 0.7981  0.8824  0.9043  0.9086  0.895   0.8926  0.8926  0.8926  0.8926
   0.8926  0.8926  0.8926  0.8926  0.8926  0.8926  0.8926  0.8926  0.8926
   0.8926  0.8926]
 [ 0.8721  0.9038  0.9183  0.9202  0.9135  0.9041  0.903   0.903   0.903
   0.903   0.903   0.903   0.903   0.903   0.903   0.903   0.903   0.903
   0.903   0.903 ]
 [ 0.8935  0.9147  0.9255  0.9298  0.9229  0.9118  0.9071  0.906   0.906
   0.906   0.906   0.906   0.906   0.906   0.906   0.906   0.906   0.906
   0.906   0.906 ]]

```

```

In [28]: # Find the index of the maximum value in the accuracies table
maxindex = np.array([int(len(Accs)*np.argmax(Accs)/(len(Accs.flatten()))),
print('The index of the maximum accuracy ('+str(Accs[maxindex[0],maxindex[1]]

besthp = hyperparams[maxindex[1]]
bestns = samples[maxindex[0]]
# Determine which sample count-hyperparameter combination this corresponds
print('This corresponds to a hyperparameter of C = '+ str(besthp) + ' when

```

The index of the maximum accuracy (0.9298) is: [6 3]
This corresponds to a hyperparameter of C = 7.84759970351e-07 when training on 1000

```

In [29]: besthp = 7.84759970351e-07
bestns = 10000

```

```

In [30]: # Load test data
testpath = r"mnist\test.mat"

```

```

mnist_test = utils.loaddata(testpath,_DATA_DIR,'testX')
predictions = tf.TrainAndPredictNsamples(trainsetarrays[:bestns],trainsetL

In [27]: IDs = np.arange(len(predictions))
numpycsv = np.c_[IDs,predictions]
np.savetxt(_LOCAL_PATH+r'\MNIST_testpredictions.csv',numpycsv,fmt='%i',del

In [31]: from matplotlib import pyplot as plt

In [32]: hpC1 = 13

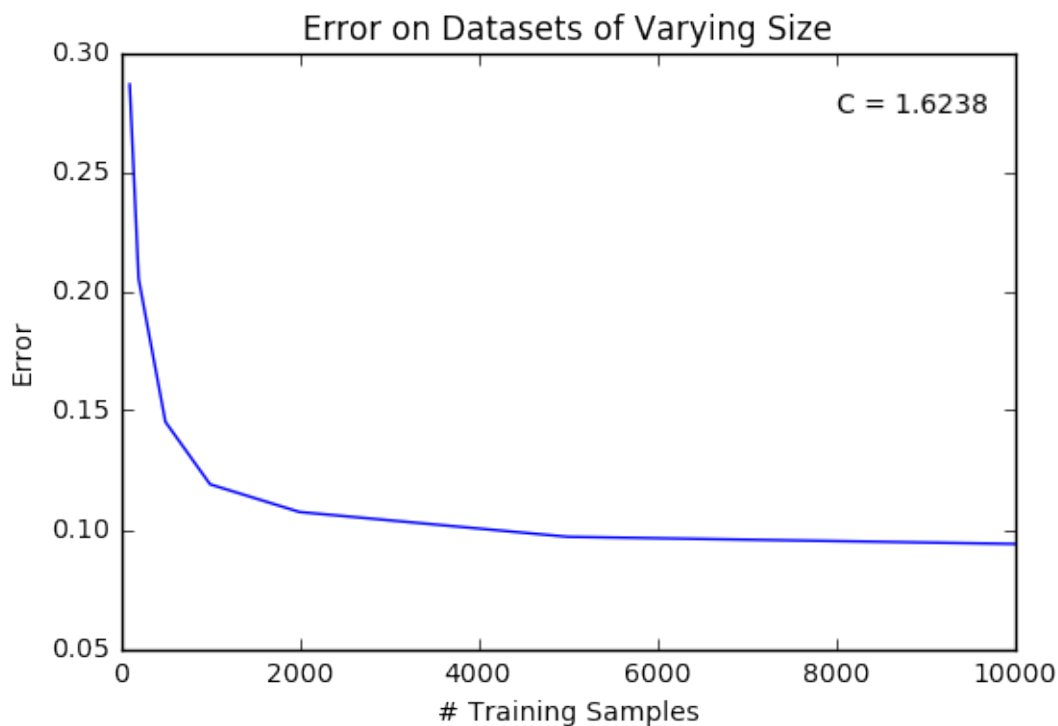
In [33]: errors = np.ones_like(Accs[:,hpC1])-Accs[:,hpC1]

In [42]: fig = plt.figure()
plt.plot(samples,errors)
plt.title('Error on Datasets of Varying Size')
plt.xlabel('# Training Samples')
plt.ylabel('Error')
plt.text(8000,0.275,'C = '+str(round(hyperparams[hpC1],4)))

Out[42]: <matplotlib.text.Text at 0x23b99f887f0>

In [43]: plt.show()

```



```

In [47]: fig.savefig(_LOCAL_PATH+r'\Figures\MNIST_SampleAcc.jpg')

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```
In [ ]: # Export data to csv files for report
        np.savetxt(_LOCAL_PATH+r'\MNIST_Accuracies.csv', Accs, fmt='%f', delimiter=',',
        np.savetxt(_LOCAL_PATH+r'\MNIST_hyperparams.csv', hyperparams, fmt='% .8f', del
```