CS441: Applied ML - HW 1

Parts 1-2: MNIST

Include all the code for generating MNIST results below

```
In [ ]: # initialization code
        import numpy as np
         from tensorflow import keras
         from keras.datasets import mnist
         # from tensorflow import keras
        %matplotlib inline
         from matplotlib import pyplot as plt
         from scipy import stats
        def load mnist():
           Loads, reshapes, and normalizes the data
           (x_train, y_train), (x_test, y_test) = mnist.load_data() # loads MNIST dat
           x_{train} = np.reshape(x_{train}, (len(x_{train}), 28*28)) # reformat to 768-d
           x_{\text{test}} = \text{np.reshape}(x_{\text{test}}, (\text{len}(x_{\text{test}}), 28*28))
           maxval = x_train.max()
           x_train = x_train/maxval # normalize values to range from 0 to 1
           x \text{ test} = x \text{ test/maxval}
           return (x_train, y_train), (x_test, y_test)
         def display_mnist(x, subplot_rows=1, subplot_cols=1):
           Displays one or more examples in a row or a grid
           if subplot_rows>1 or subplot_cols>1:
             fig, ax = plt.subplots(subplot_rows, subplot_cols, figsize=(15,15))
             for i in np.arange(len(x)):
               ax[i].imshow(np.reshape(x[i], (28,28)), cmap='gray')
               ax[i].axis('off')
           else:
               plt.imshow(np.reshape(x, (28,28)), cmap='gray')
               plt.axis('off')
           plt.show()
```

```
In []: # example of using MNIST load and display functions
  (x_train, y_train), (x_test, y_test) = load_mnist()
  display_mnist(x_train[:10],1,10)
  print('Total size: train={}, test ={}'.format(len(x_train), len(x_test)))
```



Total size: train=60000, test =10000

1. Retrieval, Clustering, and NN Classification

```
In [ ]: # Retrieval
        import math
        def get_nearest(X_query, X):
          ''' Return the index of the sample in X that is closest to X_query accordi
              to L2 distance '''
          # TO DO
          mindis = math.inf
          n = X.shape[0]
          res = -1
          for i in range(n):
            dis = np.linalg.norm(X[i]-X_query)
            if dis<mindis:</pre>
              mindis = dis
              res = i
          if res>=0:
            return res
          else:
            print("error happens!")
        # print(x train.shape)
        # print(x_test.shape)
        j = get_nearest(x_test[0], x_train)
        print(j)
        j = get_nearest(x_test[1], x_train)
        print(j)
       53843
       28882
In [ ]: # K-means
        def kmeans(X, K, niter=10):
          Starting with the first K samples in X as cluster centers, iteratively ass
          point to the nearest cluster and compute the mean of each cluster.
          Input: X[i] is the ith sample, K is the number of clusters, niter is the r
          Output: K cluster centers
          1.1.1
          # TO DO -- add code to display cluster centers at each iteration also
          centers = np.copy(X[:K])
          for i in range(niter):
            choice = np.array([get_nearest(x, centers) for x in X])
            for j in range(K):
              centers[j] = np.mean(X[choice==j],axis=0)
            if i==0 or i==9:
              print(f"iter={i+1}")
              display_mnist(centers,1,30)
          return centers
        K=30
        centers = kmeans(x_train[:1000], K)
       iter=1
```

504192131435361728694091129328

iter=10

509192331435361768694041207521

incorrect rate: 8%

2. Make it fast

```
In []: # install libraries you need for part 2
!apt install libomp-dev
!pip install faiss-cpu
import faiss
import time
```

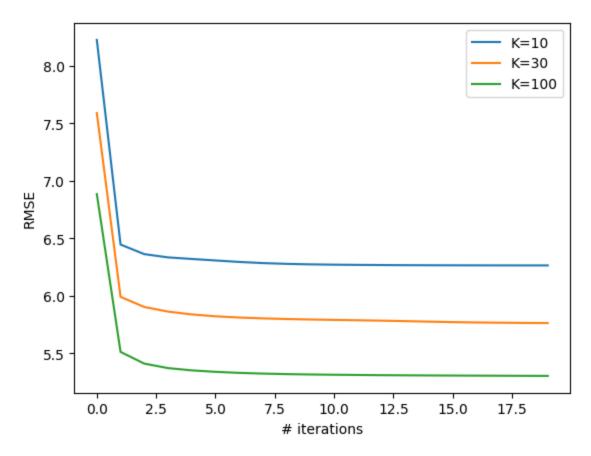
The operation couldn't be completed. Unable to locate a Java Runtime that su pports apt.

Please visit http://www.java.com for information on installing Java.

Requirement already satisfied: faiss-cpu in /Users/janghl/anaconda3/envs/am l/lib/python3.10/site-packages (1.7.4)

```
In [ ]: # retrieval
        # TO DO (check that you're using FAISS correctly)
        def fast_nearest(X_query, X, K=1, LSH=False):
          if not LSH:
            index = faiss.IndexFlatL2(x_train.shape[1]) # set for exact search
          else:
            dim = X.shape[1]
            index = faiss.IndexLSH(dim, dim)
          index.add(X) # add the data
          dist, idx = index.search(X guery, K) # returns index and sq err for each s
          return np.array(dist), np.array(idx)
        dist, idx = fast_nearest(x_test, x_train)
        print(idx[0])
        print(idx[1])
       [53843]
       [28882]
In [ ]: # K-means
        def kmeans_fast(X, K, niter=10):
```

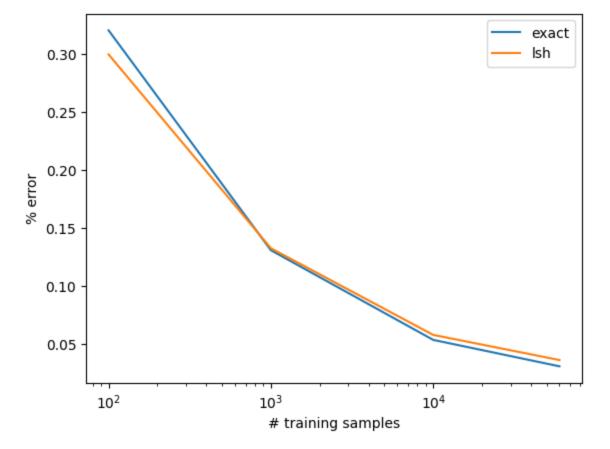
Starting with the first K samples in X as cluster centers, iteratively ass point to the nearest cluster using faiss and compute the mean of each clus Input: X[i] is the ith sample, K is the number of clusters, niter is the r Output: K cluster centers # TO DO (you can base this on part 1, but use FAISS for search) # if you include display code, you need to re-organize the plotting code b rmse = []centers = np.copy(X[:K]) for i in range(niter): dist, idx = fast nearest(X, centers) choice = np.array(idx).reshape(idx.shape[0]) rmse.append(np.sqrt(np.sum(dist)/dist.shape[0])) for j in range(K): centers[j] = np.mean(X[choice==j],axis=0) return centers, rmse K=10 centers, rmse = kmeans_fast(x_train, K, niter=20) plt.plot(np.arange(len(rmse)), rmse, label='K=10') centers, rmse = kmeans_fast(x_train, K, niter=20) plt.plot(np.arange(len(rmse)), rmse, label='K=30') K=100 centers, rmse = kmeans_fast(x_train, K, niter=20) plt.plot(np.arange(len(rmse)), rmse, label='K=100') plt.legend(), plt.ylabel('RMSE'), plt.xlabel('# iterations') plt.show()

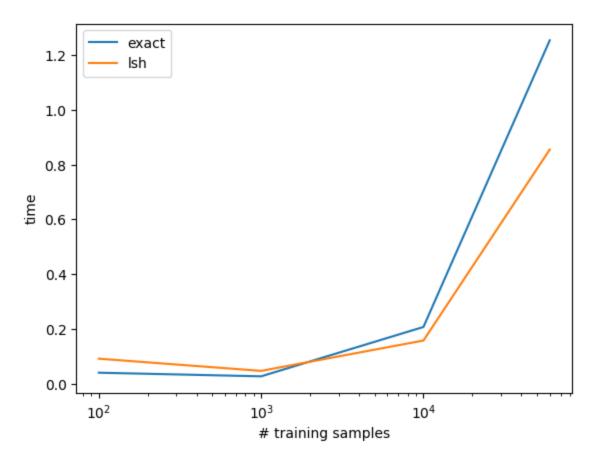


```
In [ ]: # 1-NN
        nsample = [100, 1000, 10000, 60000]
        # TO DO
        def ONN(x_train, x_test, s, LSH):
          begin = time.time()
          train = x_train[:s]
          right = 0
          if LSH:
            dim = x_{test.shape}[1]
            index = faiss.IndexLSH(dim, dim)
          else:
            index = faiss.IndexFlatL2(x_test.shape[1])
          index.add(train)
          dist, idx = index.search(x_test,1) # returns index and sq err for each san
          idx = np.array(idx).reshape(idx.shape[0])
          for i in range(len(x_test)):
            if y_test[i] == y_train[idx[i]]:
               right += 1
          acc = right/len(x_test)
          end = time.time()
          timing = end-begin
          return acc, timing
        acc exact = []
        acc_lsh = []
        timing_exact = []
        timing_lsh = []
```

```
for s in nsample:
    acc_exact_elem, timing_exact_elem = ONN(x_train, x_test, s, LSH=False)
    acc_lsh_elem, timing_lsh_elem = ONN(x_train, x_test, s, LSH=True)
    acc_exact.append(1-acc_exact_elem)
    acc_lsh.append(1-acc_lsh_elem)
    timing_exact.append(timing_exact_elem)
    timing_lsh.append(timing_lsh_elem)

plt.semilogx(nsample, acc_exact, label='exact')
plt.semilogx(nsample, acc_lsh, label='lsh')
plt.legend(), plt.ylabel('% error'), plt.xlabel('# training samples')
plt.semilogx(nsample, timing_exact, label='exact')
plt.semilogx(nsample, timing_lsh, label='lsh')
plt.legend(), plt.ylabel('time'), plt.xlabel('# training samples')
plt.show()
```





```
In [ ]: # Confusion matrix
         from sklearn import metrics
         # TO DO
         index = faiss.IndexFlatL2(x_test.shape[1])
         index.add(train)
         dist, idx = index.search(x_test,1) # returns index and sq err for each sampl
         idx = np.array(idx).reshape(idx.shape[0])
         metrics.confusion_matrix(y_test, y_train[idx])
Out[]: array([[ 971,
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```

Part 3: Temperature Regression

Include all your code used for part 2 in this section.

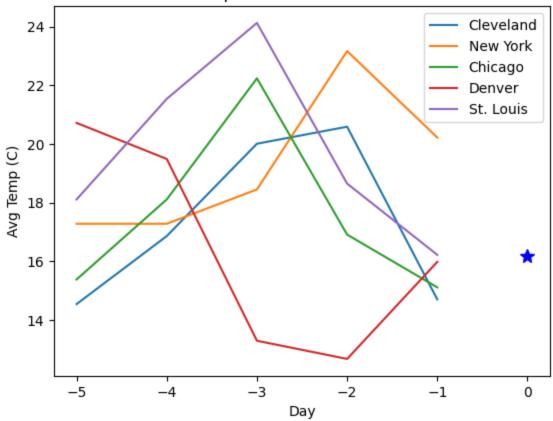
```
In []: import numpy as np
# from google.colab import drive
%matplotlib inline
```

from matplotlib import pyplot as plt

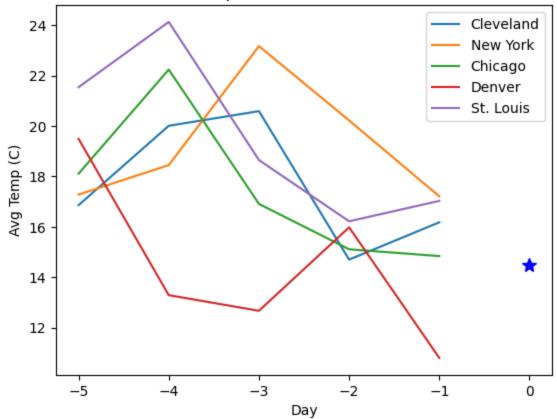
```
from sklearn.linear model import Ridge
        from sklearn.linear model import Lasso
        # load data (modify to match your data directory or comment)
        def load temp data():
          # drive.mount('/content/drive')
          # datadir = "/content/drive/My Drive/CS441/24SP/hw1/"
          datadir = "./"
          T = np.load(datadir + 'temperature_data.npz')
          x_train, y_train, x_val, y_val, x_test, y_test, dates_train, dates_val, dates_val
          T['x_train'], T['y_train'], T['x_val'], T['y_val'], T['x_test'], T['y_test
           return (x_train, y_train, x_val, y_val, x_test, y_test, dates_train, dates
        # plot one data point for listed cities and target date
        def plot_temps(x, y, cities, feature_to_city, feature_to_day, target_date):
          nc = len(cities)
          ndays = 5
          xplot = np.array([-5, -4, -3, -2, -1])
          yplot = np.zeros((nc,ndays))
          for f in np.arange(len(x)):
             for c in np.arange(nc):
               if cities[c] == feature to city[f]:
                 yplot[feature_to_day[f]+ndays,c] = x[f]
          plt.plot(xplot,yplot)
          plt.legend(cities)
          plt.plot(0, y, 'b*', markersize=10)
          plt.title('Predict Temp for Cleveland on ' + target date)
          plt.xlabel('Day')
          plt.ylabel('Avg Temp (C)')
          plt.show()
In [ ]: # load data
         (x_train, y_train, x_val, y_val, x_test, y_test, dates_train, dates_val, dat
         ''' Data format:
               x_train, y_train: features and target value for each training sample (
               x_val, y_val: features and target value for each validation sample (us
               x_test, y_test: features and target value for each test sample (used t
               dates_xxx: date of the target value for the corresponding sample
               feature to city: maps from a feature number to the city
               feature_to_day: maps from a feature number to a day relative to the ta
               Note: 361 is the temperature of Cleveland on the previous day
        1.1.1
        f = 361
        print('Feature {}: city = {}, day= {}'.format(f, feature to city[f], feature
        baseline_rmse = np.sqrt(np.mean((y_val[1:]-y_val[:-1])**2)) # root mean squa
        print('Baseline - prediction using previous day: RMSE={}'.format(baseline_rm
        # plot first two x/y for val
        plot_temps(x_val[0], y_val[0], ['Cleveland', 'New York', 'Chicago', 'Denver'
plot_temps(x_val[1], y_val[1], ['Cleveland', 'New York', 'Chicago', 'Denver'
```

Feature 361: city = Cleveland, day= -1
Baseline - prediction using previous day: RMSE=3.460601246750482

Predict Temp for Cleveland on 2018-09-27



Predict Temp for Cleveland on 2018-09-28



```
In []: # K-NN Regression
        import copy
        def regress KNN(X trn, y trn, X tst, K=1):
          Predict the target value for each data point in X_tst using a
          K-nearest neighbor regressor based on (X_trn, y_trn), with L2 distance.
          Input: X trn[i] is the ith training data. y trn[i] is the ith training lab
          Output: return y pred, where y pred[i] is the predicted ith test value
          1.1.1
          # TO DO
          dist, idx = fast_nearest(x_test, x_train, K=K)
          y pred = np.zeros(shape=X tst.shape[0])
          for i in range(len(X tst)):
            points = []
            for j in idx[i]:
              points.append(y_trn[j])
              y_pred[i] = np.mean(points)
          return y pred
        def normalize features(x, y, fnum):
          ''' Normalize the features in x and y.
              For each data sample i:
                x2[i] = x[i]-x[i,fnum]
                y2[i] = y[i]-x[i,fnum]
          111
          # TO DO
          x2 = np.zeros like(x)
          y2 = np.zeros like(y)
          xsub = np.array(np.copy(x[:,fnum]))
          for i in range(x.shape[1]):
            x2[:,i] = x[:,i] - xsub
          y2 = y - xsub
          return x2, y2
        # KNN with original features
        # TO DO
        y_predict = regress_KNN(x_train, y_train, x_test, K=5)
        # print(f"y_predict: {y_val.shape}\nx_test: {x_val.shape}")
        print(f"rmse when K=5: {np.sqrt(np.mean((y test-y predict)**2))}")
        # KNN with normalized features
        fnum = 361 # previous day temp in Cleveland
        # TO DO
        x_train_new, y_train_new = normalize_features(x_train, y_train, fnum)
        x test new, y test new = normalize features(x test, y test, fnum)
        y predict new = regress KNN(x \text{ train new, } y \text{ train new, } x \text{ test new, } K=5)
        print(f"rmse when K=5: {np.sqrt(np.mean((y_test_new-y_predict_new)**2))}")
```

rmse when K=5: 3.249556245363484 rmse when K=5: 2.9770334300836487

Part 5: Stretch Goals

Include all your code used for part 5 in this section. You can copy-paste code from parts 1-3 if it is re-usable.

```
In [ ]: # Stretch: KNN classification (Select K)
        # find the prediction for a set
        def calc_pred(valid, dist, idx, y_train, K):
          pred = np.zeros(valid.shape[0])
          for i in range(valid.shape[0]):
            count = {}
            distance = {}
            for j in range(K):
                                           #for each of K points, p is predicted numb
              p = y_train[idx[i][j]]
              d = dist[i][j]
              if p in distance:
                 distance[p] += d
                 distance[p] = d
              if p in count:
                 count[p] += 1
              else:
                count[p] = 1
            maxnum = -math.inf
                                     #find all candidates with highest votes
            candidate = []
            for c in count:
              if count[c]>maxnum:
                maxnum = count[c]
                 candidate = [c]
              elif count[c]==maxnum:
                 candidate.append(c)
            mindist = math.inf
            result = -1
            for can in candidate:
              if distance[can]<mindist:</pre>
                 result = can
            pred[i] = result
          return pred
        def KNN_classification(x_train, y_train, K):
          train = x_train[:50000]
          valid = x_{train}[50000:]
          answer = y train[50000:]
          dist, idx = fast_nearest(valid, train, K=K)
          correct = 0
          pred = calc_pred(valid, dist, idx, y_train, K)
          for i in range(len(valid)):
            if pred[i] == answer[i]:
              correct += 1
```

```
print(f"incorrect rate for K={K}: {100*(1-correct/len(valid))}%")
        (x_train, y_train), (x_test, y_test) = load_mnist()
        hyperparameter = [1, 3, 5, 11, 25]
        for K in hyperparameter:
          KNN_classification(x_train, y_train, K)
        # Perform K=3 on the test set
        best K = 3
        dist, idx = fast nearest(x test, x train, K=best K)
        correct = 0
        pred = calc_pred(x_test, dist, idx, y_train, best_K)
        for i in range(len(x test)):
          if pred[i] == y test[i]:
            correct += 1
        print(f"incorrect rate for K={best_K}: {100*(1-correct/len(x_test))}%")
       incorrect rate for K=1: 2.880000000000005%
       incorrect rate for K=3: 2.8200000000000003%
       incorrect rate for K=5: 2.949999999999997%
       incorrect rate for K=11: 3.1499999999999972%
       incorrect rate for K=25: 3.879999999999946%
       incorrect rate for K=3: 2.959999999999996%
In [ ]: # Stretch: KNN regression (Select K)
        (x_train, y_train, x_val, y_val, x_test, y_test, dates_train, dates_val, dat
        # K-NN Regression
        hyperparameter = [1, 3, 5, 11, 25]
        for K in hyperparameter:
          # KNN with original features
          y predict = regress KNN(x train, y train, x val, K=K)
          print(f"rmse when K={K}: {np.sqrt(np.mean((y_val-y_predict)**2))}")
          # KNN with normalized features
          fnum = 361 # previous day temp in Cleveland
          x_train_new, y_train_new = normalize_features(x_train, y_train, fnum)
          x_val_new, y_val_new = normalize_features(x_val, y_val, fnum)
          y predict new = regress KNN(x train new, y train new, x val new, K=K)
          print(f"(normalized)rmse when K={K}: {np.sqrt(np.mean((y_val_new-y_predict
       rmse when K=1: 5.994830467058024
       (normalized) rmse when K=1: 4.962816257477145
       rmse when K=3: 5.0719820622291065
       (normalized) rmse when K=3: 4.081979762608333
       rmse when K=5: 4.814445285619017
       (normalized) rmse when K=5: 3.7840627858326146
       rmse when K=11: 4.635825782051689
       (normalized) rmse when K=11: 3.6293441508410265
       rmse when K=25: 4.482832870675747
       (normalized) rmse when K=25: 3.544085254919383
```

```
In [ ]: # Best on test set
        y_predict = regress_KNN(x_train, y_train, x_test, K=25)
        print(f"rmse when K={K}: {np.sqrt(np.mean((y test-y predict)**2))}")
       rmse when K=25: 3.041778206247129
In [ ]: # Stretch: K-means (more iters vs redos)
        def test(niter, nredo):
          kmeans = faiss.Kmeans(x_train.shape[1], 30, niter=niter, nredo=nredo, seed
          kmeans.train(x train)
          dist, idx = kmeans.index.search(x train, 1)
          rmse = np.sqrt(np.sum(dist) / x_train.shape[0])
          return rmse
        (x_train, y_train), (x_test, y_test) = load_mnist()
        first trail multiple = []
        first_trail_single = []
        second trail multiple = []
        second trail single = []
        for i in range(5):
          first trail multiple.append(test(10,5))
          first trail single.append(test(50,1))
          second_trail_multiple.append(test(4,5))
          second trail single.append(test(20,1))
        print(f"for first_trail_multiple: mean={np.mean(first_trail_multiple)}, std=
        print(f"for first trail single: mean={np.mean(first trail single)}, std={np.
        print(f"for second_trail_multiple: mean={np.mean(second trail multiple)}, st
        print(f"for second trail single: mean={np.mean(second trail single)}, std={r
       for first trail multiple: mean=5.789932893969209, std=0.0
       for first trail single: mean=5.769123994738427, std=0.0
       for second_trail_multiple: mean=5.819028949632977, std=0.0
       for second trail single: mean=5.775398976117003, std=0.0
In []: # from https://gist.github.com/jonathanagustin/b67b97ef12c53a8dec27b343dca4a
        # For use in Colab. For local, just use jupyter nbconvert directly
        import os
        # @title Convert Notebook to PDF. Save Notebook to given directory
        NOTEBOOKS DIR = "/content/drive/My Drive/CS441/24SP/hw1" # @param {type:"str
        NOTEBOOK_NAME = "CS441_SP24_HW1_Solution.ipynb" # @param {type:"string"}
        from google.colab import drive
        drive.mount("/content/drive/", force_remount=True)
        NOTEBOOK PATH = f"{NOTEBOOKS DIR}/{NOTEBOOK NAME}"
        assert os.path.exists(NOTEBOOK PATH), f"NOTEBOOK NOT FOUND: {NOTEBOOK PATH}"
        !apt install -y texlive-xetex texlive-fonts-recommended texlive-plain-generi
        !jupyter nbconvert "$NOTEBOOK_PATH" -- to pdf > /dev/null 2>&1
        NOTEBOOK_PDF = NOTEBOOK_PATH.rsplit('.', 1)[0] + '.pdf'
        assert os.path.exists(NOTEBOOK_PDF), f"ERROR MAKING PDF: {NOTEBOOK_PDF}"
        print(f"PDF CREATED: {NOTEBOOK_PDF}")
```