```
In [ ]:
         import matplotlib.pyplot as plt
         import seaborn as sns; sns.set() # for plot styling
         import numpy as np
         import pandas as pd
         from sklearn.cluster import k_means, kmeans_plusplus, KMeans, AgglomerativeClust
         from sklearn.decomposition import PCA
         import math
         import matplotlib.pyplot as plt
         import matplotlib as mpl
         from sklearn.mixture import GaussianMixture
         from sklearn.model selection import GridSearchCV, train test split, cross val sc
         from sklearn.linear_model import LogisticRegression
         from sklearn.pipeline import Pipeline
         import math
         from random import randint
```

Assignment 3 - Clustering

For this assignment you'll need to use some clustering knowledge to build a function that can generate handwritten numbers from a provided number.

The modelling parts of this assignment are not very complex, the application of the clustering algorithms is very similar to the examples from class. This will require a little more manipulation of data, and building a little bit of structure around the models, that's where some of the challenge lies.

Requirements

- Use clustering to take the X data (the features/pixels) of the MNIST dataset, and group it into clusters.
 - Do not use the targets from the dataset at all.
- Assign labels to your clusters, so there is now a label for each cluster. You'll need to manually do a little mapping here by eye.
- Use GMM to build a function that can generate a new digit from the information in the cluster.
- Write a function, writeNumber, that can take in an integer (you can assume it is between 1
 and 20 digits, this is mostly for printing purposes, the modelling part isn't impacted by this
 at all) and print out that integer as a generated handwritten number.
- When generating the handwritten numbers, each version of a digit should be different. I.e. If the number printed is 22222, there should not be 5 identical 2s, they should vary a bit like real writing.

Deliverables

Submit your .ipynb file to the Moodle dropbox. PLEASE make sure that the file runs BY ITSELF outside of importing libraries. It should not reference any other files, either data or code.

Within your file, create a function called writeNumber, which takes a number as an input, and prints it out as a series of handwritten digits. The function should be callable like this, if I wanted to print the number 218201

```
writeNumber(218201)
```

This would produce something that looks, somewhat, like this:

```
Writing Example
```

The exact appearance of the generated writing can vary, it likely won't be as well defined as this picture. As long as I can read it as a number, it is fine.

Hint: the number of clusters may vary.

Grading

- 70% Code works. This is mostly a yes/no thing, if it doesn't work I'll try to see if it was a small or large issue in the code, but it largely works or it doesn't.
- 20% Numbers appearance. Is the writing OK? I'm not going to be overly picky, if they are reasonably legible, that is fine.
- 10% Code legibility. Is a quick read over of the code clear? Sections, comments, etc...

•

```
In []:
         #Look at an image
         def showDigit(digit, label, size=28):
             some digit = digit
             #turn array into the correct shape
             some_digit_image = np.array(some_digit).reshape(size, size)
             #imshow displays an array like an image
             plt.imshow(some digit image, cmap=mpl.cm.binary)
             plt.title(label)
             plt.axis("off")
             plt.show()
         #Display multiple digits
         def showDigits(digits, labels, indexes, size=28):
             #Make a grid that is the right size
             pics = len(indexes)
             cols = 8
             rows = math.ceil(pics/cols)
             fig, axes = plt.subplots(rows, cols, figsize=(14,6))
             plt.axis("off")
             #loop through the list of indexes, grab images and labels, plot in the "next
             for i in range(0, pics):
```

```
n = indexes[i]
some_digit = digits[n:n+1]
some_digit_image = np.array(some_digit).reshape(size, size)
ax = axes[i//cols, i%cols]
ax.axis("off")
ax.imshow(some_digit_image, cmap=mpl.cm.binary)
#ax.set_title('Ind: {} - Lbl: {}'.format(indexes[i], labels[n]))
plt.tight_layout()
plt.axis("off")
plt.show()
```

Load Data

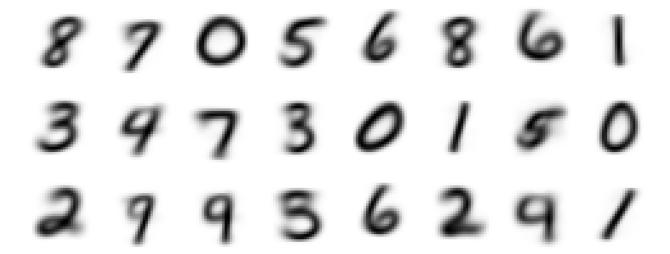
Please do not change this (substantially), probably outside of choosing between the full data and a subset. Don't load the target.

Note: testing will be much faster with a subset of records.

```
In []: #Load Data
    from sklearn.datasets import fetch_openml
    mnist = fetch_openml('mnist_784', version=1)
    #mnist = mnist[0:15000]
    X = mnist["data"]
    print(X.shape)
(70000, 784)
```

Cluster

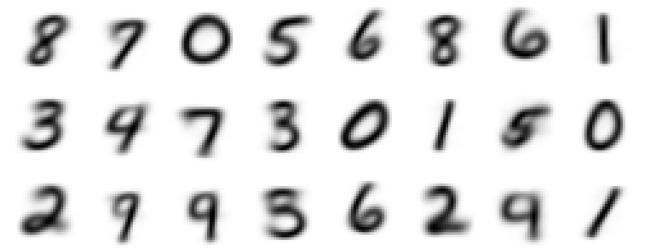
We need to break the data into clusters first.



Cluster with PCA

```
In []:
    kmeans = KMeans(n_clusters=24, init="k-means++", random_state=12)
    clus_pca = PCA()
    clus_trans = clus_pca.fit_transform(X)
    clusters = kmeans.fit_predict(clus_trans)

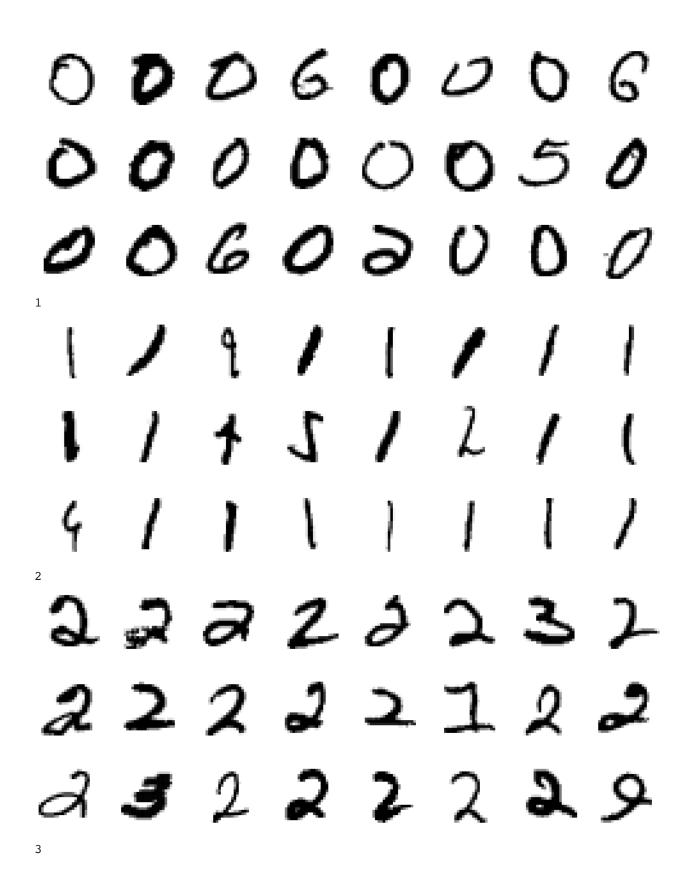
    centers = clus_pca.inverse_transform(kmeans.cluster_centers_).reshape(24, 28, 28 showDigits(digits=centers, labels=y_null, indexes=indexes, size=28)
```

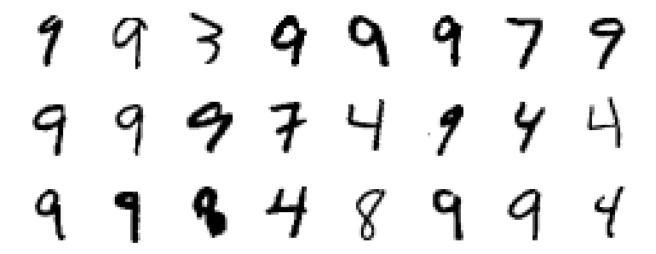


Visually Match the Clusters

Note: The label here is not the actual value of that number, they're assigned sequentially by the clustering, which does not know what each number is.

```
9
                                                       3446
                                19
                                                       3339
                                7
                                                       3260
                                1
                                                       3247
                                14
                                                       3227
                                13
                                                       3200
                                22
                                                       3083
                                21
                                                       2936
                                8
                                                       2890
                                0
                                                       2865
                                20
                                                       2798
                                                       2782
                                4
                                16
                                                       2766
                                5
                                                       2735
                                10
                                                       2636
                                23
                                                       2433
                                2
                                                       2318
                                3
                                                       2317
                                12
                                                       2226
                                15
                                                       2031
                                                       1749
                                Name: label, dtype: int64
In [ ]:
                                  map\_dict = \{0:8, 1:7, 2:0, 3:5, 4:6, 5:8, 6:6, 7:1, 8:3, 9:4, 10:7, 11:3, 12:0, 12:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13:0, 13
                                  df["label"] = df["label"].map(map_dict)
                                  df["label"].value_counts()
                                                   11154
Out[]:
                                1
                                                       8893
                                5
                                                       8883
                                6
                                                       7329
                                0
                                                       6575
                                3
                                                       6535
                                7
                                                       5883
                                2
                                                       5702
                                8
                                                       5600
                                                       3446
                                4
                                Name: label, dtype: int64
In []:
                                   for i in range(10):
                                                   samp ind = df[df["label"] == i].index.values
                                                  rows = df.loc[samp_ind,:].sample(24)
                                                  print_ind = rows.index.values
                                                  y tmp = np.full(10, i)
                                                  ind tmp = np.array(range(0,10))
                                                   #print(print ind)
                                                  print(i)
                                                   showDigits(df.drop(columns={"label"}).to_numpy(), y_tmp, print_ind)
```



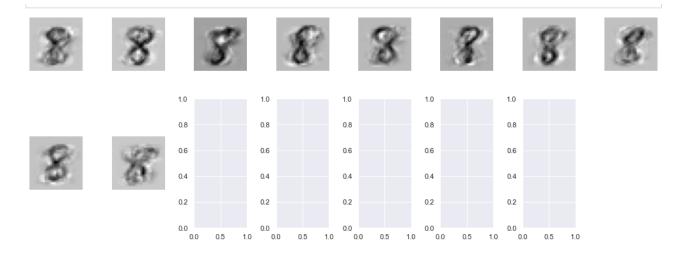


Data is Labeled

We now have a label for each cluster. They are moderately accurate by visual verification.

Create Generators with GMM

```
In [ ]:
         np.array(df[df["label"]==1].drop(columns={"label"})).shape
        (8893, 784)
Out[]:
In [ ]:
         generators = []
         for i in range(10):
             gmm = GaussianMixture(covariance type='full', random state=0)
             tmp_data = np.array(df[df["label"] == i].drop(columns={"label"}))
             gmm.fit(tmp data)
             generators.append(gmm)
             print(str(i), "Converged:", gmm.converged_)
        0 Converged: True
        1 Converged: True
        2 Converged: True
        3 Converged: True
        4 Converged: True
        5 Converged: True
        6 Converged: True
        7 Converged: True
        8 Converged: True
        9 Converged: True
In [ ]:
         data new = generators[8].sample(16)
         data new = data new[0]
         print(data_new.shape)
        (10, 784)
        (10, 784)
In [ ]:
         showDigits(digits=data_new, labels=y_null, indexes=[1,2,3,4,5,6,7,8,9,0], size=2
```



Print a Number

```
In []:
         def get digit(number, i):
             return number // 10**i % 10
         def num_printer(number):
             cols = int(math.log10(number))+1
             fig, axes = plt.subplots(rows, cols, figsize=(14,6))
             i = 0
             while i < cols:</pre>
                 n = get_digit(number,i)
                 dig = generators[n].sample(100)
                 dig = dig[0][randint(0,100)]
                 ax = axes[cols-1-i] #Need to find why it is backwards
                 img = np.array(dig).reshape(28,28)
                 ax.imshow(img, cmap=mpl.cm.binary, interpolation='nearest')
                 #ax.imshow(img, cmap='gray_r')
                 #title_str = "I:", str(i), "N:", str(n)
                 #ax.set title(title str)
                 i += 1
             plt.tight layout()
             plt.setp(plt.gcf().get_axes(), xticks=[], yticks=[])
             plt.show()
```

In []: num_printer(222226666888)























