## Statistical Learning for Engineers (EN.530.641) Homework 7

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Out: 11/04/2022 due: 11/11/2022 by midnight EST

This is exclusively used for Fall 2022 EN.530.641 SLE students, and is not to be posted, shared, or otherwise distributed.

1 In this problem, you will perform image compression which is basically principal component analysis (PCA). First download a dataset in Scikit-Learn by:

from sklearn.datasets import fetch\_openml

The data MNIST is about hand-written zip codes. Then load data by:

X, y = fetch\_openml('mnist\_784', return\_X\_y=True)

Then standardize the data by using StandardScaler in sklearn.preprocessing.

Now apply PCA. You will have to use PCA from sklearn.decomposition. Set the number of principal components n\_components by 0.95, which mean only 95% of the variance will be preserved.

- (a) Generate the plot of "cumulative explained variance" vs. dimensions (or number of principal components). To do that, you will have to use cumsum from numpy with pca.explained\_variance\_ratio as its input.
- (b) From the plot, what is the number of principal components (i.e., dimensions) corresponding to preserving 95% cumulative variance?
- (c) Select one image randomly from the original dataset, and plot "before" and "after" PCA compression.

## Submission Guideline

- Submit your homework answers in a single pdf format, including plots and so on, to "HW7\_analytical" on the gradescope.
- Submit all your python codes in a single .zip file that contains codes for each problem (name them by including the problem number). Name your single zip file submission as "Your-Name\_HW7.zip". For example, "JinSeobKim\_HW7.zip" for a single zip file. Submission will be done through "HW7\_computational" on the gradescope.

SLE (EN.530.641) 2

• Just in case you have related separate files, please make sure to include *all the necessary files*. If TAs try to run your function and it does not run, then your submission will have a significant points deduction.

• Make as much comments as possible so that the TAs can easily read your codes.