

Problem 1: Assessing Performance on Blind Predictions From HW3 (10 points)

In Homework 3, we asked you to make predictions on a blind data set. Here, we want to assess how well you did on that problem.

For HW 3, problem 5, please assess the AUC between the true outcomes in “hw3_problem_5_y_blind.csv” versus your predictions.

- (a) What was your hold-out AUC?
- (b) A “good” AUC here is about .79 using **4 components**. How did your results compare to this? Were you able to reduce the number of components by incorporating the PCA into the validation procedure?
- (c) Discovering that 4 components was ideal would require a much larger parameter search space than you may of used. How does the larger search space affect multiple comparisons?

Problem 2: Permutation Testing (25 points)

We want to examine the impact of dataset size on permutation testing. To evaluate this, try running a cross-validation grid search procedure (you can choose the procedure) on permuted labels with N datapoints, where N is [20,35,50,100,200,400]. Use 10 repeats for each setting.

- (a) Plot the mean and standard deviation of the results as a function of the number of data.
- (b) Looking at the trends, what can we say about the bias in the predictive performance results and the effect of more data?
- (c) How many distinct models did your cross-validation procedure try? How would increasing or decreasing the search space affect the results?

Problem 3: Forecasting Time Series (30 points)

The famous CO2 dataset from the Mauna Loa observatory is in the class repository. In the homework template, I have placed code to preprocess the dataset to address missing data. Then, there are two versions of the data: the first is the raw values, and the second is “detrended” data, where a linear trend has been removed. To start building an understanding of time series, please use the time series validation methods and evaluate the following:

- (a) A linear regression model to forecast one step ahead on the raw data. Determine the optimal number of lags.
- (b) A k-Nearest Neighbor model to forecast one step ahead on the raw data. Determine the optimal number of neighbors.

- (c) On the raw data, which method worked better? Why?
- (d) A linear regression model to forecast one step ahead on the detrended data. Determine the optimal number of lags.
- (e) A linear regression model to forecast one step ahead on the detrended data. Determine the optimal number of lags.
- (f) A k-Nearest Neighbor model to forecast one step ahead on the detrended data. Determine the optimal number of neighbors.
- (g) On the detrended data, which method worked better? Why? Was performance much closer than before?
- (h) The “detrended” data simply removes a linear fit line. Did a linear fit line remove the background “trend?”

Problem 6: (5 points)

How many hours did it take you to complete this assignment?

Affirm that you adhered to the Duke Honor Code.