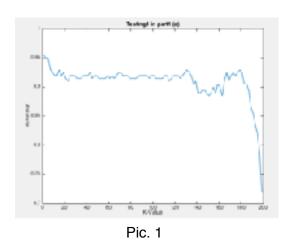
COMP24111 EX 2 Report

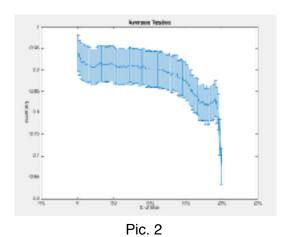
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Part 1 Training and Testing

- a) Using *shufflerows()* function to get original data in *data*. Then using *horzcat()* and *vercat()* to merge the data and its labels.
- b) K = 1 means only finding 1 training data that is closest to the testing data, which will also define the testing data. When K gets bigger, more training data could be used to define the testing data, whose value will be defined by the larger number of training data.
- c) Pic.1 is the accuracy of different K value. Training is helping a machine to set a standard of two different data. While testing is using training data to define the testing data. Pic.1 is the accuracy of varying K value.





d) Repeat the process for (c) many tomes will not get the same behaviour, because the testing data is different. Pic. 2 is the average accuracy, and standard deviation as error bars.

Part 2 Feature Extractors for Digit

First, reduce the unimportant data from dataset by setting every pixel less than 100 to 0.

Second, 2 dimension to 1 dimension.

Pic. 3 is the result of accuracy as feature extractors.



Part 3 Perceptrons

Training time is less than 0.3s.

When training interaction time is big enough (eg. more than 50), accuracy could be 100%.

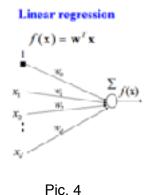
Part 4 Logistic Regression

As I learned in lecture, perceptron is a linear regression, which may get some big activations. What logistic regression does is to diminish the activations to a range of 0~1. That is the probability of an event happening.

In our 2-class KNN example, in order to judge the handwriting number is 3 or 8, we can get the probability of the testing data. If the probability of being 3 is bigger than 50%, then that could be 3, otherwise it could be 8.

But in a lot of cases, things cannot be defined totally true or wrong. So perceptrons may not be the best choice. Logistic regression can classify things by the probabilities of happening.

Pic. 4 and Pic. 5 are the conception diagram of linear regression and logistic regression.



Logistic regression $f(\mathbf{x}) = p(y = 1 | \mathbf{x}, \mathbf{w}) = g(\mathbf{w}^T \mathbf{x})$ $\begin{bmatrix} \mathbf{x} \\ \mathbf{w} \end{bmatrix} \quad \begin{bmatrix} \mathbf{y} \\ \mathbf{x} \end{bmatrix} \quad \begin{bmatrix} \mathbf{y} \\ \mathbf{y} \end{bmatrix} \begin{bmatrix}$

Pic. 5