

Hw 7

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1. DERIVE 8.16 $\hat{p}(c_i | \bar{x}) = \frac{k_i}{K}$

$$\hat{p}(c_i | \bar{x}) = \frac{\hat{p}(\bar{x} | c_i) \hat{p}(c_i)}{\sum_j \hat{p}(\bar{x} | c_j) \hat{p}(c_j)}$$

where $\hat{p}(\bar{x} | c_i) = \frac{k_i}{N_i V^k(\bar{x})}$ and $\hat{p}(c_i) = \frac{N_i}{N}$

$$\therefore \hat{p}(c_i | \bar{x}) = \left(\frac{k_i}{N_i V^k(\bar{x})} \cdot \frac{N_i}{N} \right) / \left(\sum_j \frac{k_j}{N_j V^k(\bar{x})} \frac{N_j}{N} \right)$$

$$\hat{p}(c_i | \bar{x}) = \frac{\frac{1}{V^k(\bar{x}) N} \cdot k_i}{\frac{1}{V^k(\bar{x}) N} \sum_j k_j} = \frac{k_i}{\sum_j k_j}$$

$$\hat{p}(c_i | \bar{x}) = \frac{k_i}{K}$$

2D Histogram Estimator

Accuracy with $h = 1$: **42.40%**

Halving the bin size to 0.5 decreased the accuracy by over two-fold (18.14%). Doubling the bin size to 2 increased the accuracy to 56.04%. The maximum attainable accuracy was around 60% with a bin size around 4.

Naïve Estimator

Accuracy with $h = 1$: **41.07%**

Halving the bin size to 0.5 had the same effect that it did with the histogram estimator. The accuracy dropped to 17.70%. Doubling the bin size to 2 also increased the accuracy to 55.59%. The Naïve Estimator was able to attain a marginally greater accuracy than the 2D Histogram Estimator with results above 60.21% when a bin size of 4 was used.

Kernel Estimator (Gaussian Kernel)

Accuracy with $h = 1$: **60.55%**

Halving the h -value to 0.5 decreased the accuracy to **57.21%** since the data became less smooth. If the data is smoothed out too much the accuracy also decreased. For instance, at $h=20$, the accuracy dropped below 60%.

K-Nearest Neighbor

Accuracy with $k = 1$: **52.48%**

An optimal value for k can be heuristically determined by taking the square root of N . Since $N = 1797$ for the test data, an adequate estimate for an optimal k is 42. This k values gives an accuracy of **61.49%**.

64D Histogram Estimator

Observation: The computational time and memory space required to utilize a histogram estimator in the 64-Dimensional hyperspace is exponentially greater than that of 2-Dimensional space. This provides an example of the practical utility of principal component analysis in decreasing the power and time to execute a program. The obtainable accuracy will be higher when using more dimensions, but the trade-off is time and memory. An optimal number of dimensions should be sought out to balance computational time/space and the error.