

QUESTIONS:

Q1: [2 points] [Bayesian Decision Theory]

Assume a discriminant function of the form: $g_i(x) = \ln p(x | w_i) + \ln P(w_i)$

Assume that $x \in R$, for class 1: $x \sim N(\mu_1, \sigma)$ and for class 2: $x \sim N(\mu_2, \sigma)$. Also assume that $P(w_1) = P(w_2)$.

- a) [0.5 points] For $\mu_1 = -5$, $\mu_2 = -10$, and $\sigma = 5$, plot the pdf's of the two classes inputs and also the separating surface.
- b) [0.5 points] If $P(w_1)$ increases to 0.8, where would the separating surface be?
- c) [1 points] Generate random datasets for a) and b), compute the estimates for the mean and plot histograms and the computed separating points on the histograms.

Dataset for Q2 and Q3:

Optdigits data by Alpaydin and Kaynak, from UCI Machine Learning Repository:

<ftp://ftp.ics.uci.edu/pub/ml-repos/machine-learning-databases/optdigits/>

You need the files:

optdigits.names	explanation of data
optdigits.tra	training data
optdigits.tes	test data

Q2) [1 points] [Multivariate Analysis]

Assume that each class i 's ($i=0..9$) inputs are distributed according to a normal with mean μ_i and covariance matrix Σ_i . Also assume that $\Sigma_i = \Sigma$ (common covariance matrix) and Σ is diagonal.

Implement $g(x)$ discriminant function clearly (add comments) in your code and also write its formula into the report.

Compute and report the confusion matrix, error for each class and overall error on both training and test datasets.

Hint1: Remove the features with 0 variance.

Hint2: You can use built-in functions for mean and covariance/variance calculations.

Q3) [2 points] [Dimensionality Reduction]

Q3a) [1 points] Project the instances in `optdigits.tra` dataset using the LDA algorithm into a two dimensional space as shown in the figure below.

Q3b) [0.5 points] Using the parameters of the LDA transformation you computed for the `optdigits.tra` dataset, show the two dimensional projection of `optdigits.tes` dataset in another plot.

Q3c) [0.5 points] Assume that for each class i , $\Sigma_i = \Sigma$ (common covariance matrix) and Σ is diagonal. Recompute and replot the LDA projection. Compare to the results in Q3a) and Q3b).

Hint1: do not re-compute the LDA parameters for the `optdigits.tes` set, just reuse the ones you computed for the `optdigits.tra` set.

Hint2: Do not use a built-in LDA function, implement it yourself.

