

LVC 3 - Glossary of Notations

\mathbf{X}_i = Vector containing values of input features corresponding to i^{th} record, where i ranges from 1 to n

Y_i = Value of output variable corresponding to i^{th} record

X_i = i^{th} component of a vector \mathbf{X}

\hat{Y} = Predicted value of the output variable

E = Expected value or average

$\phi(X)$ = A transformed version of the feature vector X

$h(X)$ = Non-linear classifier function

θ^T = Transpose of vector θ

μ_k = Mean vector for class k

C_k = Covariance matrix for class k

π_k = Probability of a data point belonging to class k . These are called **prior probabilities**

$N(\mu_k, C_k)$ = Normal distribution with mean μ_k and covariance C_k

γ_k = Normalising constant for class k in the normal distribution equation

$P(Y = k | X)$ = Probability of data point belonging to class k given the input features X . These are called **posterior probabilities**

$P(X|Y = k)$ = Probability of X given the output class $Y = k$

C = When covariances of all the classes are the same

C_{def} = cost of someone being a defaulter

C_{lost} = cost of losing a customer

$\hat{\pi}_k = \frac{\text{Number of samples of class } k}{\text{Total number of samples}} = \text{Estimate of } \pi_k$

$\hat{\mu}_k$ = Estimate of μ_k

\hat{C}_k = Estimate of C_k

$\hat{\theta}$ = The weight vector

$L(\text{data}; \theta)$ = Likelihood function of the observed data

γ = Regularization hyperparameter

\mathbf{w} = Weight in the likelihood equation for unbalanced data

$\|x\|$ = Distance of point x from the origin

w = Weight in the distance equation/formula