Quadratic Discriminant Analysis

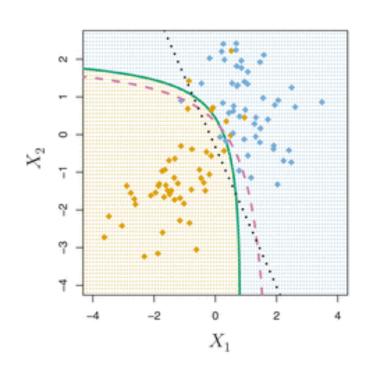






Quadratic Discriminant Analysis (QDA): Intuition

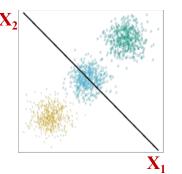
- QDA is simply using a non-linear discriminant function
- The graph illustrates a simple model with an outcome variable Y with K=2 classes and p=2 predictors X₁ and X₂
- The dotted line shows the LDA discriminant function
- The QDA method tries various quadratic discriminant functions (e.g., dotted curve) and selects the one that provides the most accurate classification
- As with polynomial regression models QDA has better training fit, more overfitting problems, and more variance
- But it is a more flexible classifier



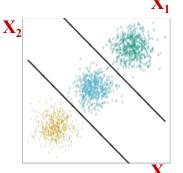


LDA Illustration

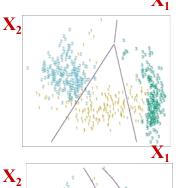




Logistic regression model with 3 categorical outcomes (color coded) variable and 2 predictors → observations in the middle category are not well differentiated



Same data, model estimated with LDA → the linear boundaries do a better job at classifying the observations



Less differentiate data with LDA model → the linear boundaries do an OK job at classifying observations but a few points are misclassified

Same data with Quadratic DA model → does better at classifying observations, but the flexibility of the curve makes the model more over-fitted with higher variance





lda() {MASS} → "Linear Discriminant Analysis" function in the {MASS}
"Modern Applied Statistics with S" package to fit LDA models

lda.fit=lda($y\sim x1+x2+etc.$, data=dataName) \rightarrow Fits an LDA model

qda () {MASS} \rightarrow "Quadratic Discriminant Analysis" function in the {MASS} "Modern Applied Statistics with S" package to fit QDA models. The syntax is identical to the lda () function.

lda.fit=qda($y\sim x1+x2+etc.$, data=dataName) \rightarrow Fits a QDA model





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