

# MATH 569      Statistical Learning

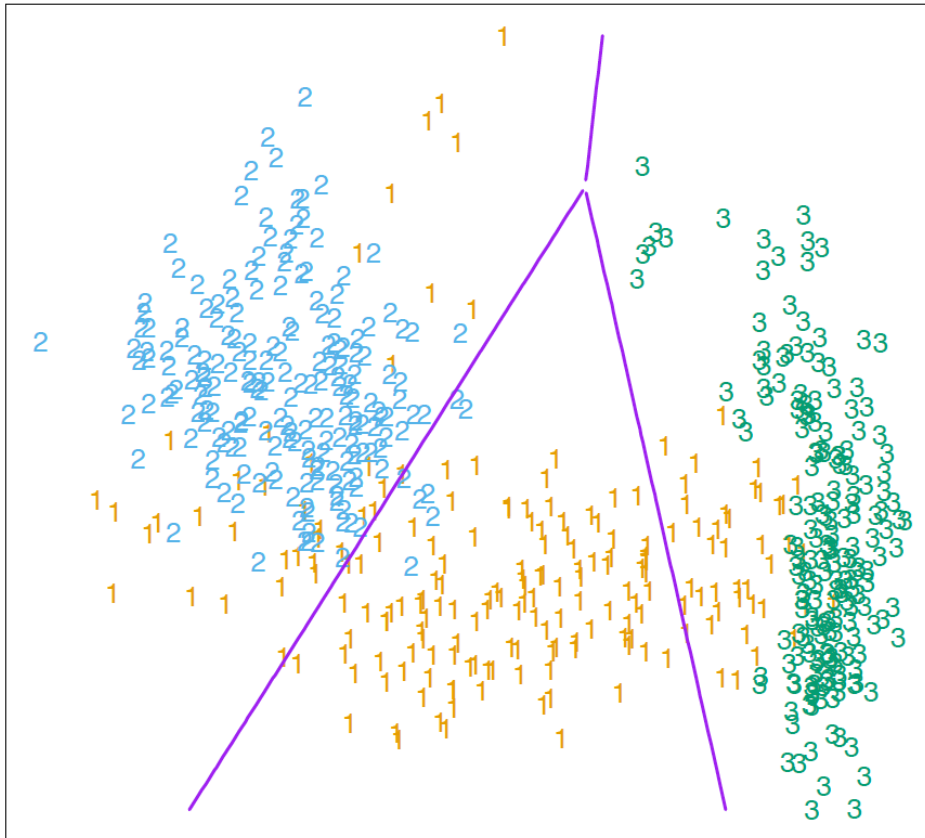
## Part III: Linear Methods of Classification

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Fig 4.1 Linear boundary vs quadratic boundary  
(found by LDA)

Linear boundary

Using only  $X_1$  and  $X_2$



Quadratic boundary

Using  $X_1, X_2, X_1X_2, X_1^2, X_2^2$

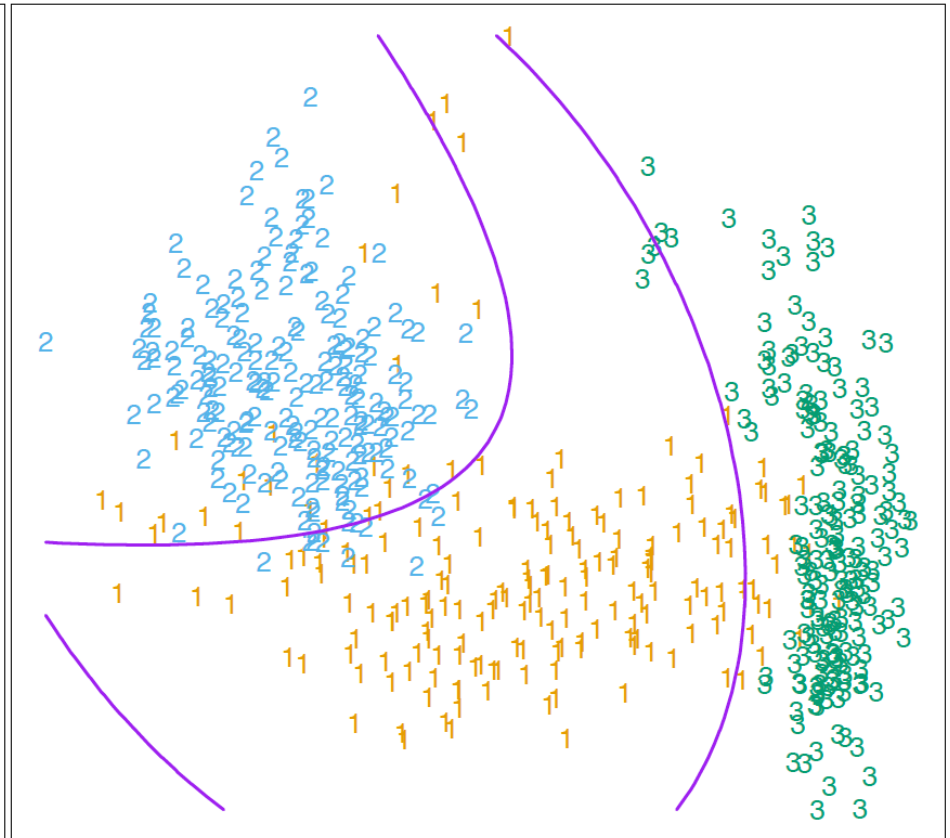
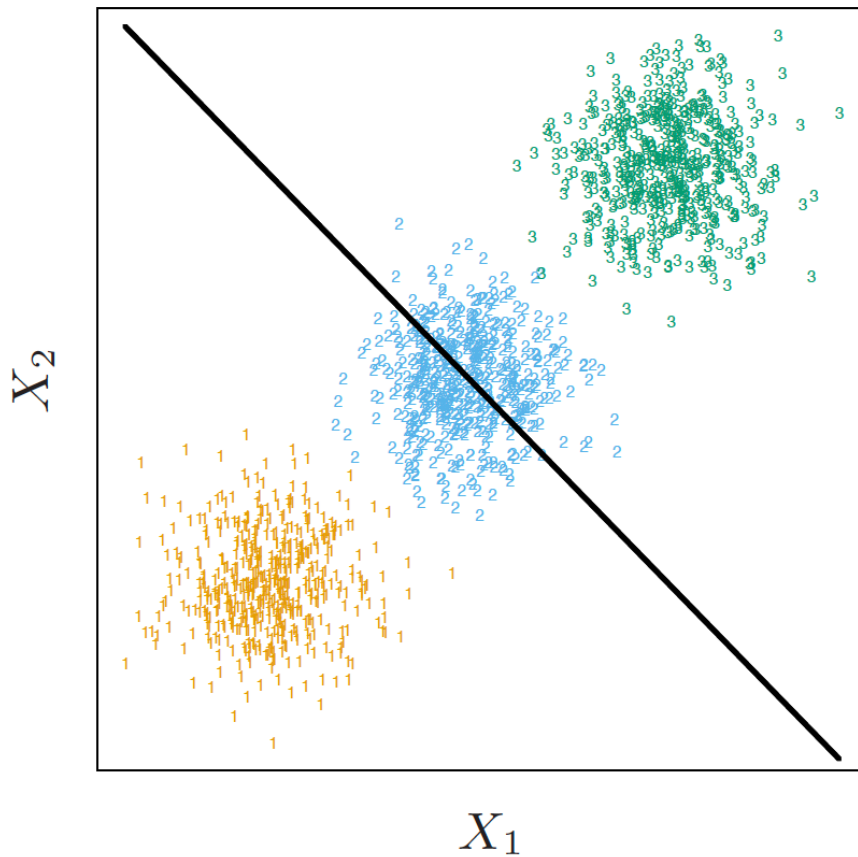


Fig 4.2 Masking effect for linear regression  
Three classes in a 2-dimensional space ( $K=3$ ,  $p=2$ )

Linear Regression



Linear Discriminant Analysis

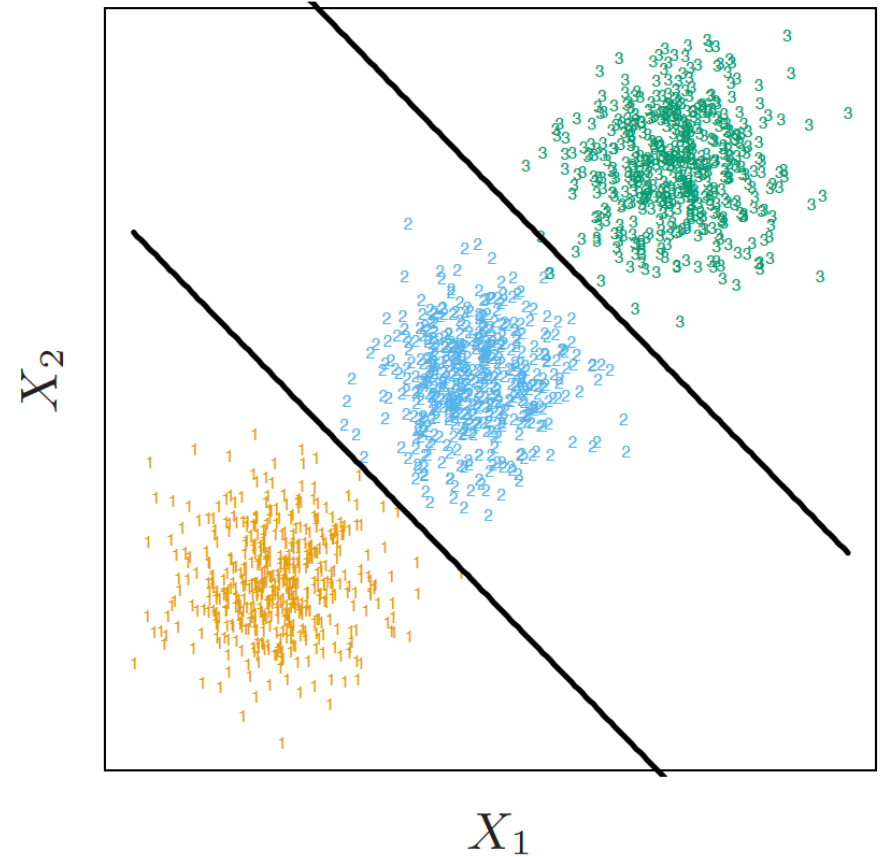
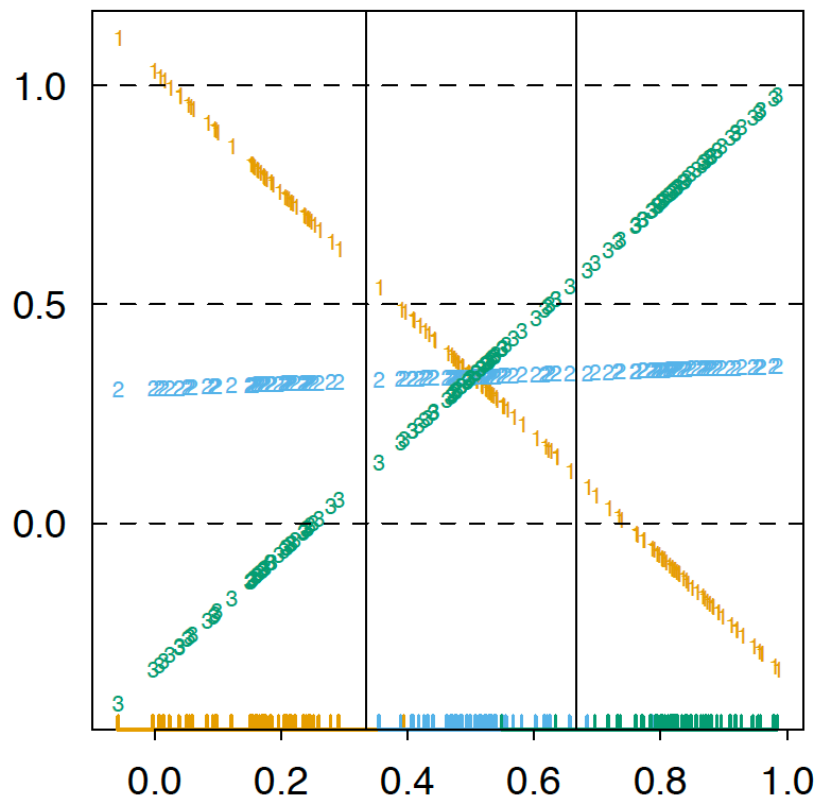


Fig 4.3 Use higher degree polynomials to fix the masking effect of linear regression

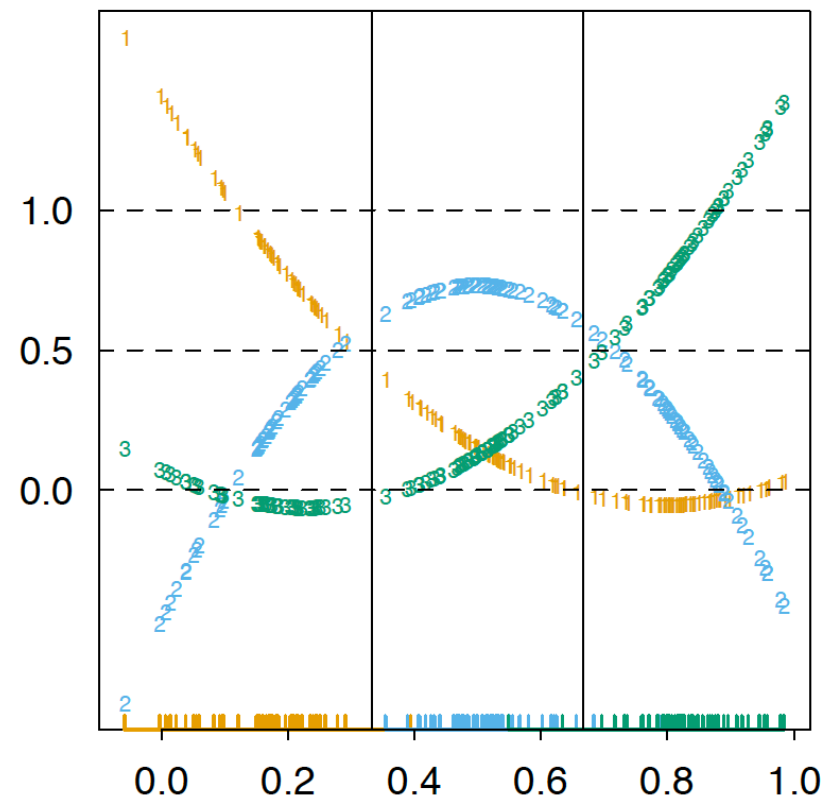
Using linear regression on linear terms,  
the middle class never dominates.  
Class 2 is masked by class 1 and class 3.

Using quadratic terms,  
class 2 is no longer masked.

Degree = 1; Error = 0.33

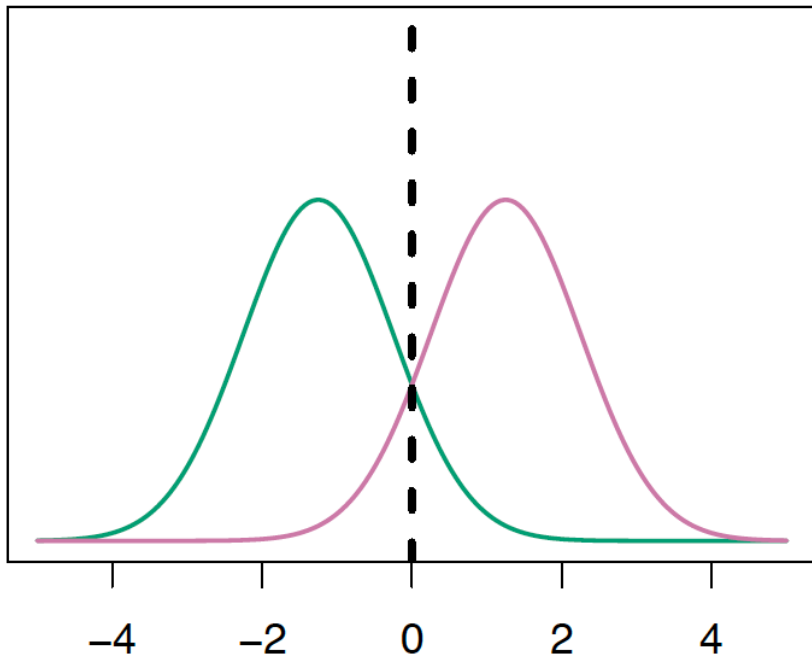


Degree = 2; Error = 0.04



## Bayes Decision Boundary

$$\pi_1 = 0.5, \pi_2 = 0.5$$



$$\pi_1 = 0.3, \pi_2 = 0.7$$

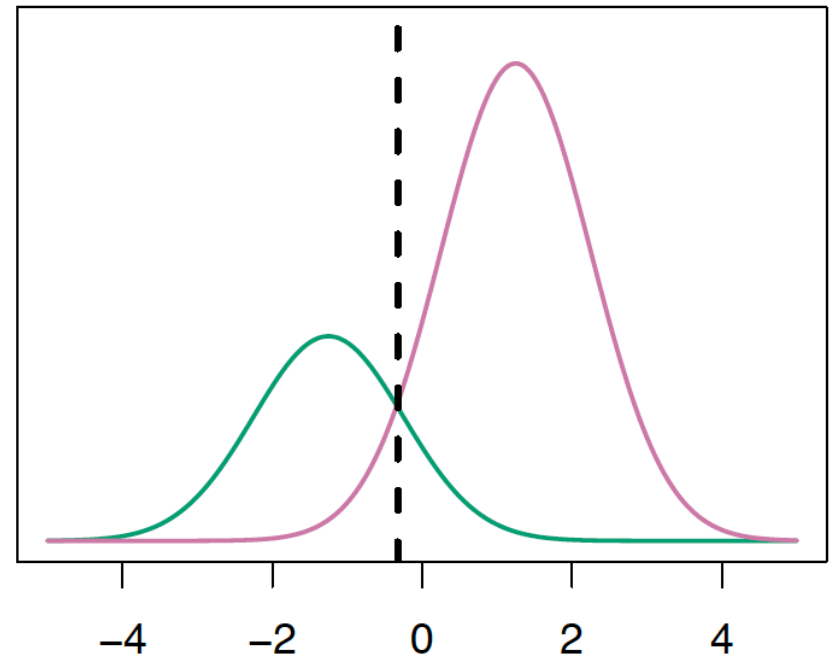


Fig 4.6 LDA and QDA both can find quadratic boundaries

Using LDA on 5 dimensional space

$X_1, X_2, X_1X_2, X_1^2, X_2^2$

Using QDA

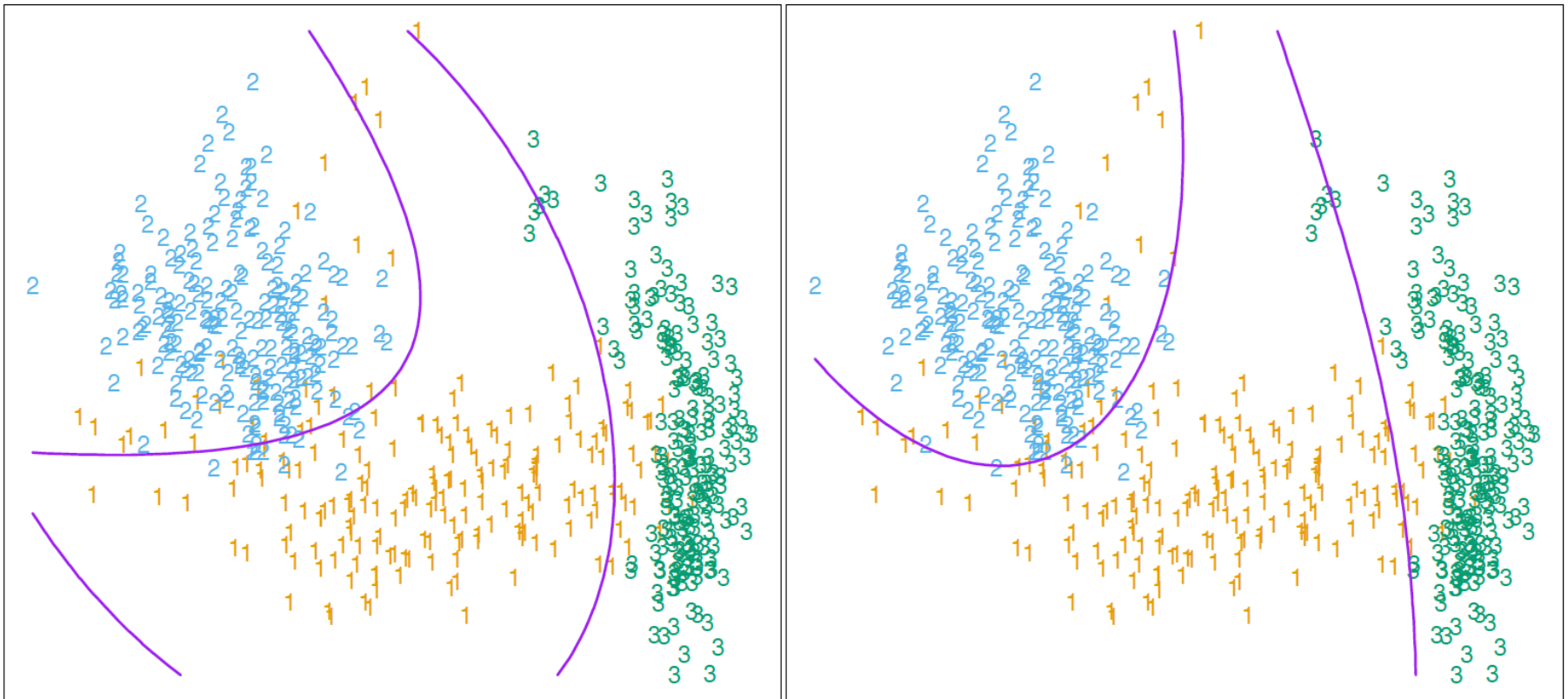
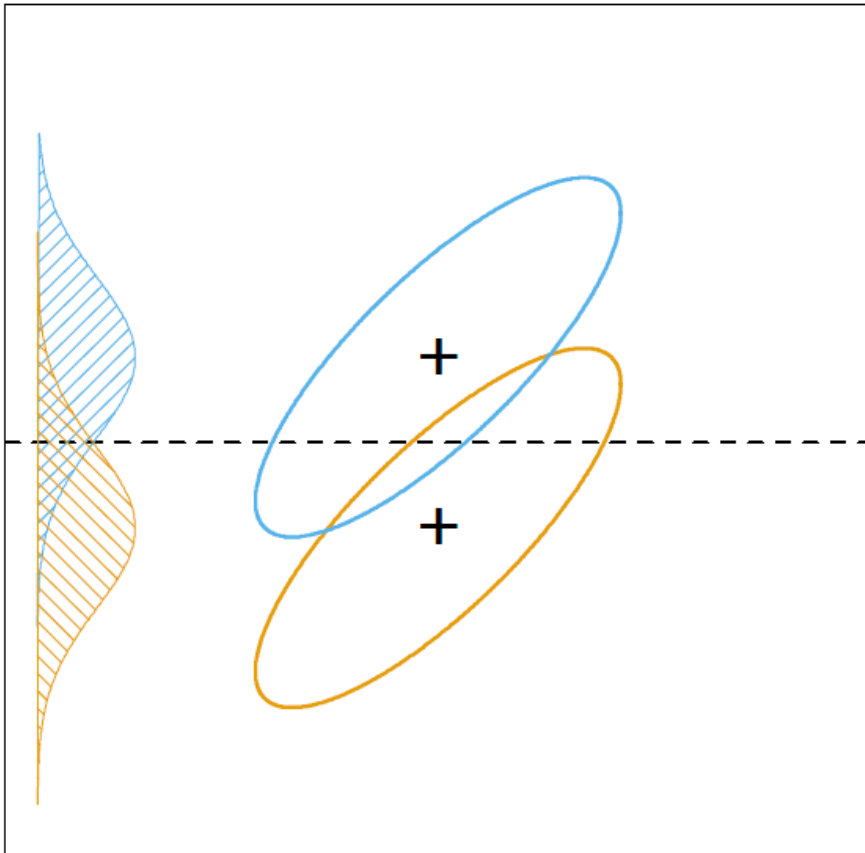


Fig 4.9 Discriminant Direction

The direction connecting two centroids may not minimize overlap



The discriminant direction minimizes this overlap for Gaussian data

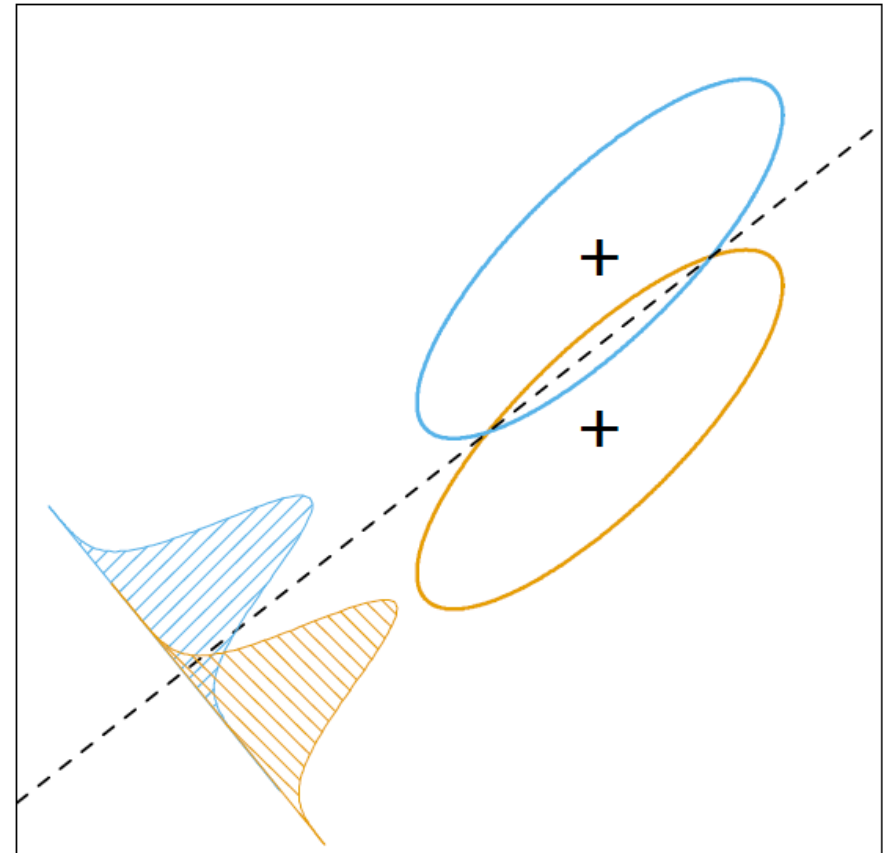


Fig 4.11 Classification in the two-dimensional subspace  
spanned by the first two canonical variates  
(Vowel training data:  $K=11$ ,  $p=10$ )

