

# Generalizing Linear Discriminant Analysis

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# Linear Discriminant Analysis

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## **Objective**

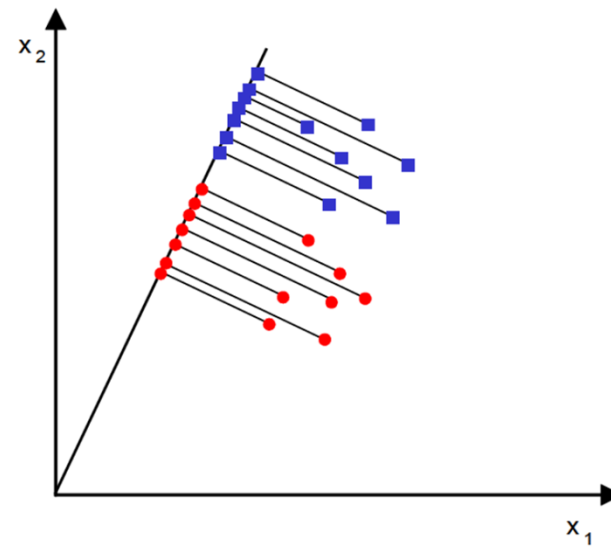
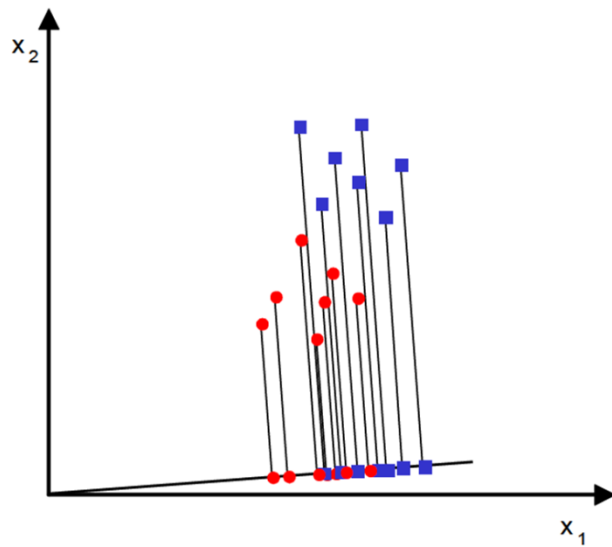
- Project a feature space (a dataset  $n$ -dimensional samples) onto a smaller
- Maintain the class separation

## **Reason**

- Reduce computational costs
  - Minimize overfitting
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- A solid blue horizontal bar spanning the width of the slide, located at the bottom.

# Linear Discriminant Analysis

Want to reduce dimensionality while preserving ability to discriminate



Figures from [1]

# Linear Discriminant Analysis

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Could just look at means and find dimension that separates means most:

$$\mu_i = \frac{1}{N_i} \sum_{x \in \omega_i} x \text{ and } \tilde{\mu}_i = \frac{1}{N_i} \sum_{y \in \omega_i} y = \frac{1}{N_i} \sum_{x \in \omega_i} w^T x = w^T \mu_i$$

Equation from [1]

# Linear Discriminant Analysis

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$$J(w) = |\tilde{\mu}_1 - \tilde{\mu}_2| = |w^T (\mu_1 - \mu_2)|$$

Equations from [1]

# Linear Discriminant Analysis

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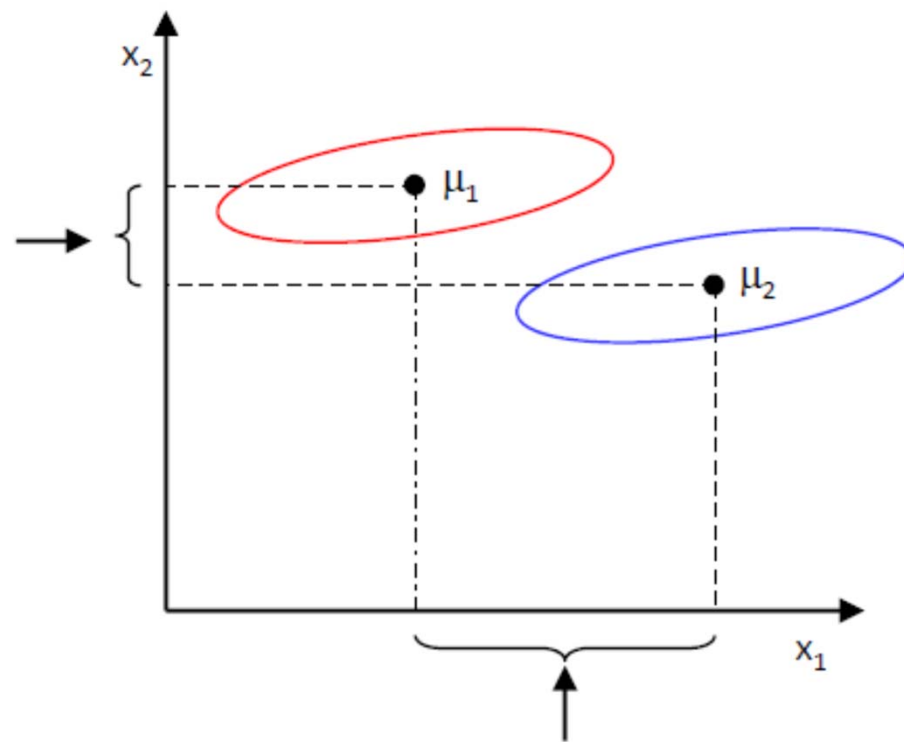


Figure from [1]