Linear Classifiers II

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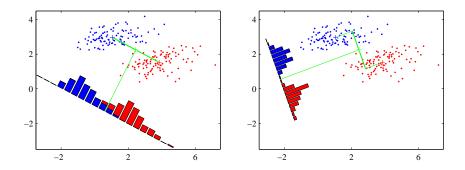
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Linear Classifiers - supplement lecture

- ➤ Supplement to the lecture about learning Linear Classifiers (perceptron, . . .)
- ▶ Better etalons by applying Fischer linear discriminator analysis.
- ► LSQ formulation of the learning task.

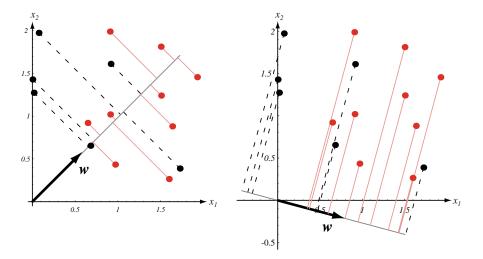
Fischer linear discriminant



- **▶** Dimensionality reduction
- ► Maximize distance between means, ...
- ▶ ... and minimize within class variance. (minimize overlap)

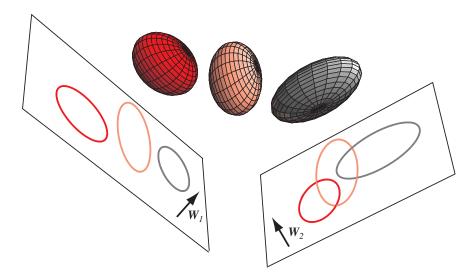
Figures from [1]

Projections to lower dimensions $y = \mathbf{w}^{\mathsf{T}} \mathbf{x}$



Figures from [2]

Projections to lower dimensions $y = \mathbf{w}^{\mathsf{T}} \mathbf{x}$



Figures from [2]

Finding the best projection

 $y = \mathbf{w}^{\top} \mathbf{x}$

 $\mathbf{m}_i = \frac{1}{n_i} \sum_{\mathbf{x} \in C} \mathbf{x}$

thresholding $y \ge -w_0$ C_1 , otherwise C_2

$$\mathit{m}_2 - \mathit{m}_1 = \mathbf{w}^{ op} (\mathbf{m}_2 - \mathbf{m}_1)$$

Within class scatter of projected samples

$$egin{aligned} \mathbf{m}_2 &= \mathbf{m}_1 \ \end{aligned}$$
 projected samples $m_i = m_i \$

ojected samples
$$m_i$$
)²

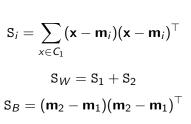
$$s_i^2 = \sum_{y \in C_i} (y - m_i)^2$$

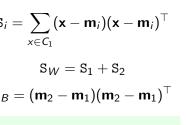
$$s_i^2 = \sum_{y \in C_i} (y - m_i)^2$$

 $J(\mathbf{w}) = \frac{(m_2 - m_1)^2}{s_r^2 + s_r^2}$

$$s_i^2 = \sum_{y \in C_i} (y - m_i)^2$$
Fischer criterion:

$$S_B$$





S_B stands for the *between* class scatter matrix

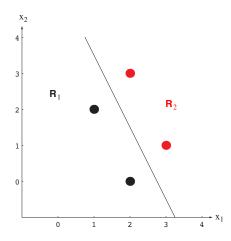
After some derivation on blackboard:

 $\mathbf{w} = S_{M}^{-1}(\mathbf{m}_2 - \mathbf{m}_1)$



LSQ approach to linear classification





Linear least squares not guaranteed to correctly classify everything on the training set. It's objective function not perfect for classification.

Outliers can shift the decision boundary.

References I

Further reading: Chapter 4 of [1], or chapter 3 and 5 of [2].

[1] Christopher M. Bishop.

Pattern Recognition and Machine Learning.

Springer Science+Bussiness Media, New York, NY, 2006.

PDF freely downloadable.

[2] Richard O. Duda, Peter E. Hart, and David G. Stork.

Pattern Classification.

John Wiley & Sons, 2nd edition, 2001.