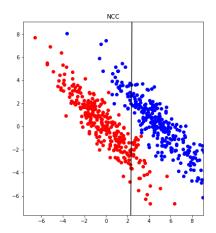
Cognitive Algorithms: Tutorial 2 Linear Discriminant Analysis

Joanina, Ken, Augustin

Problems with NCC



Content

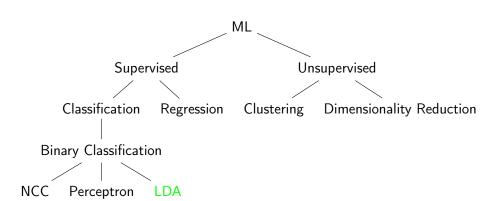
LDA

Recap: Statistics

LDA

Whitening

The Tree of CA



Random variables

Formalization of objects or quantities that depend on randomness

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$$X:\Omega \to \mathbb{R}$$

• Expected Value

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- Variance

- Variance
- Covariance

Covariance Matrix

• A set of d random variables has a covariance matrix:

$$\begin{bmatrix} \mathsf{Cov}(X_1, X_1) & \dots & \mathsf{Cov}(X_1, X_d) \\ \vdots & \ddots & \vdots \\ \mathsf{Cov}(X_d, X_1) & \dots & \mathsf{Cov}(X_d, X_d) \end{bmatrix}$$

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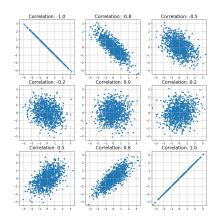
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$$\xrightarrow{\mathsf{Task} \ 1.2} \begin{bmatrix} ? & ? \\ ? & ? \end{bmatrix}$$

- Variance
- Covariance

→ How does this relate to the real world?

Correlation visualized (for Gaussians)



Positive semi-definite matrices

ullet A Matrix $\mathbf{A} \in \mathbb{R}^{d imes d}$ is positive semi-definite iff

$$x^T A x \ge 0$$
 for all $x \in \mathbb{R}^d$

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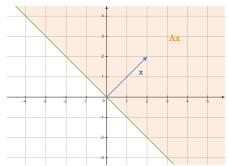
$$x^T A x \ge 0$$
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What does this mean?

$$\mathbf{x}^\mathsf{T} \mathbf{A} \mathbf{x} = \langle \mathbf{x}, \mathbf{A} \mathbf{x} \rangle = \underbrace{\| \mathbf{x} \|}_{\geq 0} \underbrace{\| \mathbf{A} \mathbf{x} \|}_{\geq 0} \cos(\alpha)$$

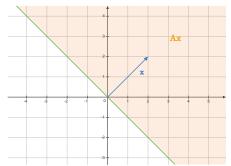
Positive semi-definite matrices II

• Ax is in the half space of x or orthogonal to x!



Positive semi-definite matrices II

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• A symmetric matrix is PSD iff all eigenvalues are non-negative

Task 3 - PSD: Pretty Sweet, Dude!

We care in particular about PSD matrices because all covariance matrices are PSD. One interesting property about PSD matrices is that all of their eigenvalues are non-negative. In particular it holds that all real, symmetric matrices A with non-negative eigenvalues are PSD, and vice versa. Prove this statement in both directions!

Task: Possible or not

• Given three 1-dimensional random variables X, Y and Z with the following properties:

$$Var(X) = Var(Y) = Var(Z) = 10, Cov(X, Y) = 8,$$

$$Cov(X, Z) = 1, Cov(Y, Z) = 8$$

• Is this possible?

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$$C = \begin{bmatrix} 10 & 8 & 1 \\ 8 & 10 & 8 \\ 1 & 8 & 10 \end{bmatrix}$$

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• Show that C is not positive semi-definite!

Method 1: Quadratic form

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$$\lambda(C) \approx \{-0.82, 9, 21.82\}$$

Method 2: Eigenvalues

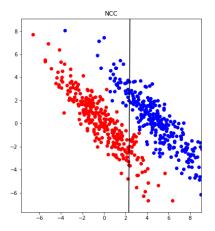
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• Works (well) for proving and disproving



Problems with NCC



The Fisher criterion

$$w^* = \underset{w}{\text{arg\,max}} \; \frac{w^\mathsf{T} S_\mathrm{b} w}{w^\mathsf{T} S_\mathrm{w} w}$$

Orthogonal Matrices

orthonormal column vectors

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- $U^T = U^{-1}$

Eigendecomposition

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⇒ Always possible if matrix is real and symmetric!