

The Why and How of Nonnegative Matrix Factorization

Topic Presentation

Group 02

LINMA2380 — Matrix computations

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Summary

1 Applications

2 Connections to other problems

Applications - Image processing


Goal : Facial Feature Extraction





Data matrix : $X \in \mathbb{R}_+^{p \times n}$

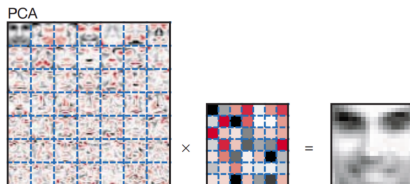
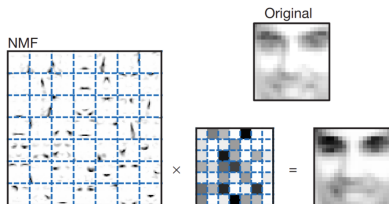
- p : total number of pixels
- n : number of faces
- $X(i, j)$: the gray-level of the i -th pixel in the j -th face

Applications - Image processing

$$\underbrace{X(:,j)}_{\text{jth facial image}} \approx \sum_{k=1}^r \underbrace{W(:,k)}_{\text{facial features}}$$


$$\underbrace{H(k,j)}_{\text{importance of features in jth image}} = \underbrace{WH(:,j)}_{\text{approximation of jth image}}$$



Applications - Image processing



Applications - Text mining

Goal : Topic Recovery and Document Classification

Data matrix : $X \in \mathbb{R}^{n \times m}$

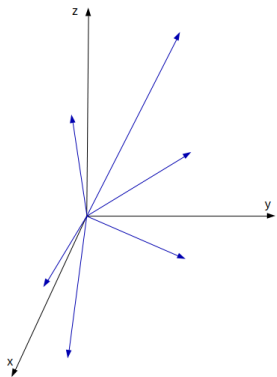
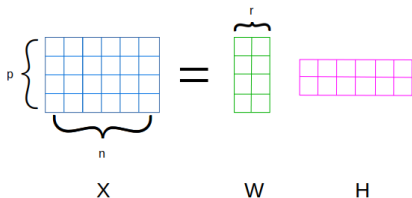
- each column : a document
- each line : a word
- $X(i, j)$: number of times the i -th word appears in the j -th document

$$\underbrace{X(:, j)}_{j\text{th document}} \approx \sum_{k=1}^r \underbrace{W(:, k)}_{k\text{th topic}} \underbrace{H(k, j)}_{\substack{\text{importance of } k\text{th topic} \\ \text{in } j\text{th document}}}, \quad \text{with } W \geq 0 \text{ and } H \geq 0.$$

Nonnegative rank

Definition (Nonnegative rank)

Given $X \in \mathbb{R}_+^{p \times n}$, the nonnegative rank of X , denoted $\text{rank}_+(X)$ is the minimum r s.t. $\exists W \in \mathbb{R}_+^{p \times r}, H \in \mathbb{R}_+^{r \times n}$ with $X = WH$.



Computational Geometry : Nested polytopes problem

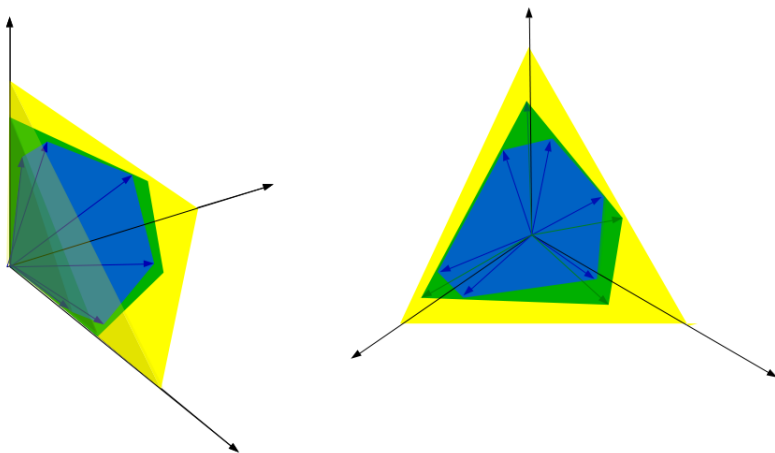


Figure: Finding a polytope with minimum nb of vertices nested between 2 polytopes

Graph Theory : Bipartite dimension

Let $G(X) = (V_1 \cup V_2, E)$ be a bipartite graph induced by X (i.e. $(i, j) \in E \Leftrightarrow X_{ij} \neq 0$).

Definition (Biclique and bipartite dimension)

- A *biclique* (or a *complete bipartite graph*) is a bipartite graph s.t. every vertex in V_1 is connected to every vertex in V_2 .
- The *bipartite dimension* (or the *minimum biclique cover*) $bc(G(X))$ is the minimum number of bicliques needed to cover all edges in E .

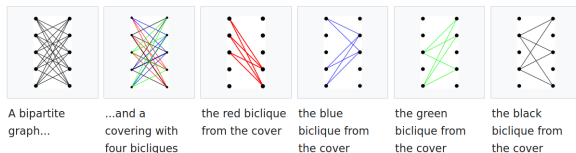


Figure: Example for biclique edge cover [biclique]



