

# How to assess quality of BMF algorithms?

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# Motivation

- Boolean matrix factorization (BMF).
- Method for analysis of Boolean data.
- Various algorithms (more than 25).
- How to assess their quality?

# Boolean Matrix Factorization

- A general aim: for a given matrix  $I \in \{0, 1\}^{n \times m}$  find matrices  $A \in \{0, 1\}^{n \times k}$  and  $B \in \{0, 1\}^{k \times m}$  for which  $I$  (approximately) equals  $A \circ B$
- $\circ$  is the Boolean matrix product

$$(A \circ B)_{ij} = \max_{l=1}^k \min(A_{il}, B_{lj}).$$

$$\begin{pmatrix} 10111 \\ 01101 \\ 01001 \\ 10110 \end{pmatrix} = \begin{pmatrix} 110 \\ 011 \\ 001 \\ 100 \end{pmatrix} \circ \begin{pmatrix} 10110 \\ 00101 \\ 01001 \end{pmatrix}$$

- Discovery of  $k$  factors that exactly or approximately explain the data.
- Factors = interesting patterns in data.

# Computational Complexity

- Basic feature of each algorithm.
- We prefer algorithm with the smaller complexity.
- Big O notation (hides several issues).
- Better way: relative time complexity.
- “One algorithm is three-times faster than other.”
- Time (and space) complexity is not critical issue for the most of current algorithms.
- Runnable on ordinary PC.

## Approximation Factor

- Optimization version of the basic decomposition problem is NP-hard.
- No polynomial time algorithm (computing exact solution) exists.
- Based on heuristic - approximation factor.
- Recent results on approximability: basic decomposition problem is NP-hard to approximate within factor  $n^{1-\epsilon}$ .
- Lower bound is not encouraging.
- Current algorithm produce much better results.

# Quality of Factors

- 1 Geometry of factorization
- 2 Interpretability of individual factors
  - Knowledge discovery view
- 3 Quality of a set of extracted factors
  - Reduction of dimensionality
  - Explanatory view

# Explanation of Data by Factors

- Distance (error function)

$$E(C, D) = ||C - D|| = \sum_{i,j=1}^{m,n} |C_{ij} - D_{ij}|.$$

- Two components of  $E$

$$E(I, A \circ B) = E_u(I, A \circ B) + E_o(I, A \circ B), \text{ where}$$

$$E_u(I, A \circ B) = |\{\langle i, j \rangle ; I_{ij} = 1, (A \circ B)_{ij} = 0\}|,$$

$$E_o(I, A \circ B) = |\{\langle i, j \rangle ; I_{ij} = 0, (A \circ B)_{ij} = 1\}|.$$

- Coverage quality

$$c(l) = 1 - E(I, A \circ B) / ||I||.$$

# Two Basic Viewpoint to BMF

## ■ Discrete Basis Problem

- Given  $I \in \{0, 1\}^{n \times m}$  and a positive integer  $k$ , find  $A \in \{0, 1\}^{n \times k}$  and  $B \in \{0, 1\}^{k \times m}$  that minimize  $\|I - A \circ B\|$ .
- Emphasizes the importance of the first few (presumably most important) factors.
- *Miettinen P., Mielikainen T., Gionis A., Das G., Mannila H., The discrete basis problem, IEEE Transactional Knowledge and Data Engineering 20(10)(2008), 1348–1362*

## ■ Approximate Factorization Problem

- Given  $I$  and prescribed error  $\varepsilon \geq 0$ , find  $A \in \{0, 1\}^{n \times k}$  and  $B \in \{0, 1\}^{k \times m}$  with  $k$  as small as possible such that  $\|I - A \circ B\| \leq \varepsilon$ .
- Emphasizes the need to account for (and thus to explain) a prescribed (presumably reasonably large) portion of data.
- *Belohlavek R., Trnecka M., From-below approximations in Boolean matrix factorization: Geometry and new algorithm, Journal of Computer and System Science 81(8)(2015), 1678–1697.*

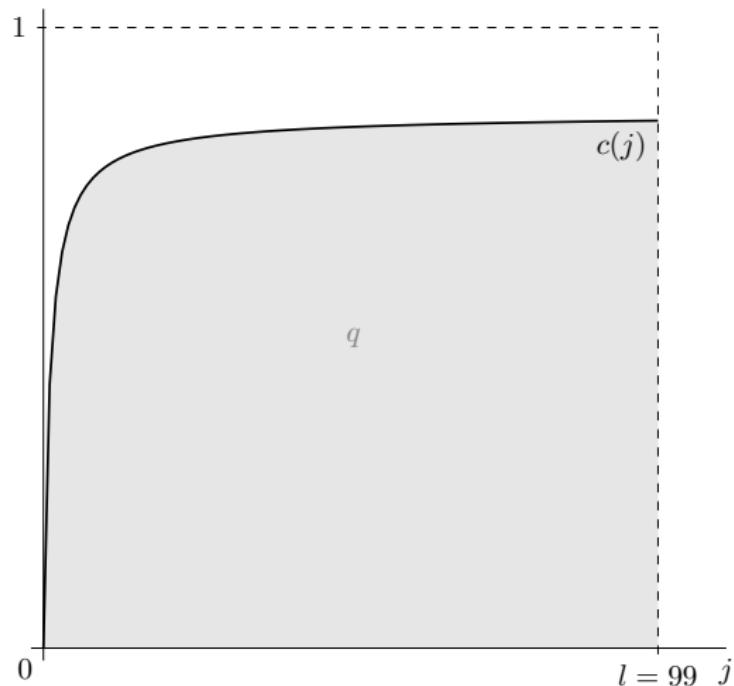
## Quality Measure

- $w_l = l/k$  for the DBP view
- $w_l = 1 + (E(I, A \circ B) - \varepsilon)/(||I|| - \varepsilon)$  for AFP view
- $w_l = (l/k + 1 + (E(I, A \circ B) - \varepsilon)/(||I|| - \varepsilon))/2$  combined view.

$$q = 1 - \left( \sum_{j=0}^l w_j \frac{E(I, A \circ B)}{||I||} \right) / \left( \sum_{j=0}^l w_j \right).$$

- Reflect natural requirement for a good decomposition.

## Interpretation



**Figure:** Measure of quality of BMF algorithm

## Experimental Evaluation

- ASO—Miettinen P., Mielikainen T., Gionis A., Das G., Mannila H., *The discrete basis problem*, *IEEE Transactional Knowledge and Data Engineering* 20(10)(2008), 1348–1362.
- GRECOND—Belohlavek R., Vychodil V., *Discovery of optimal factors in binary data via a novel method of matrix decomposition*, *Journal of Computer and System Science* 76(1)(2010), 3–20.
- NAIVECOL—Ene A. et al., *Fast exact and heuristic methods for role minimization problems*. Proc. SACMAT 2008, pp. 1–10.
- PANDA—Lucchese C., Orlando S., Perego R., *Mining top-K patterns from binary datasets in presence of noise*, *SIAM DM* 2010, pp. 165–176.
- HYPER—Xiang Y., Jin R., Fuhr D., Dragan F. F., *Summarizing transactional databases with overlapped hyperrectangles*, *Data Mining and Knowledge Discovery* 23(2011), 215–251
- GREESS—Belohlavek R., Trnecka M., *From-below approximations in Boolean matrix factorization: Geometry and new algorithm*, *Journal of Computer and System Science* 81(8)(2015), 1678–1697.

# Results

Table: Numbers of factors and coverage quality

Dataset		ASSO	GRECOND	NAIVECOL	HYPER	PANDA	GREESS
Mushroom	$c = 80\%$	19	29	32	42	NA	31
	$c = 90\%$	34	46	47	57	NA	47
	$c = 95\%$	50	62	62	70	NA	61
	$c = 100\%$	NA	120	110	123	NA	105
	$k = 10$	0.556	0.582	0.512	0.285	0.346	0.546
	$k = 20$	0.652	0.715	0.674	0.502	0.346	0.696
	$k = 30$	0.720	0.812	0.789	0.664	0.346	0.793
	$k = 40$	0.765	0.873	0.862	0.780	0.346	0.865

# Results

Table: BMF algorithm quality

Dataset		ASSO	GRECOND	NAIVECOL	HYPER	PANDA	GREESS
Mushroom	$q_{0.8}$	0.622	0.740	0.729	0.657	0.344	0.733
	$q_{0.9}$	0.695	0.801	0.786	0.709	0.344	0.794
	$q_{0.95}$	0.725	0.827	0.810	0.728	0.344	0.819
	$q_1$	0.745	0.844	0.827	0.749	0.344	0.835
	$q_{10}$	0.556	0.582	0.511	0.285	0.346	0.545
	$q_{20}$	0.650	0.712	0.671	0.498	0.346	0.693
	$q_{30}$	0.715	0.805	0.781	0.654	0.346	0.786
	$q_{40}$	0.756	0.861	0.848	0.760	0.346	0.851
	$q_{10,0.9}$	0.764	0.876	0.863	0.798	0.344	0.870
	$q_{20,0.8}$	0.763	0.874	0.860	0.792	0.344	0.867

# Conclusion

- We point out an important problem in BMF: assessment of quality of BMF algorithms.
- We identify key aspects of such assessment.
- We propose quantitative ways how to measure quality of BMF algorithms.

# Thank you