

Vector Space Model: Distance Functions

(Teaching Inspired by Research)

Prof. Dr. Marcin Grzegorzek and
the Medical Data Science Team



UNIVERSITÄT ZU LÜBECK
INSTITUT FÜR MEDIZINISCHE INFORMATIK


- ① Introduction
- ② Distance Functions for Real Points
- ③ Distance Functions for Binary Points
- ④ Distance Functions for Sequences
- ⑤ Conclusion

- 1 Introduction
- 2 Distance Functions for Real Points
- 3 Distance Functions for Binary Points
- 4 Distance Functions for Sequences
- 5 Conclusion

Contents of the Course

Week	Lecture	Practical Exercises
1	(05/04) Introduction to Medical Information Retrieval (MIR)	(05/04) Introduction to Python
2	(12/04) Main Components and Classification of MIR Systems	(12/04) Introduction to Python
3	(19/04) Metadata in Medical Information Retrieval Systems	(19/04) CBIR in Medical Applications
4	(26/04) No Lecture due to a Business Trip	(26/04) CBIR in Medical Applications
5	(03/05) Set Theoretic Model: Boolean Retrieval	(03/05) CBIR in Medical Applications
6	(10/05) Set Theoretic Model: Fuzzy Retrieval	(10/05) Flask Tutorial
7	(17/05) Vector Space Model: Similarity Measures	(17/05) Flask Tutorial

Contents of the Course

8	(24/05) Vector Space Model: Distance Functions	(24/05) HTML
9	(31/05) Vector Space Model: Latent Semantic Indexing	(31/05) HTML
10	(07/06) Probabilistic Model	(07/06) HTML
11	(14/06) Text-based Retrieval of Medical Information	(14/06) Deep Learning
12	(21/06) Audio-based Retrieval of Medical Information	(21/06) Deep Learning
13	(28/06) Image-based Retrieval of Medical Information	(28/06) Relevance Feedback
14	(05/07) Demonstrators from Current Research Projects	(05/07) Relevance Feedback
15	(12/07) Summary and Conclusions	(12/07) Evaluation 

Generally about Distance Functions

A metric on a set \mathbb{R}^I is a distance function

$$d : \mathbb{R}^I \times \mathbb{R}^I \longrightarrow [0, \infty) \quad ,$$

if for all $\mathbf{x}, \mathbf{y}, \mathbf{z} \in \mathbb{R}^I$ all of the following conditions are satisfied:

(1) $d(\mathbf{x}, \mathbf{y}) \geq 0$;

(2) $d(\mathbf{x}, \mathbf{y}) = 0 \Leftrightarrow \mathbf{x} = \mathbf{y}$;

(3) $d(\mathbf{x}, \mathbf{y}) = d(\mathbf{y}, \mathbf{x})$;

(4) $d(\mathbf{x}, \mathbf{z}) \leq d(\mathbf{x}, \mathbf{y}) + d(\mathbf{y}, \mathbf{z})$.

- 1 Introduction
- 2 Distance Functions for Real Points**
- 3 Distance Functions for Binary Points
- 4 Distance Functions for Sequences
- 5 Conclusion

Minkowski Distance – General Form

A popular metric extensively used for IR is the Minkowski distance:

$$d(\mathbf{x}, \mathbf{y}) = \left(\sum_{i=1}^I \omega_i |x_i - y_i|^p \right)^{\frac{1}{p}} .$$

Minkowski Distance – Examples

Selected examples of the Minkowski distance for different values of p and $\omega_{i=1,\dots,l} = 1$:

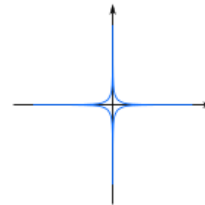
$$p = 1 \quad \Rightarrow \quad \sum_{i=1}^l |x_i - y_i|$$

$$\lim_{p \rightarrow \infty} \left(\sum_{i=1}^l |x_i - y_i|^p \right)^{\frac{1}{p}} = \max_{i=1,\dots,l} |x_i - y_i|$$

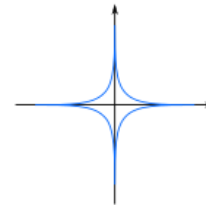
$$\lim_{p \rightarrow -\infty} \left(\sum_{i=1}^l |x_i - y_i|^p \right)^{\frac{1}{p}} = \min_{i=1,\dots,l} |x_i - y_i|$$

Minkowski Distance – Unit Circles

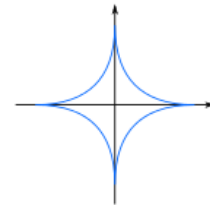
Assuming $\omega_{i=1,\dots,l} = 1$:



$$p = 2^{-2} \\ = 0.25$$



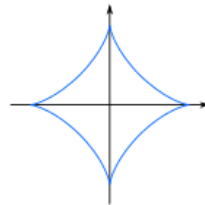
$$p = 2^{-1.5} \\ = 0.354$$



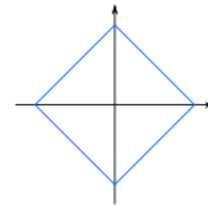
$$p = 2^{-1} \\ = 0.5$$

Minkowski Distance – Unit Circles

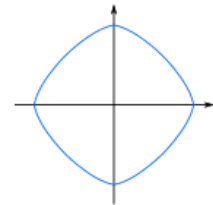
Assuming $\omega_{i=1,\dots,l} = 1$:



$$\begin{aligned} p &= 2^{-0.5} \\ &= 0.707 \end{aligned}$$



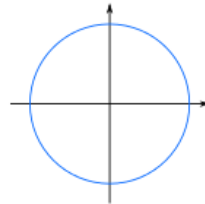
$$\begin{aligned} p &= 2^0 \\ &= 1 \end{aligned}$$



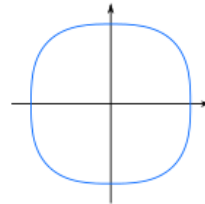
$$\begin{aligned} p &= 2^{0.5} \\ &= 1.414 \end{aligned}$$

Minkowski Distance – Unit Circles

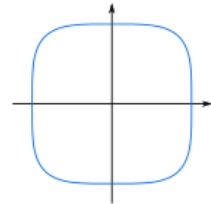
Assuming $\omega_{i=1,\dots,l} = 1$:



$$p = 2^1 \\ = 2$$



$$p = 2^{1.5} \\ = 2.828$$

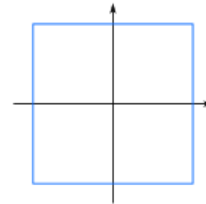


$$p = 2^2 \\ = 4$$

Minkowski Distance – Unit Circles

Assuming $\omega_{i=1,\dots,l} = 1$:

...



$$p = 2^\infty \\ = \infty$$

- 1 Introduction
- 2 Distance Functions for Real Points
- 3 Distance Functions for Binary Points**
- 4 Distance Functions for Sequences
- 5 Conclusion

General Information

- Binary points store information about fulfilling or not fulfilling of properties.
- Graphically, such points can be represented as corner of a hypercube.

Comparison of Properties

$p \in P$	p fulfilled for \mathbf{x}_1	p not fulfilled for \mathbf{x}_1
p fulfilled for \mathbf{x}_2	$n_{1/1}$	$n_{0/1}$
p not fulfilled for \mathbf{x}_2	$n_{1/0}$	$n_{0/0}$

Example:

$$\mathbf{x}_1 = (0, 0, 0, 0, 1, 1, 1, 1)^T \quad \mathbf{x}_2 = (1, 1, 0, 1, 1, 1, 0, 0)^T$$

↓

$$n_{0/0} = 1 \quad n_{0/1} = 3 \quad n_{1/0} = 2 \quad n_{1/1} = 2$$

Minkowski Distance for Binary Points

General form assuming $\omega_{i=1,\dots,l} = 1$:

$$d(\mathbf{x}_1, \mathbf{x}_2) = \left(\sum_{i=1}^l |x_{1,i} - x_{2,i}|^p \right)^{\frac{1}{p}} .$$

For binary points:

$$d = (n_{1/0} + n_{0/1})^{1/p}$$

- 1 Introduction
- 2 Distance Functions for Real Points
- 3 Distance Functions for Binary Points
- 4 Distance Functions for Sequences**
- 5 Conclusion

General Information

- A sequence is just a list of data elements of the same type.
- The number of data elements describing a document can vary.

Earth Mover's Distance – Data Format

- The following data type is considered here:
 $\text{tuple}(\mathbf{x}_i : \text{array}[1 \dots l](\text{real}) ; w_{\mathbf{x}_i} : \text{real}).$
- It describes the i -th element of the sequence \mathbf{X} .

Earth Mover's Distance – Story Behind

- Computing the distance between the sequence \mathbf{X} with m elements and the sequence \mathbf{Y} with n elements, we consider the elements of \mathbf{X} to be mounds and the elements of \mathbf{Y} to be holes in the ground.
- The points \mathbf{x}_i and \mathbf{y}_j can be interpreted as the positions of the mounds and the holes in a l -dimensional space.
- The volumes of the mounds/holes are given by $w_{\mathbf{x}_i}$ and $w_{\mathbf{y}_j}$ respectively.
- The distance between \mathbf{X} and \mathbf{Y} is defined as the minimum cost of transporting the earth from the mounds to the holes.

Earth Mover's Distance – Minimisation of Costs

- Thus, the goal is to minimise the transportation costs.
- A particular constellation of the whole earth transportation process between mounds and holes can be described by a matrix $F = [f_{ij}]$ with f_{ij} standing for the earth volume moved from the mound \mathbf{x}_i into the hole \mathbf{y}_j .
- The overall transportation costs can be now computed as follows:

$$T_{\text{costs}}(\mathbf{X}, \mathbf{Y}, F) = \sum_{i=1}^m \sum_{j=1}^n d(\mathbf{x}_i, \mathbf{y}_j) f_{ij} \quad .$$

Earth Mover's Distance – Final Result

The final distance value:

$$d_{\text{EMD}}(\mathbf{X}, \mathbf{Y}) = \frac{\min_{|f_{ij}|} \left(\sum_{i=1}^m \sum_{j=1}^n d(\mathbf{x}_i, \mathbf{y}_j) f_{ij} \right)}{\sum_{i=1}^m \sum_{j=1}^n f_{ij}} .$$

Earth Mover's Distance

- 1 Introduction
- 2 Distance Functions for Real Points
- 3 Distance Functions for Binary Points
- 4 Distance Functions for Sequences
- 5 Conclusion**

Final Statements

- Most frequently, documents to be compared with each other for retrieval are described by feature vectors of the same dimensionality. In this case, well-known distance functions defined for real points, e.g., the Minkowski distance, can be applied.
- In more heterogeneous IR scenarios, the documents are represented by data structures with less consistency. In this case other methods (e.g., distance functions for sequences) are necessary for their comparison.