



DISTANCE MEASURE

SYRACUSE UNIVERSITY
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DISTANCE MEASURES

Similarity and distance: Two opposite concepts

Similarity measures how close or similar two examples are.

Distance measures how far or different two examples are.

The definitions of **distance functions** are dependent on variable types: numeric, nominal. Many data sets contain mixed types of attributes.

Example: How similar are these two people?

i = (Refund = No, Married, Income = 120K)

j = (Refund = Yes, Married, Income = 90K)

NUMERIC ATTRIBUTES

If the data have all numeric attributes, distance measures can compare the numeric values of the attributes.

Some popular ones include *Minkowski distance*

$$d(i, j) = \sqrt[q]{(|x_{i1} - x_{j1}|^q + |x_{i2} - x_{j2}|^q + \dots + |x_{ip} - x_{jp}|^q)}$$

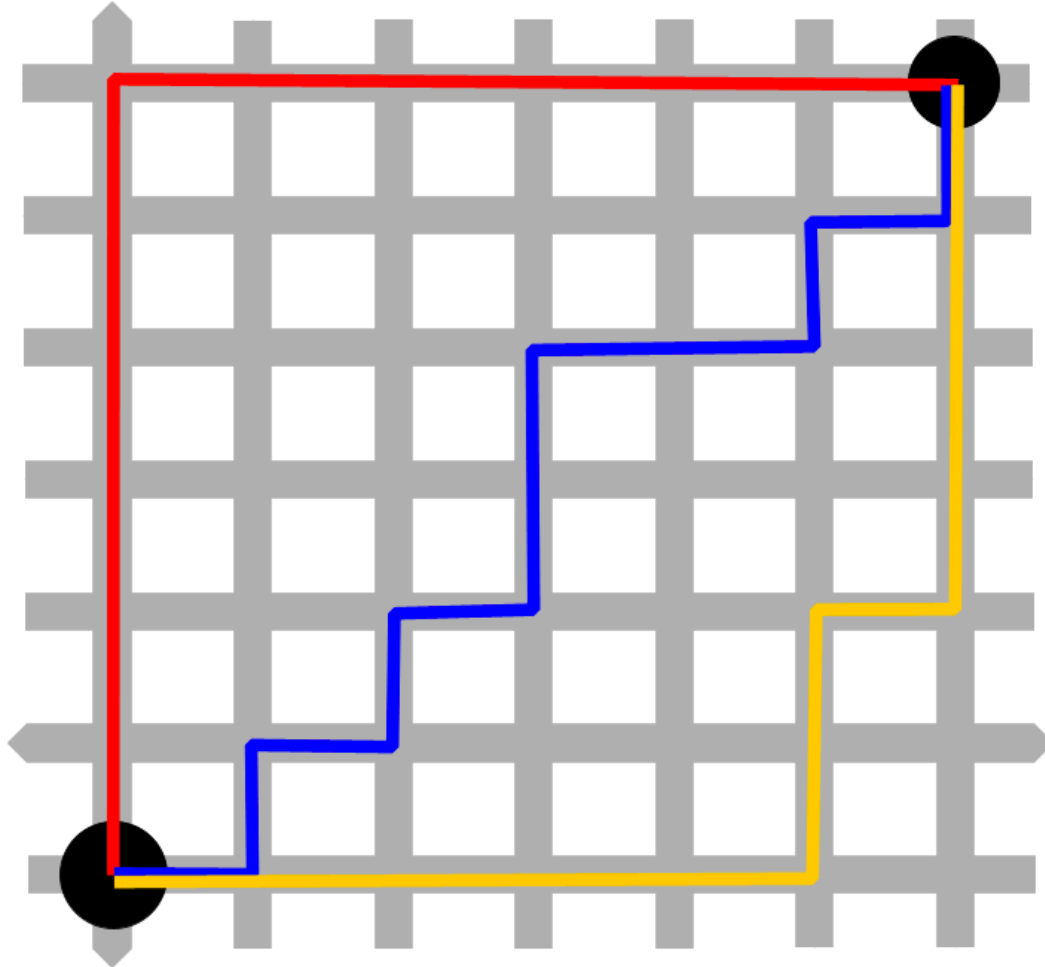
where $i = (x_{i1}, x_{i2}, \dots, x_{ip})$ and $j = (x_{j1}, x_{j2}, \dots, x_{jp})$ are two p -dimensional data instances, and q is a positive integer.

If $q = 1$, d is *Manhattan distance*.

Taking the absolute value of the differences between attribute values

$$d(i, j) = |x_{i1} - x_{j1}| + |x_{i2} - x_{j2}| + \dots + |x_{ip} - x_{jp}|$$

MANHATTAN DISTANCE

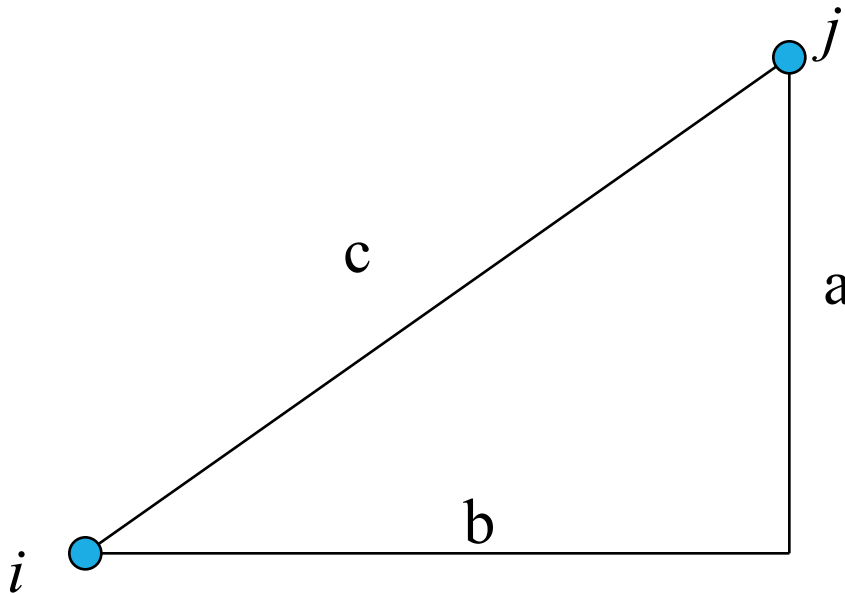


https://upload.wikimedia.org/wikipedia/commons/d/de/Manhattan_distance_bgiu.png

EUCLIDEAN DISTANCE

When $q = 2$, d is *Euclidean distance*:

$$d(i, j) = \sqrt{(|x_{i1} - x_{j1}|^2 + |x_{i2} - x_{j2}|^2 + \dots + |x_{ip} - x_{jp}|^2)}$$



$$d_1(i, j) = a + b$$

$$d_2(i, j) = \sqrt{a^2 + b^2} = c$$

PROPERTIES OF DISTANCE MEASURE

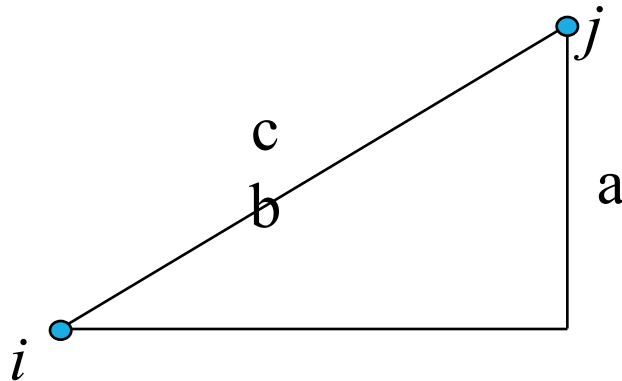
A distance measure should satisfy the following requirements:

$d(i, j) \geq 0$ (nonnegative value)

$d(i, i) = 0$ (zero distance to itself)

$d(i, j) = d(j, i)$ (symmetric measure)

$d(i, j) \leq d(i, k) + d(k, j)$ (shortest distance between two points)



DISTANCE BETWEEN NOMINAL VALUES

Example: How similar are these two people?

i = (Refund = Yes, Married, Income = 120K)

j = (Refund = No, Divorced, Income = 90K)

Taxpayer	Refund	Marital Status	Income in Thousands
i	Yes	Married	120
j	No	Divorced	90

METHOD 1: SIMPLE MATCHING

Taxpayer	Refund	Marital Status	Income in Thousands
<i>i</i>	Yes	Married	120
<i>j</i>	No	Divorced	90

m : Number of matches; p : Total number of nominal variables

$$d(i, j) = \frac{p}{p} m$$

METHOD 2: CONVERT NOMINAL TO BINARY VARIABLES

Taxpayer	Refund	Marital Status	Income in Thousands
<i>i</i>	Yes	Married	120
<i>j</i>	No	Divorced	90

Convert a nominal attribute to multiple binary attributes, and treat binary attributes as numeric (0 or 1).

Taxpayer	Refund	Married?	Divorced?	Single?	Income
1	1	1	0	0	120
2	0	0	1	0	90

BINARY VARIABLES: SYMMETRIC OR ASYMMETRIC

All patients run through many tests.

How different are their test results?

Patient	Test 1	Test 2	Test 3	Test 4	Test 5	Test 6
Jack	1	0	1	0	0	0
Mary	1	0	1	0	1	0

BINARY VARIABLES: SYMMETRIC OR ASYMMETRIC

A contingency table for
binary data

Gives the number of attributes of each
pair of values

		<i>Mary</i>		
		1	0	<i>sum</i>
<i>Jack</i>	1	<i>a</i>	<i>b</i>	<i>a+b</i>
	0	<i>c</i>	<i>d</i>	<i>c+d</i>
	<i>sum</i>	<i>a+c</i>	<i>b+d</i>	<i>p</i>

Patient	Test 1	Test 2	Test 3	Test 4	Test 5	Test 6
Jack	1	0	1	0	0	0
Mary	1	0	1	0	1	0

SYMMETRIC BINARY ATTRIBUTES

Distance measure for symmetric binary attributes:

		Object j		
		1	0	sum
Object i	1	a	b	$a+b$
	0	c	d	$c+d$
sum		$a+c$	$b+d$	p

$$d(i, j) = \frac{b+c}{a+b+c+d}$$

ASYMMETRIC BINARY ATTRIBUTES

If most test results are negative, d will be much greater than a , b , and c . Sharing many negative test results is not that informative to doctors.

		Object j		
		1	0	sum
Object i	1	a	b	$a+b$
	0	c	d	$c+d$
	sum	$a+c$	$b+d$	p

Distance measure for asymmetric binary attributes:

$$d(i, j) = \frac{b + c}{a + b + c}$$

DISTANCE BETWEEN ORDINAL VALUES

Method 1: Treat as nominal.

Method 2: Treat as numeric.

ATTRIBUTES OF MIXED TYPES

A database may contain different types of attributes: Symmetric binary, asymmetric binary, nominal, ordinal, numerical

How to compute the distance between examples with heterogeneous attributes?

Calculate distance for each type of attribute and aggregate.

SIMILARITY MEASURE

If defining a distance measured in $[0,1]$ range, similarity can be defined as $1 - d$.

Other similarity measures:

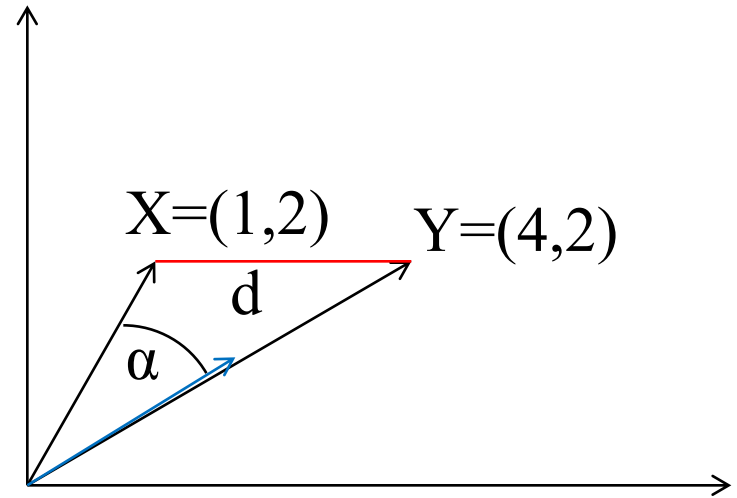
Cosine similarity measure

VECTOR SPACE REPRESENTATION AND COSINE SIMILARITY

Distance and similarity measures

Euclidean distance

$$\begin{aligned}d &= \sqrt{(x_1 - y_1)^2 + (x_2 - y_2)^2} \\&= \sqrt{(1 - 4)^2 + (2 - 2)^2} = 3\end{aligned}$$



Cosine similarity

$$\begin{aligned}\cos(\alpha) &= \frac{x \cdot y}{\|x\| \|y\|} = \frac{x_1 y_1 + x_2 y_2}{\sqrt{x_1^2 + x_2^2} \sqrt{y_1^2 + y_2^2}} \\&= \frac{1 \cdot 4 + 2 \cdot 2}{\sqrt{1^2 + 2^2} \sqrt{4^2 + 2^2}} = \frac{8}{\sqrt{5} \sqrt{20}} = 0.8\end{aligned}$$

COSINE SIMILARITY

In the range of $[0,1]$:

“0” means two vectors are perpendicular to each other.

“1” means same vector direction and length.

Commonly used in information retrieval and text mining to compare document similarity

High-dimensional space

Each word in the vocabulary is a dimension.