## Approximate Searches

## Similarity searches in Postgresql using metric spaces

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## We are going to talk about

- Searches with exact match
- Metric Spaces
- Approximate Searches
- Edit Distance
- Example
- Pivoting - Indexing
- Future issues


## Requirements to Approx searches

- Postgresql
- C language
- SQL language
- PI-pgsql language


## Exact Match

## Select * from customers where name $=$ 'Enrico'

Result :

Enrico
| address
2, red street

Ottawa
There is match!!

## Exact Search : when can I use it?

It make sense to use an exact search when we presume that our result is inside our database -> .....where name='Enrico'

## EXACT SEARCH

## Other kind of search ?



Images Search


Economic Search

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Sounds Search


DNA Search

## Other kind of search ?



## Similarity Searches

## Similarity Searches

## Some Examples:

- Text retrievial
- Image searches
- Sound searches
- Other .....


## Similarity Searches



In the exact searches we think our dataset as a set of rows and we make searches inside it

When we make a similatity search we have to think our dataset as a set of objects. The query is an object too that can belong or not belog to the dataset

## Similarity Searches

The "similiarity search" is the search of objects that belong to the dataset and that are closer to the query object

## Edit distance

## Edit distance or Levenshtein distance:

The Edit distance between two strings is given by the minimum number of operations needed to transform one string into the other, where an operation is an insertion, deletion, or substitution of a single character.

## Edit distance

For example the distance between "kitten" and "sitting" is 3

- kitten $\rightarrow$ sitten (substitution of 's' for ' $k$ ')
- sitten $\rightarrow$ sittin (substitution of 'i' for 'e')
- sittin $\rightarrow$ sitting (insert 'g' at the end)


## Edit distance

Now we can compare:

- Strings

Every kind of object starting from its features -> Every object has features

- We find an object that has the smallest distance from our query.


## Edit distance



## Some Problems

Objects can have many dimensions : tipical 50-60 dimensions 1 dimension for each feature (for example 3 for RGB images, and so on...)

We can spend much time to calculate the edit distance.

## Metric Spaces

A metric space is a set $S$ with a global distance function (the metric d) so that for every two points $x, y$ in $S$, returns the distance between them as a nonnegative real number $d(x, y)$.

## Metric Spaces

## The distance function $d(x, y)$ must be :

- non-negative: $d(x, y)>=0$
- Strictly Positive : $d(x, y)=0$ iff $x=y$
- Symmetric: $d(x, y)=d(y, x)$
- Have to satisfy the triangle inequality : $d(x, z)<=d(x, y)+d(y, z)$


## Metric Spaces



## Database objects are seen as points in a metric space

## Query point can belong to the dataset



Multimedia Dataset

## Query point can not belong to the dataset



Multimedia Dactaset

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Query

## Dataset vs points



Objects and queries are seen as points in a multidimensional metric space

## Metric Spaces: Searches

- Nearest neighbor : search of the object more close to the query point
- K - Nearest neighbor : search of the K objects more close to the query point
- Range query : search of objects that are inside the circle with a given radius $r$ and center in query point q


## Nearest Neighbor

(D9) (Db) (DC) (DC)

The point p1 is the nearest neighbor for the query point q

## - Nearest Neighboor: an example

SELECT parola, editdistance (parola, 'contorno') from parole order by 2 limit 5;
quattro terre
comparsa donna
scomparso nonno
volto
quattro


## Range Query



Points p1,p8,pd are inside the circle

## Range Query : an example

SELECT parola, editdistance (parola, 'contorno') from parole where editdistance(parola,'contorno') < = 5; quattro terre
comparsa donna
scomparso
nonno
volto
quattro

contorno
parola | editdistance


## Similarity searches

- Instead of words we can use any kind of strings
- We can compare n-ple of values(a1,a2,.., an) that represents features of objects.


## The dark side

- High dimensional spaces tipical 50,60 dimensions for each object
- A lot of memory
- A lot of time to calculate edit distance


## The triangle inequality



$$
d(q, p 1)<=d(q, P v)+d(P v, P 1)
$$

Better solutions than brute force

## The triangle inequality

Let $(X, d)$ be a metric space, where $X$ is the universe of valid objects and $d$ is the metric of the space, and let $U$ a subset of objects of $\mathrm{X}|\mathrm{U}|=\mathrm{n} \mathrm{U}$ is our database.
$(\mathrm{q}, \mathrm{r})=\{\mathrm{u}$ that belongs to U so that $\mathrm{d}(\mathrm{u}, \mathrm{q})<=r\}$ Range Query
Given a quey (q,r) and a set
triangle inequality it follows
also that $\mathrm{d}(\mathrm{pi}, \mathrm{q})<=\mathrm{d}(\mathrm{pi}, \mathrm{x})+\mathrm{d}(\mathrm{f}$
From boit inequailites, it fofere that a lower bound on d(a
Is $d(q, x)>=|d(p i, x)-d(p i, q)|-\frac{1}{l}$ objects $u$ of interest are those that satisfy $d(q, u)<=r$, so all the objects that satisfy the exclusion condition can pe excluded, without actually evaiuating d(q,u)

## Do you need some coffee ?

## Are you still alive?

Yes?

Ok, Now we will enjoy :)

## Building an Index - Pivoting



Some examples:
$d(P v, P 4)=1$
$d(P v, P 3)=2$
...... and so on

We can choose a point as pivot point and when we insert a new item we can precalculate the distance beetween our pivot and the new point

## Building an Index - Pivoting



## Building an Index - Range Query



## Q is our query point

## Building an Index - Range Query



Q is our query point

## Building an Index - Range Query


$d(q, p v)=3$


## Building an Index - Range Query



## Building an Index - 2 Pivots

If we have 2 or more pivots we consider as candidate points all the point that are in the
intersections of the distance calculated among pivots


## Building an Index - 2 Pivots



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## Building an Index - 2 Pivots



## An example:

Range query query point q radius $r=2$

## Building an Index - 2 Pivots



## Other features

- We can implement an algoritm also for KNN queries using our index structure
- We can use an approximated editdistance function (using AC or PAC alghoritms) to minimize computational time.


## Status of work



## INDEX FOR METRIC SPACES

TO IMPROVE RANGE QUERIES AND K-NN QUERIES

## The Future, but



## ...................................I'm alone



## The End: we talked about

- Metric Spaces
- Approximate Searches
- Edit Distance
- Example
- Pivoting - Indexing
- Future issues


## Bibliography

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