## SIMILARITY JOIN FOR BIG GEOGRAPHIC DATA

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## Motivation

## Geographic Data

Partitioning

## The Problem

- Cloud-based systems are crucial to processing and analyzing large amounts of data
- Similarity Joins (SJ) are a key data processing and analysis tool
- Very little work on Similarity Joins has been done for big geographic data
Our Contribution
- We propose MRSimJoin -a MapReduce-based algorithm to efficiently solve the SJ problem
- The algorithm is general enough to be used with data that lies in any metric space
- Our focus is on the study of this operation with big geographic data
- Thorough evaluation of performance and scalability with real world and synthetic geographic data sets
- Geographic data uses latitude $(\varphi)$ and longitude ( $\lambda$ ) coordinates to represent a location on a sphere
- There are several methods of calculating distance between two points
- Euclidean Distance
- Great Circle Distance
- Tunnel Distance
- This presentation considers the case of Euclidean Distance on a plane where a spherical earth is projected using equirectangular projection
- Euclidean Distance is fast to compute and accurate at small distances
- Given two points

$$
\begin{aligned}
& -r_{1}=\left(\varphi_{1}, \lambda_{1}\right) \\
& -r_{2}=\left(\varphi_{2}, \lambda_{2}\right)
\end{aligned}
$$

- The Euclidean Distance between them is as follows:
$\operatorname{geoDist}\left(r_{1}, r_{2}\right)=R \sqrt{\left(\Delta_{\varphi}\right)^{2}+\left(\cos \left(\varphi_{m}\right) \Delta_{\lambda}\right)^{2}}$
- MRSimJoin iteratively partitions the data into smaller partitions until each partition is small enough to be efficiently processed by a single-node SJ routine
- This process is done in multiple rounds, each corresponding to a MapReduce job
- Each round outputs result links and intermediate data requiring further partitioning



## Partitioning in an MRSimJoin Round

- Data partitioning is performed using a set of K pivots (conceptually similar to QuickJoin), which are a random subset of the records to be partitioned
- The process generates two types of partitions: base partitions and windowpair partitions

1) A base partition contains all the records that are closer to a given pivot than to any other pivot
2) A window-pair partition contains the records in the boundary between two base partitions


Partitioning a base partition


Partitioning a window-pair partition

Tests run over 2 million (SF1) records

## Increasing Scale Factor



Increasing Epsilon
Output Size

- MRSimJoin
-MRThetaJoin


