**Indexing Larger Spatial Data using Hadoop, a closer look at spatialhadoop and its complementary QuadTree implementation**

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**Building Index**

In general, database storage consists of persisted index structure and data storage, but neither part should infinitely scale up. In big data environment, we leverage the advantage of distribution and parallelism to achieve scalability.

Indexing process partitions large data set into even-sized data chunks (the number of chunks is determined by the volume of spatial dataset). Each data chunk is a small but fully-fledged database storage.

Data

Index

Data chunk, size limit 64MB

By default, the block size of Hadoop distributed file system (HDFS for short hereafter) is 64MB. Therefore our default chunks should be close to but no larger than that size, in order to guarantee i/o efficiency.

**Partitioning**

In our case, input data are formatted texts where each line represents a polygon object. Mapper input takes <line number, line text string> as key-value pairs. Mapreduce framework will split the files into smaller splits before submitting them to mappers. Multiple mappers will work in parallel to process the input splits. The level of parallelism is determined by the volume of data set or can be customized by the user.

Mappers convert the text string into a spatial object and emits <mapperInstanceID, object> pairs for reducers. The emitted key-value pairs which share the same key will be sent to same reducer, and the reducer will collect an appropriate number of objects (a number adjusted to make data chunk close to 64MB) and build spatial index for them. Upon finishing index building, the index and raw data will be combined as a single data chunk and emit as <id, chunk> as a single binary file to the file system.

**Executing a Query**

Executing a range query is also a Mapreduce job, where input is a number of data chunks built from the first Mapreduce job. In each mapper, a customized input format reader was defined to load the data chunks into memory, where their original data structure are restored – the first section of file restored as index structures, and the second section of file are simply loaded into a large byte array.

Query was also executed in the mapper phase, index structure returns a set of object ids (id is a combination of file position and length) for any given range search. According to the object ids, the real content of the object will be retrieved from the byte array.

After that, a set of queried spatial objects are sent to reducer with key being the queryID, value being the spatial object. In this way, results belong to the same query will be aggregated in the same reducer.

**QuadTree and Modifications**