#### MapReduce in Nutch

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# MapReduce: Background

- Invented by Google
  - http://labs.google.com/papers/mapreduce.html
- Platform for reliable, scalable computing.
- Implemented in Java as a part of Nutch
- Programmer specifies two primary methods:
  - map(k, v)  $\rightarrow \langle k', v' \rangle^*$
  - reduce(k',  $\langle v' \rangle^*$ )  $\rightarrow \langle k', v' \rangle^*$
  - also partition(), compare(), & others
- All v' with same k' are reduced together, in order.

#### MapReduce Diagram



### MapReduce: Pros & Cons

- Not always a natural fit,
  - but, with moderate force, many things will fit.
- Not always optimal,
  - but not far off, and often cheaper in the end.
- Developing large-scale systems is expensive
- Shared platform:
  - minimizes development & debug time
  - maximizes optimizations, tools, etc.

## Nutch Algorithms

- *inject* urls into a crawl db, to bootstrap it.
- loop:
  - generate a set of urls to fetch from crawl db;
  - *fetch* a set of urls into a *segment*;
  - *parse* fetched content of a segment;
  - *update* crawl db with data parsed from a segment.
- *invert links* parsed from segments
- *index* segment text & inlink anchor text

#### Data Structure: Crawl DB

- CrawlDb is a directory of files containing: <URL, CrawlDatum>
- CrawlDatum:

<status, date, interval, failures, linkCount, ...>

• Status:

{db\_unfetched, db\_fetched, db\_gone, linked, fetch success, fetch fail, fetch gone}

# Algorithm: Inject

• MapReduce1: Convert input to DB format

In: flat text file of urls Map(line)  $\rightarrow$  <url, CrawlDatum>; status=db\_unfetched Reduce() is identity;

Output: directory of temporary files

- MapReduce2: Merge into existing DB Input: output of Step1 and existing DB files Map() is identity.
   Reduce: merge CrawlDatum's into single entry
  - Out: new version of DB

### Algorithm: Generate

• MapReduce1: select urls due for fetch

In: Crawl DB files

 $Map() \rightarrow if date \ge now, invert to < CrawlDatum, url >$ 

Partition by value hash (!) to randomize

Reduce:

compare() order by decreasing CrawlDatum.linkCount output only top-N most-linked entries

• MapReduce2: prepare for fetch

Map() is invert; Partition() by host, Reduce() is identity. Out: Set of <url,CrawlDatum> files to fetch in parallel

### Algorithm: Fetch

• MapReduce: fetch a set of urls In: <url,CrawlDatum>, partition by host, sort by hash  $Map(url,CrawlDatum) \rightarrow \langle url,FetcherOutput \rangle$ multi-threaded, async map implementation calls existing Nutch protocol plugins FetcherOutput: <CrawlDatum, Content> Reduce is identity Out: two files: <url,CrawlDatum>, <url,Content>

### Algorithm: Parse

- MapReduce: parse content
  In: <url, Content> files from Fetch
  - Map(url, Content) → <url, Parse> calls existing Nutch parser plugins
  - Reduce is identity.
  - Parse: <ParseText, ParseData>
  - Out: split in three: <url,ParseText>, <url,ParseData> and <url,CrawlDatum> for outlinks.

# Algorithm: Update Crawl DB

 MapReduce: integrate fetch & parse out into db In: <url,CrawlDatum> existing db plus fetch & parse out Map() is identity

Reduce() merges all entries into a single new entry overwrite previous db status w/ new from fetch sum count of links from parse w/ previous from db Out: new crawl db

#### Algorithm: Invert Links

• MapReduce: compute inlinks for all urls In: <url,ParseData>, containing page outlinks Map(srcUrl, ParseData>  $\rightarrow$  <destUrl, Inlinks> collect a single-element Inlinks for each outlink limit number of outlinks per page Inlinks: <srcUrl, anchorText>\* Reduce() appends inlinks Out: <url, Inlinks>, a complete link inversion

### Algorithm: Index

• MapReduce: create Lucene indexes In: multiple files, values wrapped in <Class, Object> <url, ParseData> from parse, for title, metadata, etc. <url, ParseText> from parse, for text <url, Inlinks> from invert, for anchors <url, CrawlDatum> from fetch, for fetch date Map() is identity Reduce() create a Lucene Document

call existing Nutch indexing plugins

Out: build Lucene index; copy to fs at end

### MapReduce Extensions

- Split output to multiple files
  - saves subsequent i/o, since inputs are smaller
- Mix input value types
  - saves MapReduce passes to convert values
- Async Map
  - permits multi-threaded Fetcher
- Partition by Value
  - facilitates selecting subsets w/ maximum key values

# Summary

- Nutch's major algorithms converted in 2 weeks.
- Before:
  - many were undistributed scalabilty bottlenecks
  - distributable algorithms were complex to manage
  - collections larger than 100M pages impractical
- After:
  - all are scalable, distributed, easy to operate
  - code is substantially smaller & simpler
  - should permit multi-billion page collections