# TinyGoogle Discussion

### Main Task

- Develop and infrastructure to support two types of queries
  - IndexDoc(D)
    - Produce an Index of a D
      - Count the occurrence of each work in the document
    - Merge the index of d with the Main Index
  - Retrieve[<Di, (i=1, N)>, <Ki, (i=N>)]
    - Parse keyword set
    - For each key, k, identify all documents where k appears
    - Rank order all documents
    - Return query result

### Indexing a document

I can not do
everything, but
still I can do
something; and
because I cannot
do everything, I
will not refuse to
do something I
can do



Word	Count
And	1
Because	1
But	1
Can	4
Do	4
Everything	2
I	5
Not	3
Refuse	1
Something	2
Still	1
То	1
Will	1

### Doc Indexing - Program Structure

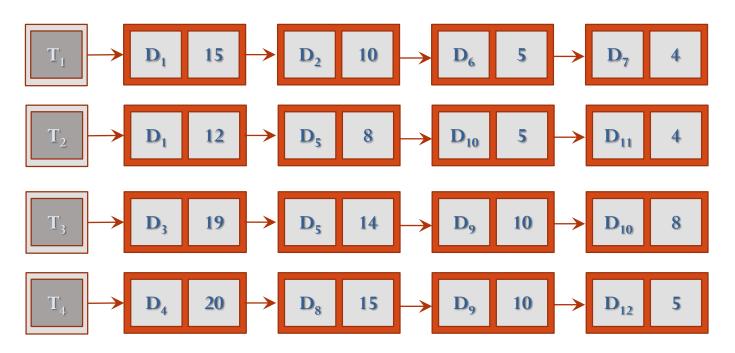
- A three-phased program can be used to distribute the work over several machines the final document index and merge it with main index
- Phase I Document Processing
  - Each machine will process a fraction of the document set
- Phase II Count Aggregation
  - Partial key word counts from individual machines are combined into the document index
- Phase III Merge document index into main index
  - Divide document index into chunks and assign machines to merge checks into main index

## Query Retrieval – Program Structure

- Assign keywords to different workers
  - Main index must be stored distributed for parallel access
- Produce documents
- Rank order document,

#### Inverted Index Structure

 Given a query, retrieval involves fetching posting with query terms and traversing the postings to compute the result set

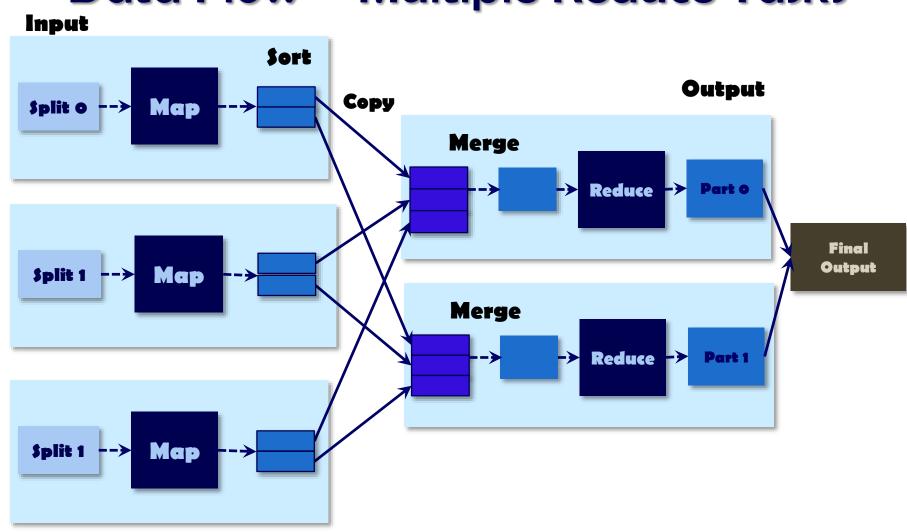


# MapReduce

### MapReduce - TinyGoogle Data Flow

- A MapReduce job is a unit of work to be performed
  - Job consists of the "MapReduce Program", the Input data and the Configuration Information and Runtime System
- The MapReduce job is divided it into two types of tasks map tasks and reduce tasks
- The Input data is divided into fixed-size pieces called \*plit\*
  - One map task is created for each split
- Configuration information and runtime system
  - Preprocessing of input data Location of input indicates where the input lies, and the output is stored

## Data Flow – Multiple Reduce Tasks



#### **Document Index**

- Define WordCount as Multiset;
- For Each Document in DocumentSet, S

```
T {} = tokenize(d in S);
For Each Token in T
   WordCount[token]++;
  DocIndex(<t, cont>, for t in T)
```

### WordCount Program - II

#### Phase I

- Define WordCount as Multiset;
- For Each D in DocSubset {
  - T = tokenize(D);
  - For Each Token in T {
    - WordCount[token]++;
- }
- **-** }
- SendToPhaseII(wordCount);

#### Phase II

- Define TotalWordCount as Multiset;
- For each WordCount Received From firstPhase {
  - MultisetAdd
     (TotalWordCount,
     WordCount);
- **-** }

## Inverted Index Algorithm - Mapper

- Individual Documents are processed in parallel by the mappers
- Input to the mapper <key, value> pairs
  - Key = Document IDs,
  - Value = Actual document content
- Document is analyzed and tokenized
  - Term frequencies, f, are computed by iterating over all analyzed documents
- Mapper emits intermediate <key, value> pairs
  - Key = Term,
  - Value = Posting

### Inverted Index Algorithm - Mapper

```
class Mapper
  procedure Map(did id, doc d)
{

   forall term t in doc d do
      EMIT( term t, posting [id, H{t}] )
}
```

### **Shuffle and Sort Phase**

- In the shuffle and sort phase, MapReduce runtime performs a large distributed computation to group postings by term
  - With no additional effort by the programmer, execution framework brings together all postings that belong in the same posting list
- The outcome of the shuffle and sort phase simplifies significantly the task of the reducer

### Inverted Index Algorithm - Reducer

- Producer First Step
  - Producer creates an empty list and appends all postings associated with the same key to the list
    - Key = term
- Producer Second Step
  - Posting are sorted, based on document identifiers
- Producer Final Step
  - Entire posting list is emitted as value, with term as key
    - Typically, the posting list is first compressed before emission
  - The final <key, value> pairs are written to disk and comprise the inverted index

### Inverted Indexing – Final Outcome

- Each reducer writes its output in a separate file in the distributed file system,
  - Final index is split across R files, where R is the number of reducers.
    - Further consolidation of the file is not needed
  - Separately, an index to the postings lists must be built
    - Index allows retrieval engine to fetch the postings list, for a given term, by opening the appropriate file and seeking to the correct byte offset position in the file
    - Typically in the form of mappings from term to (file, byte offset) pairs