# PageRank Implementation in MapReduce

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## Hadoop version

- Your code will be tested under EMR AMI version 2.4.2
- You can develop and test your code using Hadoop 1.0.3, which is corresponding to the AWS EMR AMI version 2.4.2
- http://archive.apache.org/dist/hadoop/ common/hadoop-1.0.3/
- You may have some incompatible issue using other Hadoop versions.

## Development process

- 1 First install Hadoop in your local computer and test your code with a smaller dataset.
- (2) Then migrate to AWS EMR and test your code with the large dataset.

If you have problem with environment configuration, you can also use Amazon EMR as a build environment to compile programs for use in your cluster.

http://docs.aws.amazon.com/ElasticMapReduce/latest/DeveloperGuide/emr-build-binaries.html

## PageRank Driver function

```
int main(String[] args) {
 // job 1 extract wiki and remove red links
  PageRank.parseXml("wiki/data", "wiki/ranking/iter0-raw")
 // job 2 wiki adjacency graph generation
  PageRank.getAdjacencyGraph("wiki/ranking/iter0-raw", "wiki/ranking/iter0")
 // job 3 total number of pages
  PageRank.calTotalPages("wiki/ranking/iter0", "wiki/ranking/N")
 // job 4: iterative MapReduce
  for(int run =0; run<8; run++) {
  PageRank.calPageRank("wiki/ranking/iter"+String(run),
                       "wiki/ranking/iter"+String(run+1))
  // job 5: Rank page in the descending order of PageRank
  PageRank.orderRank()
```

#### **Extract links**

#### 1) Use XmlInputFormat

Mahout's XmlInputFormat will process XML files and extract out the XML between two configured start / end tags. So if your XML looks like the following: <main>

```
<person>
    <name>Bob</name>
     <dob>1970/01/01</dob>
     </person>
</main>
```

and you've configured the start / end tags to be <person> and </person>, then your mapper will be passed the following <LongWritable, Text> pair to its map method:

```
LongWritable: 10 Text: "<person>\n <name>Bob</name>\n <dob>1970/01/01</dob>\n </person>"
```

Extract title and links.

- 1 Title A can be simply extracted between <title> A </title>. No complex rule is needed to extract A. Just take what it is between <title> and <\title>.
- 2 Extract the wikilinks. We are no longer suffering to make a complex regular expression to extract the wikilinks since the red links will be thrown away in this job.

```
Extract title and wikilink in the page
<page>
  <title> AccessibleComputing</title> -- extract AceessibleComputing
for simplicity.
   <redirect title = "Computer accessibility"> --ignore the redirect title
   <text> [[Computer accessibility]]
</page>
<page>
  <title> Anarchism </title> --extract Anarchism
  <text> .... Is a [[political philosophy]] that advocates [[stateless
society | stateless societies]] of defined as [[self-goverance|self-
governed]]....
```

#### Extract wikilink

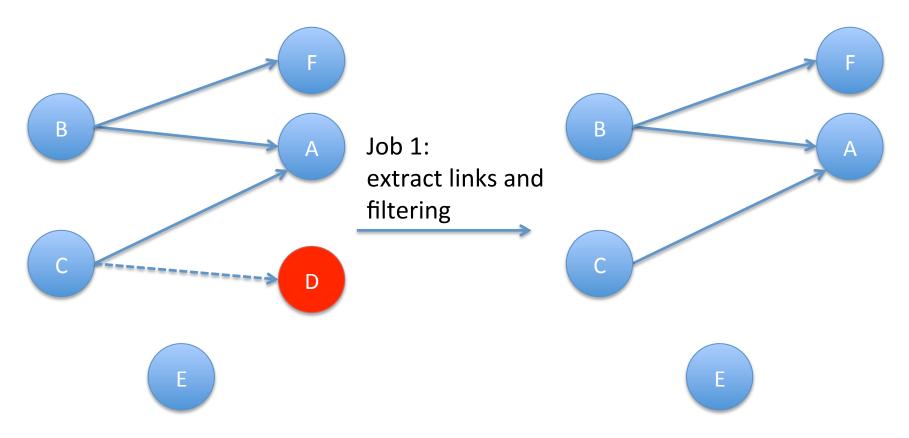
- 1 Assume case sensitive.
- 2 No other sophisticated processing is needed.
- 3 Replace empty space in title and wikilink with '\_'.

### Job1: remove red links

What is red links?
 http://en.wikipedia.org/wiki/Wikipedia:Red\_link

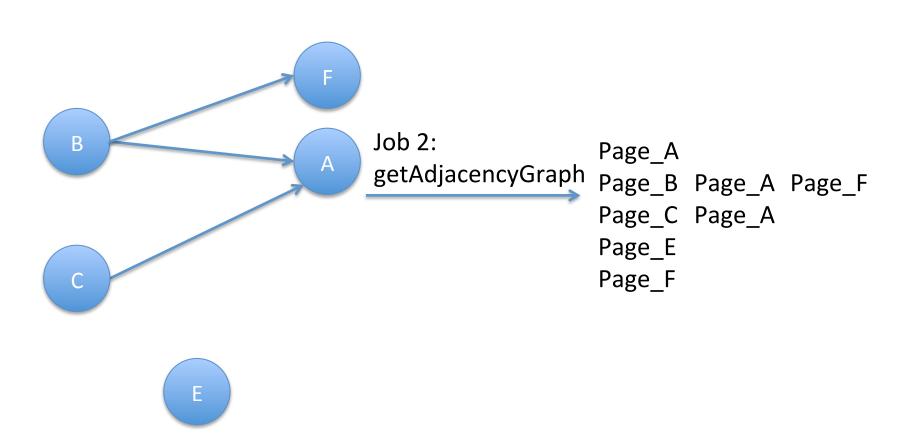
 Let's remove all the red links first before the PageRank calculation.

 How to remove red links in the project? Write a MapReduce job to throw away the red links.



D needs to be removed since there is no page D in the wiki dataset although page C mentions link D in its page. The nonexistence of page D simply means we are not able to find a <title>D<\title> in the dataset.

# Job2: adjacency graph



# Job2: adjacency graph

The output of Job2 should be a adjacency graph. The adjacency graph format is:

```
<src> <links in page>
```

- 1. The <src> is the title of the page.
- 2. The links in page> is the list of wikilinks found in page <src>.
- 3. The ks in page> should not contain red links
- 4. The links in page> should not contain duplicate links and it should not contain a link which points to the page itself.
- 5. All the data is separated by a tab.

### Job3: N calcuation

N is simply the number of <title>...<\title> pairs discovered in the dataset.

# Job4: PageRank Calculation

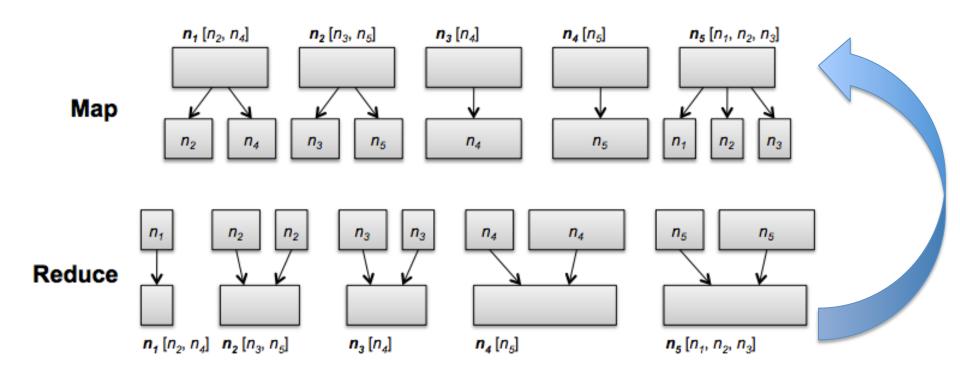
#### Algorithm 5.3 PageRank (simplified)

In the map phase we evenly divide up each node's PageRank mass and pass each piece along outgoing edges to neighbors. In the reduce phase PageRank contributions are summed up at each destination node. Each MapReduce job corresponds to one iteration of the algorithm. This algorithm does not handle dangling nodes and the random jump factor.

```
    class Mapper.

      method Map(nid n, node N)
 2:
          p \leftarrow N.PageRank/|N.AdjacencyList|
3:
          Emit(nid n, N)
                                                ▶ Pass along graph structure
 4:
          for all nodeid m \in N. Adjacency List do
 5:
             EMIT(nid m, p)
                                         ▶ Pass PageRank mass to neighbors
6:
 1: class Reducer
      method Reduce(nid m, [p_1, p_2, \ldots])
 2:
          M \leftarrow \emptyset
 3:
          for all p \in \text{counts } [p_1, p_2, \ldots] do
4:
             if IsNode(p) then
 5:
                 M \leftarrow p
                                                   ▶ Recover graph structure
6:
             else
7:
                                    8:
                 s \leftarrow s + p
          M.PageRank \leftarrow s
9:
          EMIT(nid m, node M)
10:
```

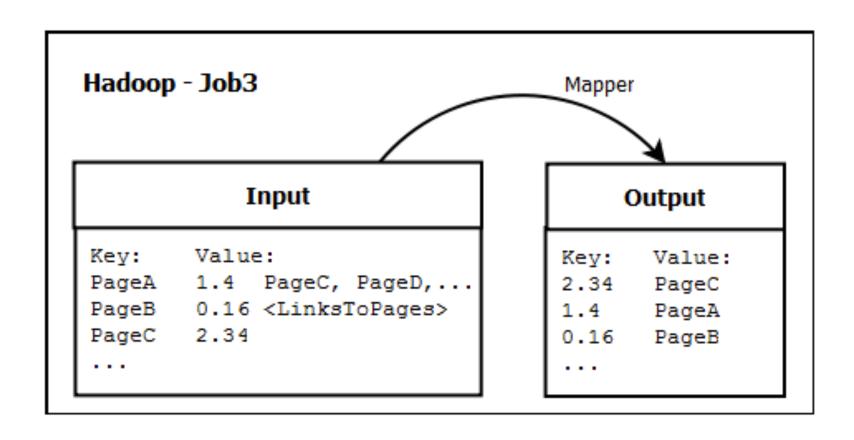
# Job4: PageRank Calculation



# Job 5: PageRank Ordering

- Filtering: only print out the page with PageRank >= 5/N
  (in the Map function)
- 2. Emit (PageRank, Page)
- 3. Only one Reducer
- Output the result in the descending order of PageRank.
   Here you will need to override the default sorter to sort in decreasing order. extends WritableComparator

# Job 5: PageRank Ordering

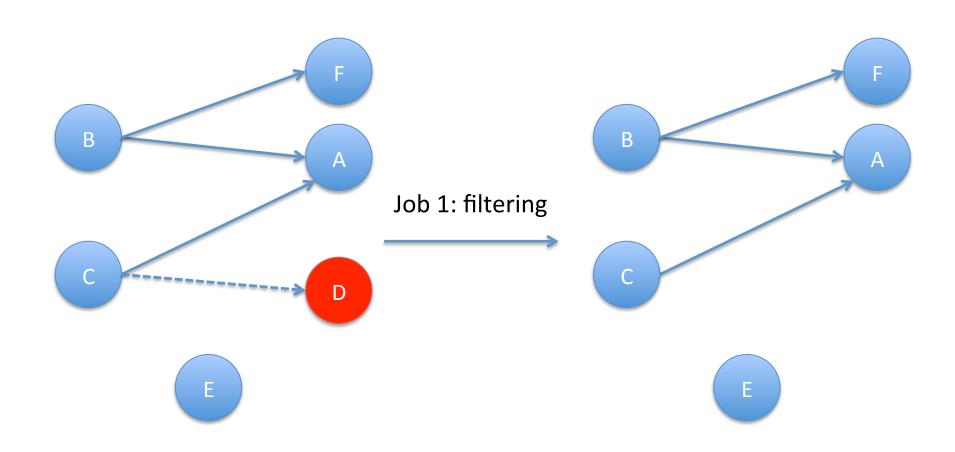


# Project inputs & outputs

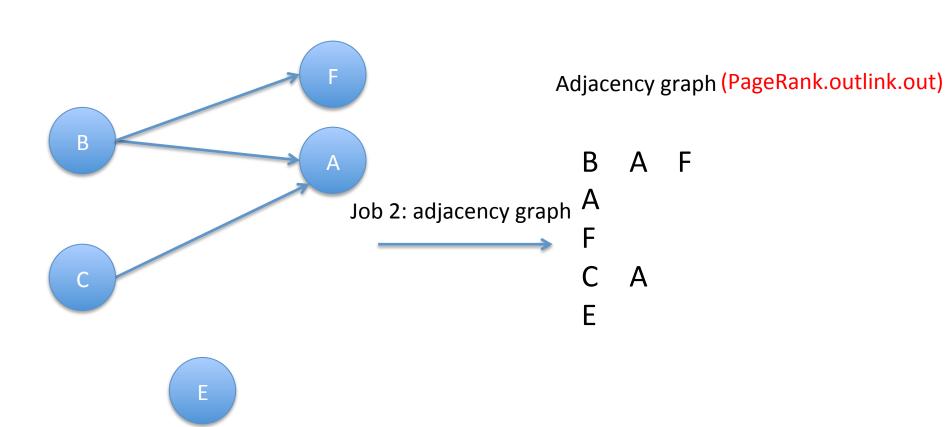
### Input

```
spring-2014-ds/data/enwiki-latest-pages-articles.xml
your-bucket-name
results/PageRank.outlink.out
results/PageRank.n.out
results/PageRank.iter1.out (output file for iteration 1)
results/PageRank.iter8.out (output file for iteration 8)
logs/ (the job log directory)
job/PageRank.jar (your job jar)
tmp/ (temporary files, you might or might not need it)
```

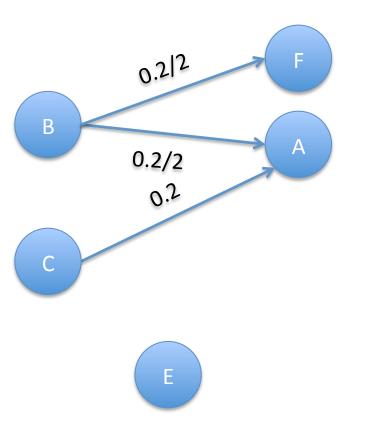
# Example Job 1: remove red links



# Example Job 2: adjacency graph



# Examples: Iteration 1

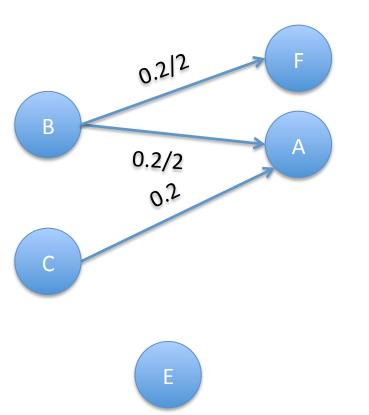


- Total number of pages N=5. It contains A,B,C,E,F
- E is a standalone page. It has no links.
   But E and its PageRank need to be printed out.

$$P(A) = (1-0.85)/5 + 0.85*(0.2/2 + 0.2/1) = 0.285$$
  
 $P(F) = (1-0.85)/5 + 0.85*(0.2/2) = 0.115$   
 $P(B) = (1-0.85)/5 = 0.03$   
 $P(C) = (1-0.85)/5 = 0.03$   
 $P(E) = (1-0.85)/5 = 0.03$ 

In our project, we don't use teleport to deal with the sink node for simplicity. At the initial point, the sum of the PageRank is 1. But the sum will gradually decrease with the iterations due to the PageRank leaking in the sink nodes.

## Examples: Iteration 2

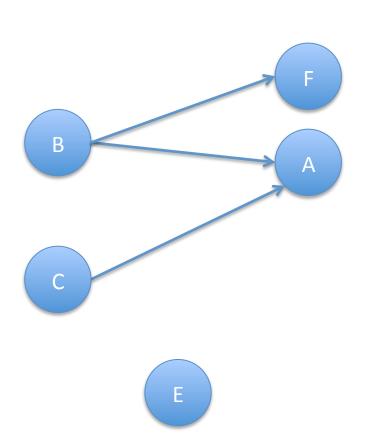


- Total number of pages N=5. It contains A,B,C,E,F
- E is a standalone page. It has no links.
   But E and its PageRank need to be printed out.

$$P(A) = (1-0.85)/5 + 0.85*(0.03/2 + 0.03/1) = 0.06826$$
  
 $P(F) = (1-0.85)/5 + 0.85*(0.03/2) = 0.04275$   
 $P(B) = (1-0.85)/5 = 0.03$   
 $P(C) = (1-0.85)/5 = 0.03$   
 $P(E) = (1-0.85)/5 = 0.03$ 

In our project, we don't use teleport to deal with the sink node for simplicity. At the initial point, the sum of the PageRank is 1. But the sum will gradually decrease with the iterations due to the PageRank leaking in the sink nodes.

# **Example: Results**



PageRank.n.out

N=5

PageRank.outlink.out

Page\_A

Page\_B Page\_A Page\_F

Page\_C Page\_A

Page\_E

Page\_F

PageRank.iter1.out

Page\_A 0.285

Page\_B 0.03

Page\_C 0.03

Page\_E 0.03

Page\_F 0.115

# Questions?