

Mining Massive Data Sets Midterm Report

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Abstract—This document is a model and instructions for L^AT_EX. This and the IEEEtran.cls file define the components of your paper [title, text, heads, etc.]. *CRITICAL: Do Not Use Symbols, Special Characters, Footnotes, or Math in Paper Title or Abstract.

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I. INTRODUCTION

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II. FIRST TASK: A-PRIORI ALGORITHM FOR FREQUENT CUSTOMERS

A. Overview of MapReduce

1) *What is MapReduce*: MapReduce is a yarn-based system commonly used for processing massive dataset:

- Performs concurrent processing by dividing the dataset into multiple chunks on the Hadoop commodity servers.
- Instead of sending the data to the machine with the logic to execute, we send the logic to the data to execute, specifically, the server.

2) *How MapReduce works*: MapReduce is executed in the following order:

- Split: Divides the dataset into multiple data batches.
- Map: Maps every element within each data batch to a $\langle \text{key}, \text{value} \rangle$ pair.
- Await Completion: Wait for all data batches to finish mapping the pairs.
- Combine: Generates $\langle \text{key}, \text{value} \rangle$ pairs in the form of a list (e.g., $[[A, 1], [A, 1]]$)
- Partition: Determines which reducer should handle each key. It uses a hash function (e.g., $\text{hash}(\text{key}) \% \text{num_reducers}$) to distribute keys evenly.
- Reduce: Processes every data assigned to it and return the output.

B. First subtask

In the first subtask, we are assigned to store the data on Hadoop Distributed File System (HDFS). After which we will implement a Hadoop MapReduce program in Java to discover groups of customers going shopping at the same date.

1) *The Mapper Class*: CustomerGroupByDateMapper

The Mapper class is responsible for reading input data and emitting key-value pairs. Key aspects of its implementation include:

- Input Processing: The input data is a CSV file with seven columns including Member_number (customer ID) and Date (transaction date).
- Filtering Headers: The Mapper ignores lines where Member_number is a header.
- Emitting Key-Value Pairs: The transaction date is used as the key, and the customer ID is used as the value. This allows all customer transactions on a given date to be grouped together during the shuffle and sort phase.

Example Output from Mapper:

(01/01/2014, 12345)
(01/01/2014, 67890)
(03/01/2014, 54321)

2) *The Reducer Class*: CustomerGroupByDateReducer

The Reducer class is responsible for aggregating the values emitted by the Mapper for each unique key. Key aspects of its implementation include:

- Collecting Unique Customer IDs: The reducer stores customer IDs in a HashSet to ensure uniqueness.
- Joining Values: The unique customer IDs are converted into a comma-separated string.
- Emitting Results: The final output consists of the transaction date as the key and the list of unique customer IDs as the value.

Example Output from Reducer:

(01/01/2014, 12345,67890)
(03/01/2014, 54321)

3) *Driver Program (Main Method)*: The driver program configures and executes the MapReduce job. It performs the following tasks:

- Setting up the Job: The job is named "Customer Date Groups" and configured to use GroupMapReduce as the main class.
- Setting Mapper and Reducer: The Mapper and Reducer classes are assigned appropriately
- Defining Input and Output: The input and output paths are provided as command-line arguments
- Job Execution: The job is submitted to Hadoop for execution, and the program exits based on its success or failure.

C. Second subtask

In the second subtask, we are assigned to implement the A-Priori algorithm to identify frequent customer pairs in the form of 02 Hadoop MapReduce programs, each corresponding to a pass.

1) *The First Pass*: Identifying Frequent Individual Customers

- Mapper Class: AprioriFirstPassMapper
 - Function: Reads transaction data and emits each customer ID as a key with a value of 1.
 - Filtering: Skips header lines and ensures valid data is processed.
 - Example Output from Mapper:

(12345, 1)
(67890, 1)
(12345, 1)
- Reducer Class: AprioriFirstPassReducer
 - Function: Aggregates the occurrences of each customer ID.
 - Filtering: Only customers meeting the support threshold (minimum occurrences) are retained.
 - Example Output from Reducer:

(12345, 2)
(67890, 1)

2) *The Second Pass*: Identifying Frequent Customer Pairs

- Mapper Class: AprioriSecondPassMapper
 - Setup: Loads frequent customers from the first pass output using Hadoop's distributed cache.
 - Processing: Reads transactions and filters out customers that did not meet the first pass threshold.
 - Pair Generation: Creates all possible pairs of frequent customers.
 - Example Output from Mapper:

(12345, 67890, 1)
(12345, 54321, 1)
- Reducer Class: AprioriSecondPassReducer
 - Function: Aggregates occurrences of customer pairs and filters based on the support threshold.
 - Example Output from Reducer:

(12345, 67890, 3)

3) *Driver Program (Main Method)*:

- First Pass Execution:
 - Runs the first MapReduce job to determine frequent individual customers.
 - Saves the output for use in the second pass.
- Second Pass Execution:
 - Loads the first pass results as cached data.
 - Runs the second MapReduce job to find frequent customer pairs.

III. SECOND TASK: PCY ALGORITHM FOR FREQUENT ITEMS

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REFERENCES

- [1] G. Eason, B. Noble, and I. N. Sneddon, "On certain integrals of Lipschitz-Hankel type involving products of Bessel functions," *Phil. Trans. Roy. Soc. London*, vol. A247, pp. 529–551, April 1955.
- [2] J. Clerk Maxwell, *A Treatise on Electricity and Magnetism*, 3rd ed., vol. 2. Oxford: Clarendon, 1892, pp.68–73.
- [3] I. S. Jacobs and C. P. Bean, "Fine particles, thin films and exchange anisotropy," in *Magnetism*, vol. III, G. T. Rado and H. Suhl, Eds. New York: Academic, 1963, pp. 271–350.
- [4] K. Elissa, "Title of paper if known," unpublished.
- [5] R. Nicole, "Title of paper with only first word capitalized," *J. Name Stand. Abbrev.*, in press.
- [6] Y. Yorozu, M. Hirano, K. Oka, and Y. Tagawa, "Electron spectroscopy studies on magneto-optical media and plastic substrate interface," *IEEE Transl. J. Magn. Japan*, vol. 2, pp. 740–741, August 1987 [Digests 9th Annual Conf. Magnetism Japan, p. 301, 1982].
- [7] M. Young, *The Technical Writer's Handbook*. Mill Valley, CA: University Science, 1989.

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