Database Project (Fall 2023)

Homework #2 (Due date: Oct 20)

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Compress 1) your codes and 2) the document as follows:

- 'DBP HW2 STUDENTID.zip'
 - ✓ Code
 - brute_force.py
 - mapper.py, reducer.py, combiner.py
 - ✓ Document: DBP HW2 STUDENTID.pdf

NOTE: You need to install matplotlib library.

- 1. [60pts] Consider you're searching for restaurants to book in Suwon. You must cho ose optimal restaurants in 'Suwon' while considering conflicting features. Use pare to-optimal set (skyline query) to filter the results from the database to keep only t hose objects that are not worse than others.
- (1) [30pts] Write the function named 'pareto_optimal' in the code file 'brute_force.py' to obtain the set of Pareto optimal restaurants in 'Suwon' from the 'restaurant.txt' dataset and plot the result.

Instructions:

- Refer to the definitions below and the baseline code which uses a brute-force algorithm to find the Pareto-optimal set. When plotting, please use the provided code. All data features should be considered. In general, higher quality and service are preferred, and lower prices are preferred.
- You can check the implementation of pareto_optimal function with the compare_result function in the main.py file.

[Definition 1: Dominant relationship]

For tuple $t_i \in \mathbb{R}^n = \{t_{i1}, t_{i2}, ..., t_{in}, ..., t_i\}$, $t_i < t_j$ (Dominant relationship) satisfies when t we conditions are true.

1) $\forall t_k \in t_i, \forall t_k \in t_i : t_k \leq t_{ik}$

For each k, t_k is less than or equal to t_{ik} .

2) $\exists t_k \in t_i, \exists t_{ik} \in t_i : t_k < t_{ik}$

There is at least one k that satisfies $t_k < t_{jk}$.

[Definition 2: Dominant tuple]

Given a set of tuples T, a set of dominant tuples is defined as:

$$Dom(T) = \{t \in T | \not\exists p \in T \not\exists t : t \leq p\}$$

 $Dom(T) = \{t \in T \mid \text{ a set of tuples } T \text{ except for } t \text{ has no element } p \text{ that satisfies } t \leq p\}$

Answer: Enter your code and result here.

pareto_optimal function code n open(input_file, 'n') as f: lines = f.readlines() lines = [line.strip().split(',') for line in lines] skyline = [] not skyline = [] not dominated: skyline.append((id_a, city_a, quality_a, service_a, price_a)) not_skyline.append((id_a, city_a, quality_a, service_a, price_a)) skyline, not_skyline Plot 14000 -10000 <u>-</u> 8000 6000 4000 300 250 0 20 150 80 100 100

(2) [30pts] Write the 'mapper.py' and 'reducer.py' code to produce the same results using MapReduce as the pareto optimal code you wrote in Problem 1 - (1).

Instructions:

- You can check the result output with the compare result function in the main.py file.

2. [40pts] Write additional code 'combiner.py'. Compare the time difference when using the combiner versus not using it (from 1-(2)) and provide an explanation for the observed variance.

Instruction:

- Run the command \$ bash base.sh and \$ bash base_combiner.sh to run MapReduce and compare the time consumed. Capture and report the images from each case, and briefly explain the result.
- If you are using M1 mac and wxw-matt:docker-hadoop repository, you must modify the version of Hadoop streaming in shell files.
- To ensure correct functionality of the shell files, the output file in HDFS must be in the /hw/output path. If you are going to use your own path, you must modify the shell files.

Report

[Without using combiner.py]

```
Shuffle Errors

BAD_ID=0

CONNECTION=0

IO_ERROR=0

WRONG_LENGTH=0

WRONG_MAP=0

WRONG_REDUCE=0

File Input Format Counters

Bytes Read=732807

File Output Format Counters

Bytes Written=21414

2023-10-19 20:27:07,146 INFO streaming.StreamJob: Output directory: /hw/output Elapsed time: 30387 ms
```

[With using combiner.py]

```
Shuffle Errors

BAD_ID=0

CONNECTION=0

IO_ERROR=0

WRONG_LENGTH=0

WRONG_MAP=0

WRONG_REDUCE=0

File Input Format Counters

Bytes Read=732807

File Output Format Counters

Bytes Written=21414

2023-10-19 20:28:33,520 INFO streaming.StreamJob: Output directory: /hw/output Elapsed time: 22012 ms
```

Using combiner.py reduced almost 8 seconds compare to one that is not using it. The purpose of the combiner is to perform local aggregation, and in this scenario, the logic for the local aggregation is the same as the global aggregation logic in the reducer. As a result, employing the combiner can help in minimizing the data transfer and improving the overall performance.