EECS-6421: EXAM

Electrical Engineering & Computer Science Lassonde School of Engineering

York University

| Family Name: Given Name: | |
|-----------------------------|--|
| Student#: | |
| EECS Account: | |

Instructor: Parke Godfrey
Exam Duration: take-home
Term: Fall 2017

Instructions

• rules

- The test is open-note, open-book.
- Use of a calculator is permitted.

answers

- Should you feel a question needs an assumption to be able to answer it, write the assumptions you need along with your answer.
- This is an exam. Do not do work with others on it.
- If you need more room to write an answer, indicate where you are continuing the answer.
 Add extra "blank" pages for space, if really needed

• points

- The number of points a given question is worth is marked.
- There are three major parts, 30 points in total.

| Marking Box | | | |
|-------------|-----|--|--|
| 1. | / 5 | | |
| 2. | / 5 | | |
| 3. | / 5 | | |
| 4. | / 5 | | |
| 5. | / 5 | | |
| Total | /25 | | |

1. [5pt] **Datalog**¬. What's that again, in English this time!?

[EXERCISE]

Consider (again) the schema

 $\begin{aligned} & \textbf{Movie}(\underline{\text{title}}, \text{director}, \text{year}) \\ & \textbf{Cast}(\underline{\text{actor}}, \underline{\text{title}}, \underline{\text{role}}) \\ & \text{FK (title) refs } \textbf{Movie} \end{aligned}$

Consider the following rules which are used in the queries to follow.

```
castIn (A, M) \leftarrow cast (A, M, R).

actor (A) \leftarrow castIn (A, M).

castOut (A, M) \leftarrow castIn (A, M), castIn (A, N), M \neq N.

diCast (A, M) \leftarrow cast (A, M, R_1), cast (A, M, R_2), R_1 \neq R_2.
```

For each of the following Datalog¬ queries, restate the query in concise, understandable English.

(a) [1pt]

$$temp(A) \leftarrow diCast(A, M).$$

 $\leftarrow temp(A).$

(b) [1pt]

 $temp (A) \leftarrow actor (A), movie (M, D, Y), not castIn (A, M).$ $\leftarrow actor (A), not temp (A).$ (c) [1pt]

 $temp(A) \leftarrow castIn(A, M_1), castOut(A, M_2).$ $\leftarrow temp(A).$

(d) [1pt]

 $temp(A) \leftarrow castIn(A, M)$, **not** castOut(A, M). $\leftarrow temp(A)$.

(e) [1pt]

 $temp\ (A) \leftarrow castIn\ (A,\ M),\ \mathbf{not}\ diCast\ (A,\ M),\ \mathbf{not}\ castOut\ (A,\ M).$ $\leftarrow temp\ (A).$

2. [5pt] Negation Semantics. A somewhat stable model.

[Concept]

(a) [3pt] Is there a Datalog \neg database \mathbf{P} such that p (a positive atomic consequence) is a consequence of \mathbf{P} with respect to the stable model semantics, but p is not a consequence of \mathbf{P} with respect to the well-founded semantics, and \mathbf{P} has a unique stable model?

If this cannot happen, explain why not. Otherwise, provide an example.

(b) [2pt] Is there a Datalog \neg database \mathbf{P} such that p (a positive atomic consequence) is a consequence of \mathbf{P} with respect to the well founded semantics, but p is not a consequence of \mathbf{P} with respect to the stable model semantics? (Note that when \mathbf{P} has no stable models, everything is a consequence of \mathbf{P} with respect to the stable model semantics.)

If this cannot happen, explain why not. Otherwise, provide an example.

3. [5pt] Sequential Reads. Speed it up.

[Analysis]

Consider each of the join algorithms that we have studied, in turn. Explain briefly whether sequential reads and writes would be advantageous in each case.

Assume that sequential reads are generally not advantageous for filescans of base tables. Base tables become fragmented on disk over time due to inserts and deletes.

(a) [1pt] **BNLJ** (block nested loop join)

(b) [1pt] **INLJ** (index-nested loop join)

(c) [1pt] **HJ** (two-pass hash join)

(d) [1pt] \mathbf{SMJ} (two-pass sort merge join)

(e) [1pt] \mathbf{MJ} (merge join, with outer and inner sorted prior)

4. [5pt] Query Optimization. Simply the best plan available.

[EXERCISE]

Schema:

```
Employee(<u>eid</u>, name, did, jobCat, salary)

JobBenefits(<u>title</u>, <u>jobCat</u>, since)

FK (title) refs Benefit

Benefit(title, description, cost)
```

Statistics:

- **Employee**: 100,000 records on 2,000 pages
 - jobCat: 500 distinct values
 - did (department ID): 100 distinct values (department #13 is accounting)
- **JobBenefits**: 3,500 records on 70 pages
 - title: 200 distinct values
 - jobCat: 500 distinct values (same values as in **Employee**.jobCat)
- Benefit: 200 records on 20 pages
 - cost: ranges over \$500,...,\$10,500

Indexes:

- Employee:
 - Clustered tree index on eid. (Index pages two deep; third layer, data-entry pages.)
 - Unclustered tree index on did, jobCat. (Index pages two deep; third layer, data-entry pages.)
- JobBenefits:
 - Clustered tree index on jobCat, title. (Index page one deep; second layer, data-entry pages.)
- Benefit:
 - Hash index on title.

Query:

```
select name, eid, B.title
  from Employee E, JobBenefits J, Benefit B
  where E.jobCat = J.jobCat
   and J.title = B.title
  and E.did = 13
  and B.cost > 10000;
```

You have an allocation of twelve buffer frames.

(a) [1pt] Estimate the cardinality—the resulting number of records—of the query.

(b) [4pt] Devise a good query plan for the query. Show the query tree, fully annotated with the chosen algorithms and access paths.

Estimate the cost of your plan. For full credit, you should have a plan that costs less than 1,500 I/O's.

5. [5pt] MapReduce. Isn't this cartography?!

[Design]

Consider as input files that look as follows.

| ${f List}$ | | \mathbf{Price} | |
|------------|-----------|------------------|------|
| parke | ice cream | broccoli | 3.99 |
| parke | tofu | hummus | 4.50 |
| parke | tomatoes | ice cream | 6.25 |
| jeff | broccoli | $_{ m milk}$ | 4.99 |
| jeff | hummus | tofu | 3.59 |
| jeff | ice cream | tomatoes | 5.00 |
| eric | milk | : | |
| | : | • | |

Input **List** represents grocery lists for people. Input **Price** yields the price for each grocery item. Assume the format of each is *key-value*.

(a) [2pt] Design a MapReduce job that outputs grocery item (e.g., "ice cream") as key and the count of the number of people as value (e.g., 2) who have that item on their list.

Make clear your map and reduce procedure pairs:

- the input of each map;
- the output key-value of each map;
- the output key-value of each reduce; and
- simple pseudo-code / clear description of what each map and reduce does.

(b) [2pt] Design a *MapReduce* job that outputs the total cost—sum of prices—for each person's grocery list. E.g.,

Do not assume that the values are sorted by the shuffle; unlike in Project #2, your platform here cannot provide this. Note that a map can take its input from more than directory. Describe your MapReduce job as in Question 5a. You may assume that you have the output from the job in Question 5a in a directory Count. Be concerned that nearly everyone may have, say, ice cream on their list. (Hint: Think about ice cream, 1, ice cream, 2, ..., ice cream, N as keys.)

⁽c) [1pt] Can map and reduce filter keys, to eliminate keys that do not meet some condition? Briefly explain why or why not.

EXTRA SPACE