

Database Systems I

CMPT 354 Summer 2024 Zhengjie Miao

Course Setup

Ask professor if he can post the due dates of the assignment

- OH slots:
 - Tue 2pm-3pm, Zhengjie@TASC 1 9407
 - Mon 11am-12pm, Xinyi@ASB9812
 - Wed 1pm-2pm, Obumneme@ASB9812
 - Thu 3pm-4pm, Zahra@ASB9812
- Sign up for Piazza!
 - The place to ask course-related questions
 - https://piazza.com/sfu.ca/summer2024/cmpt354d100
 - Access code: qc76kdzsz9l

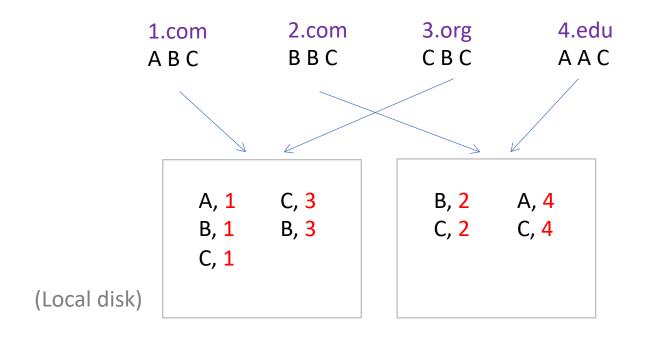
Recap: MapReduce

Online analytical processing

- Which design is better for OLAP?
 - MapReduce vs. Relational DBMS?
- Many problems can be processed in MapReduce pattern:
 - Given a lot of unsorted data Allows you to divide your data among multiple machines
 - Map: extract something of interest from each record
 - Shuffle: group the intermediate results in some way
 - Reduce: further process (e.g., aggregate, summarize, analyze, transform)
 each group and write final results

(Customize map and reduce for problem at hand)

MR Example - Map



key: document id

value: document contents

map(String key, String value):

for each word w in value:

Emit_Intermediate(w, document id);

MR Example - Shuffle & Reduce

A, 1 C, 3 B, 1 B, 3 C, 1 B, 2 A, 4 C, 2 C, 4



A, { 1, 4 } C, { 3, 1, 2, 4 }

B, { 1, 3, 2 }

Values exchanged by shuffle process

key: a word

values: a list of doc id

Why is MapReduced criticized?
Proved to be inefficient in database systems.
Ask professor why it is inefficient.



A, [1, 4] B, [1, 2, 3] C, [1, 2, 3, 4] The reduce function accepts all pairs for a given word, sorts the corresponding document IDs

Trends

Typically, you will need to maintain a transactional and analytical system, where you move data between them.

Ask him about why we would move our data frequently?

1 Hybrid Transactional and Analytical Processing

2. Cloud-Native Databases

Maintain database inside virtual machine
Cloud databases could allow infinite storage
disaggregating the storage and computation layer?
The scale of computation is independent from storage? Ask the
professor what he means

• 3. Lakehouse = Data Lake + Data Warehouse

Search up Data Lake and Data Warehouse

Lakehouse wants to result in efficient queries

- 4. Specialized Systems
 - E.g. Time-series Database, GPU Acceleration, Vector Database

Outline

An overview of data models

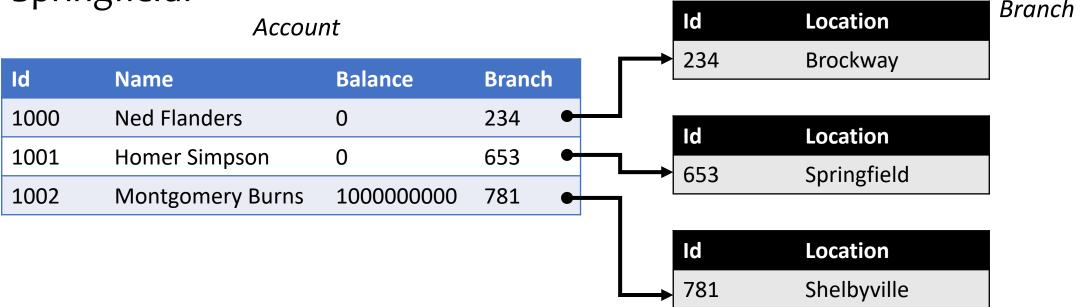
Basics of the Relational Model

- Relational Algebra
 - Core operators & Derived operators
 - Extension
 - Relational calculus

Recap: Data Models

Query: Who have accounts with o balance managed by a branch in

Springfield?



- Programmer controls "navigation": accounts → branches
 - How about branches → accounts? (e.g., find branches managing accounts with o balance)

Data Model

- Data Model
 - mathematical formalism (or conceptual way) for describing the data

If you do not have a data model, what happens?

Ask professor

- The description generally consists of three parts:
 - Structure of the data
 - Operations on the data
 - Constraints on the data

Structure of the data

- Schema (e.g., table names, attribute names)
 - Describe the conceptual structure of the data
- Different from data structure (e.g., list, array)
 - Data structure can be seen as a physical data model

Operations on the data

- Query language (e.g., RA, SQL)
 - Describe what operations that can be performed on data
- Two kinds of operations
 - operations that retrieve information ("query" the data)
 - operations that change the database ("update" the data)
- Different from programming languages (e.g., C, Java)
 - Support a set of limited operations
 - Allow for query optimizations

Your query can be optimized if the system finds another query which is equivalent and more efficient

Constraints on the data.

- Constraints (e.g., balance >= 0, account id is unique)
 - describe limitations on what the data can be.

- Different kinds of constraints
 - Domain constraints Ex: balance >= 0
 - Integrity constraints Ex: the user id is unique
- Why does it matter?
 - Ensure the correctness of data

Commonly Used Data Models

Relational Data Model

Key-Value Data Model

• Semi-structured Data Model (e.g., Json, XML)

The Relational Model in Brief

Id	Name	Balance	Branch		Id	Location
1000	Ned Flanders	0	234 •-	 	234	Brockway
1001	Homer Simpson	0	653 •	—	653	Springfield
1002	Montgomery Burns	1000000000	781 • -	—	781	Shelbyville

I thought SQL used RA, but that RA is not the query language

- Structure of the Data
 - Table structure
- Query language
 - RA, RC, SQL
- Constraints on the data
 - E.g., id is unique, balance >= 0, name is not NULL

The Key-Value Model in Brief

```
Key → Value

1000 → (Ned Flanders, 0, 234)

1001 → (Homer Simpson, 0, 653)

1002 → (Montgomery Burns, 1000000000, 781)
```

- Structure of the Data
 - (Key, Value) pairs
 - Key is an integer/string, value can be any object
- Query language
 - get(key), put(key, value)
- Constraints on the data
 - E.g., key is unique, value is not NULL

The Semistructured Model in Brief

You cannot apply the same operation to each account?

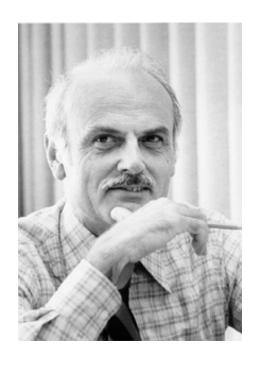
Ask professor to clarify this

- Structure of the Data
 - Tree structure (e.g., XML, json)
- Query language
 - XQuery, MongoDB QL
- Constraints
 - E.g., each <Account> has a <Id> element and a <name> element nested within

Resembles the hierarchical model "what goes around comes around"

```
<Accounts>
   <Account>
        <Id>1000</Id>
        <Name>Ned Flanders</Name>
        <Balance>0</Balance>
        <Branch>234</Branch>
   </Account>
   <Account>
        <Id>1001</Id>
        <Name>Homer Simpson</Name>
        <Balance>0</Balance>
        <Branch>653</Branch>
   </Account>
    <Account>
        <Id>1002</Id>
        <Name>Montgomery Burns</Name>
        <Balance>100000000</Balance>
        <Branch>781</Branch>
   </Account>
</Accounts>
```

Edgar F. Codd (1923-2003)



- Pilot in the Royal Air Force in WW2
- Inventor of the relational model and algebra while at IBM
- Turing Award, 1981

Relational data model

- A database is a collection of relations (or tables)
- Each relation has a set of attributes (or columns)
- Each attribute has a name and a domain (or type)
 - Set-valued attributes are not allowed Attributes must be atomic
- Each relation contains a set of tuples (or rows)
 - Each tuple has a value for each attribute of the relation
 - Duplicate tuples are not allowed
 - Two tuples are duplicates if they agree on all attributes Set semantics
- Simplicity is a virtue!

Example User

uid	name	age	рор
142	Bart	10	0.9
123	Milhouse	10	0.2
857	Lisa	8	0.7
456	Ralph	8	0.3

Ordering of rows doesn't matter (even though output is always in some order)

Group

gid	name
abc	Book Club
gov	Student Government
dps	Dead Putting Society

Member

uid	gid
142	dps
123	gov
857	abc
857	gov
456	abc
456	gov

Schema vs. instance

- Schema (metadata)
 - Specifies the logical structure of data
 - Is defined at setup time
 - Rarely changes

The schema rarely changes since we must update the change to the schema to all instances

Instance

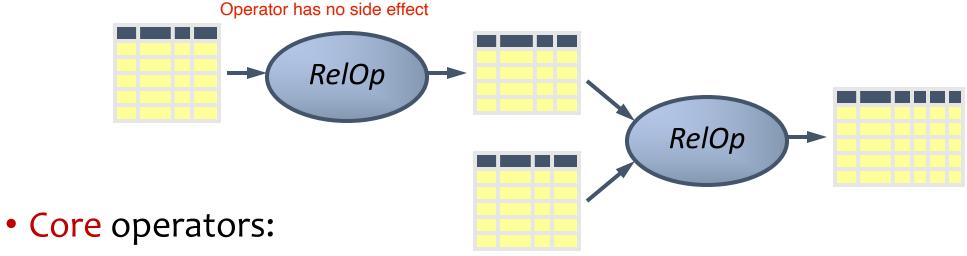
- Represents the data content
- Changes rapidly, but always conforms to the schema
- Compare to types vs. collections of objects of these types in a programming language

Example

- Schema Ask the professor what he said regarding
 - User (uid int, name string, age int, pop float)
 - Group (gid string, name string)
 - Member (uid int, gid string)
- Instance
 - User: $\{\langle 142, Bart, 10, 0.9 \rangle, \langle 857, Milhouse, 10, 0.2 \rangle, \dots \}$
 - Group: {\langle abc, Book Club \rangle, \langle gov, Student Government \rangle, \ldots \}
 - Member: {\(\((142\), dps\), \(\((123\), gov\), \(...\)}

Relational algebra

A language for querying relational data based on "operators"



- Selection, projection, cross product, union, difference, and renaming
- Additional, derived operators:
 - Join, natural join, intersection, etc.
- Compose operators to make complex queries

Selection

- Input: a table *R*
- Notation: $\sigma_p R$
 - *p* is called a selection condition (or predicate)
- Purpose: filter rows according to some criteria
- Output: same columns as R, but only rows or R that satisfy p

of

Selection example

• Users with popularity higher than 0.5 $\sigma_{pop>0.5}User$

uid	name	age	рор		uid	name	age	
142	Bart	10	0.9		142	Bart	10	
123	Milhouse	10	0.2	(T)				I
857	Lisa	8	0.7	$\sigma_{pop>0.5}$	857	Lisa	8	ĺ
456	Ralph	8	0.3					
								Ī

More on selection

- Selection condition can include any column of R, constants, comparison (=, \leq , etc.) and Boolean connectives (Λ : and, V: or, \neg : not)
 - Example: users with popularity at least 0.9 and age under 10 or above 12 $\sigma_{pop\geq 0.9 \ \land \ (age<10 \ \lor \ age>12)} User$
- You must be able to evaluate the condition over each single row of the input table!
 - Example: the most popular user

σ_{pop ≥ every pon ig User} User

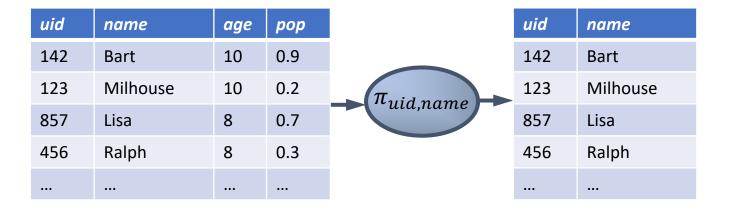
Projection

- Input: a table *R*
- Notation: $\pi_L R$
 - *L* is a list of columns in *R*
- Purpose: output chosen columns
- ullet Output: same rows, but only the columns in L

Projection example

• IDs and names of all users

 $\pi_{uid,name}$ User



More on projection

- Duplicate output rows are removed (by definition)
 - Example: user ages

$$\pi_{age}$$
 User

uid	name	age	рор
142	Bart	10	0.9
123	Milhouse	10	0.2
857	Lisa	8	0.7
456	Ralph	8	0.3
			•••

Composing Select and Project

• Names of users with popularity higher than 0.5 $\pi_{name}(\sigma_{pop>0.5}User)$

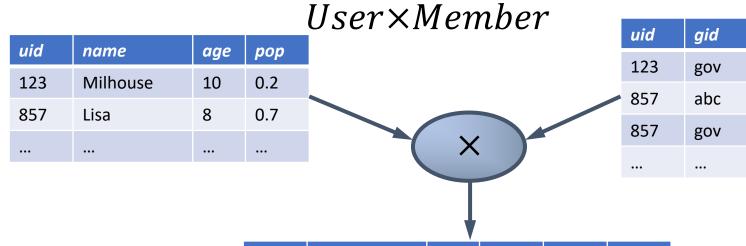
uid	name	age	рор		uid	name	age	рор
142	Bart	10	0.9		142	Bart	10	0.9
123	Milhouse	10	0.2		857	Lisa	8	0.7
857	Lisa	8	0.7					
456	Ralph	8	0.3				•••	
				$\sigma_{pop>0}$.	5			

- What about: $\sigma_{pop>0.5}(\pi_{name}(User))$
 - Invalid types. Input to $\sigma_{pop>0.5}$ does not contain pop.

Cross product

- Input: two tables *R* and *S*
- Natation: $R \times S$
- Purpose: pairs rows from two tables
- Output: for each row r in R and each s in S, output a row rs (concatenation of r and s)

Cross product example



- How many rows in result?
 - |R|*|S|
- Schema compatability?
 - Not needed.
- Duplicates?
 - None generated.

uid	name	age	рор	uid	gid
123	Milhouse	10	0.2	123	gov
123	Milhouse	10	0.2	857	abc
123	Milhouse	10	0.2	857	gov
857	Lisa	8	0.7	123	gov
857	Lisa	8	0.7	857	abc
857	Lisa	8	0.7	857	gov
•••		•••	•••	***	

A note on column ordering

• Ordering of columns is unimportant as far as contents are concerned

uid	name	age	рор	uid	gid		uid	gid	uid	name	age	рор
123	Milhouse	10	0.2	123	gov		123	gov	123	Milhouse	10	0.2
123	Milhouse	10	0.2	857	abc		857	abc	123	Milhouse	10	0.2
123	Milhouse	10	0.2	857	gov		857	gov	123	Milhouse	10	0.2
857	Lisa	8	0.7	123	gov	=	123	gov	857	Lisa	8	0.7
857	Lisa	8	0.7	857	abc		857	abc	857	Lisa	8	0.7
857	Lisa	8	0.7	857	gov		857	gov	857	Lisa	8	0.7
•••		•••	•••				•••		•••			•••

• So cross product is commutative, i.e., for any R and S, $R \times S = S \times R$ (up to the ordering of columns)

Derived operator: join

```
(A.k.a. "theta-join")
```

- Input: two tables *R* and *S*
- Notation: $R \bowtie_{p} S$
 - p is called a join condition (or predicate)
- Purpose: relate rows from two tables according to some criteria
- Output: for each row r in R and each row s in S, output a row rs if r and s satisfy p
- Shorthand for $\sigma_p(R \times S)$

Join example

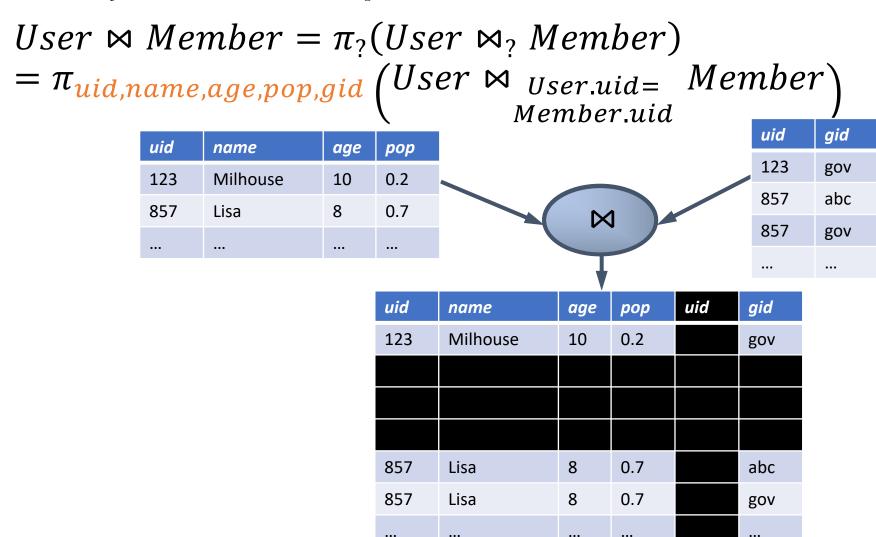
• Info about users, plus IDs of their groups $User\bowtie_{User.uid=Member.uid}Member$

		0.0	7 100							
	uid	name	age	рор	ı				uid	gid
	123	Milhouse	10	0.2					123	gov
	857	Lisa	8	0.7		M IIsor			857	abc
	837	Lisa	0	0.7	7	⊠ _{User.} Memb			857	gov
	•••	•••	•••	•••		I				
					name	age	рор	uid	gid	
refix a column refer	efix a column reference				Milhouse	10	0.2	123	gov	
vith table name and	"." to									
lisambiguate identic	ally na	med								
olumns from differe	nt tabl	es								
						8	0.7	857	abc	
				857 857	Lisa Lisa	8	0.7	857	gov	

Derived operator: natural join

- Input: two tables *R* and *S*
- Notation: $R \bowtie S$
- Purpose: relate rows from two tables, and
 - Enforce equality between identically named columns
 - Eliminate one copy of identically named columns
- Shorthand for $\pi_L(R \bowtie_p S)$, where
 - p equates each pair of columns common to R and S
 - L is the union of column names from R and S (with duplicate columns removed)

Natural join example



Union

- Input: two tables *R* and *S*
- Notation: $R \cup S$
 - R and S must have identical schema
- Output:
 - Has the same schema as R and S
 - Contains all rows in R and all rows in S (with duplicate rows removed)

uid	gid		uid	gid		
123	gov	U	123	gov	=	
857	abc		857	gov		
						8

Difference

- Input: two tables *R* and *S*
- Notation: R S
 - R and S must have identical schema
- Output:
 - Has the same schema as R and S
 - Contains all rows in R that are not in S

uid	gid		uid	gid			uid	gid
123	gov	—	123	gov	:	=	857	abc
857	abc		857	gov				

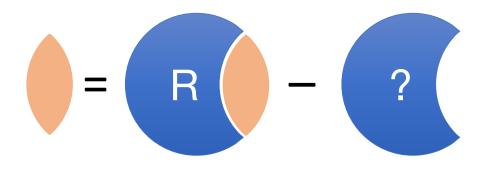
Derived operator: intersection

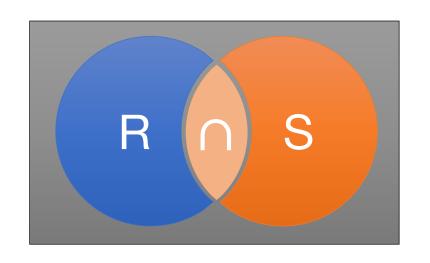
- Input: two tables *R* and *S*
- Notation: $R \cap S$
 - R and S must have identical schema
- Output:
 - Has the same schema as R and S
 - Contains all rows that are in both R and S
- Shorthand for? Could we also use the natural join?

uid	gid		uid	gid		uid	gid
123	gov	\cap	123	gov	=	123	gov
857	abc		857	gov			

Intersection

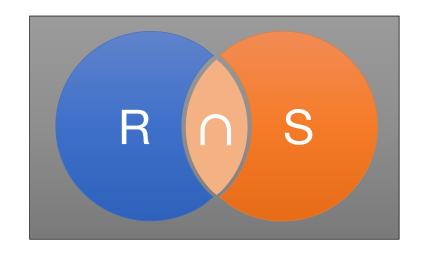
•
$$R \cap S = R - ?$$





Intersection

- $R \cap S = R (R S)$
- Also = S (S R)



$$= \left(\begin{array}{c} R \\ \end{array} \right) - \left(\begin{array}{c} R \\ \end{array} \right)$$

How can you write it without –

$$R \bowtie S$$

Renaming

• Input: a table *R*

857

abc

- Notation: $\rho_S R$, $\rho_{(A_1,A_2,...)}R$, or $\rho_{S(A_1,A_2,...)}R$
- Purpose: "rename" a table and/or its columns
- Output: a table with the same rows as R, but called differently

Renaming allows you to clarify which attributes belong to which instance

Member $\begin{array}{c|cccc} \textit{uid} & \textit{gid} \\ & \textit{123} & \textit{gov} \end{array} & \rho_{M1(uid1,gid1)} \textit{Member} & = \end{array}$

uid	gid
123	gov
857	gov

M1

Should the columns not be renamed?

Professor: they should

Renaming

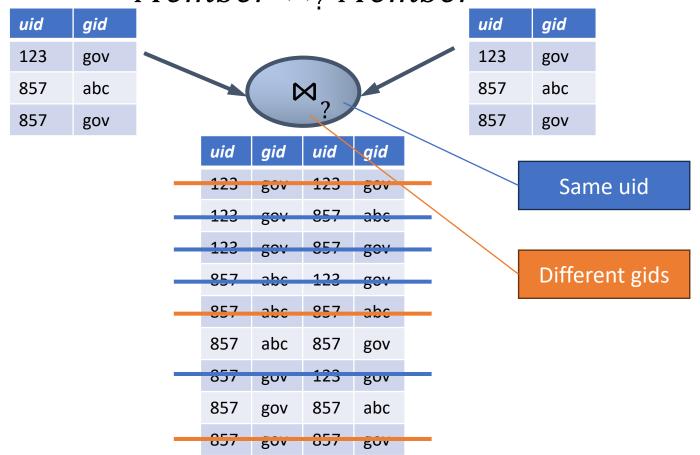
- Used to
 - Avoid confusion caused by identical column names
 - Create identical column names for natural joins
- As with all other relational operators, it doesn't modify the database
 - Think of the renamed table as a copy of the original

Men	nber		M1	
uid	gid	M l	uid	gid
123	gov	$\rho_{M1(uid1,gid1)}Member =$	123	gov
857	abc		857	gov

Note:

Renaming example

 IDs of users who belong to at least two groups *Member* ⋈_? *Member*



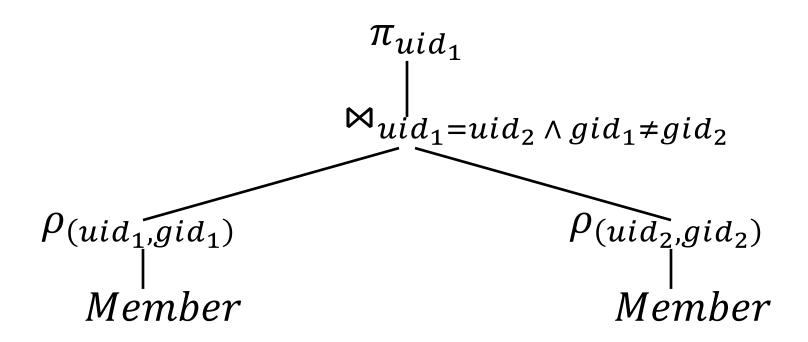
Renaming example

• IDs of users who belong to at least two groups *Member* ⋈_? *Member*

First condition will always be true Second condition will always be false $\pi_{uid}\left(\begin{matrix} Member \bowtie_{Member.uid} = Member.uid \land \\ Member.gid \neq Member.gid \end{matrix} \right)$

$$\pi_{uid_1} \begin{pmatrix} \rho_{(uid_1,gid_1)} Member \\ \bowtie_{uid_1 = uid_2 \land gid_1 \neq gid_2} \\ \rho_{(uid_2,gid_2)} Member \end{pmatrix}$$

Expression tree notation



Summary of core operators

- Selection: $\sigma_p R$
- Projection: $\pi_L R$
- Cross product: $R \times S$
- Union: $R \cup S$
- Difference: R S
- Renaming: $\rho_{S(A_1,A_2,...)}R$
 - Does not really add "processing" power

Summary of derived operators

- Join: $R \bowtie_p S$
- Natural join: $R \bowtie S$
- Intersection: $R \cap S$

- Many more
 - Semijoin, anti-semijoin, quotient, ...

An exercise

uid

857

User (uid int, name string, age int, pop float)
Group (gid string, name string)
Member (uid int, gid string)

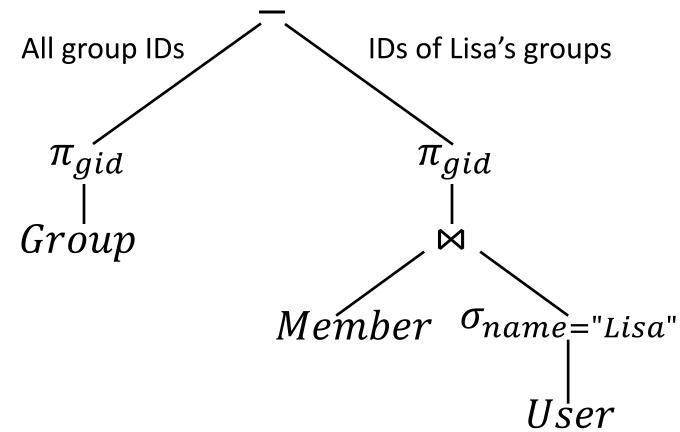
 Names of users in Lisa's groups name Writing a query bottom-up: Their names π_{name} Milhouse uid Lisa abc 123 Should we remove Lisa's name 857 gov Users in 857 gov π_{uid} Lisa's groups User uid gid name age pop 857 Lisa abc 857 Lisa 0.7 gov Lisa's groups Member name age pop Lisa 0.7 8 gid Who's Lisa? 123 gov $\sigma_{name} = "Lisa"$ Member 857 abc 857 gov

Another exercise

User (uid int, name string, age int, pop float)
Group (gid string, name string)
Member (uid int, gid string)

• IDs of groups that Lisa doesn't belong to

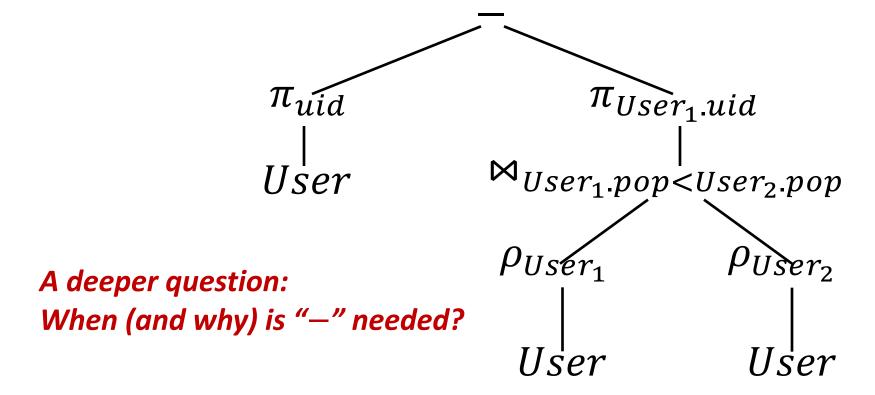
Writing a query top-down:



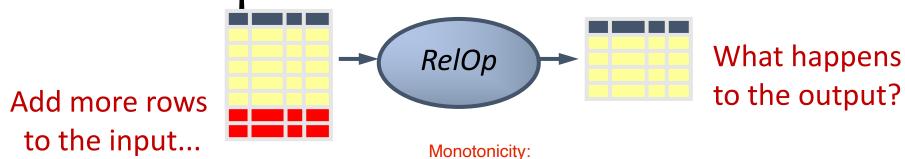
A trickier exercise

User (uid int, name string, age int, pop float)
Group (gid string, name string)
Member (uid int, gid string)

- Who are the most popular?
 - Who do NOT have the highest pop rating?
 - Whose pop is lower than somebody else's?



Monotone operators



Applying the relational operator on a new input will always output what you had earlier

- If some old output rows may need to be removed
 - Then the operator is non-monotone
- Otherwise the operator is monotone
 - That is, old output rows always remain "correct" when more rows are added to the input

 | Does monotonicity hold when removing rows from the input? |
 | Is there another property
- Formally, for a monotone operator op: $R \subseteq R'$ implies $op(R) \subseteq op(R')$ for any R, R'

Monotone operators



- If some old output rows may need to be removed
 - Then the operator is non-monotone
- Otherwise the operator is monotone
 - That is, old output rows always remain "correct" when more rows are added to the input

uid	gid		uid	gid
123	gov	_	123	gov
857	abc		857	gov
			857	abc

Monotone operators



- If some old output rows may need to be removed
 - Then the operator is non-monotone
- Otherwise the operator is monotone

gov

abc

• That is, old output rows always remain "correct" when more rows are added to the input

The old row is always

uid	gid		uid	gid		uid	gid		"correct" no matter
123	gov	<u> </u>	123	gov	=	857	abc /		what is added to R
857	abc		857	gov		693	abc	•	

Classification of relational operators

• Selection: $\sigma_p R$ Monotone

• Projection: $\pi_L R$ Monotone

• Cross product: $R \times S$ Monotone

• Join: $R \bowtie_{p} S$ Monotone

• Natural join: $R \bowtie S$ Monotone

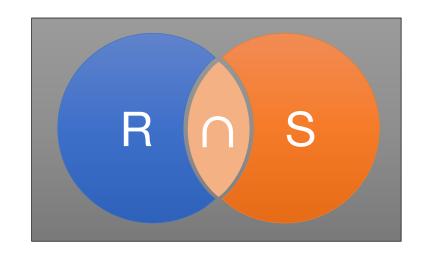
• Union: $R \cup S$ Monotone

• Difference: R - S Monotone w.r.t. R; non-monotone w.r.t S

• Intersection: $R \cap S$ Monotone?

Is intersection monotonic?

$$\bullet R \cap S = R - (R - S)$$



$$= \left(\begin{array}{c} R \\ - \end{array} \right)$$

Yes!

$$R_1 \subseteq R_2 \Rightarrow S \cap R_1 \subseteq S \cap R_2$$

Why is "—" needed for "highest"?

- Composition of monotone operators produces a monotone query
 - Old output rows remain "correct" when more rows are added to the input
- Is the "highest" query monotone?

 Ask professor if the reason we need the difference operator is
 - No! because finding the most popular person is not a monotonic query
 - Current highest pop is 0.9
 - Add another row with pop 0.91
 - Old answer is invalidated
- So it must use difference!

Why do we need core operator *X*?

- Difference
 - The only non-monotone operator
- Projection
 - The only operator that removes columns
- Cross product
 - The only operator that adds columns
- Union
 - The only operator that allows your to the data from both relations without expanding the tuple?
 - A more rigorous argument?

Ask the professor if the union operator is needed to group the data in two relations such that the joined info does not expand the tuples used in either relation?

- Selection?
 - Left as an exercise for the viewer

Selection reduces rows by filtering by a condition. Without this, we cannot use it in combination with the difference operator to only be left with the group of interest.

Extensions to relational algebra

- Duplicate handling ("bag algebra")
- Grouping and aggregation

- All these will come up when we talk about SQL
- But for now we will stick to standard relational algebra without these extensions

Why is RA a good query language?

- Simple
 - A small set of core operators
 - Semantics are easy to grasp
- Declarative?
 - Yes, compared with older languages like CODASYL
 - Though assembling operators into a query does feel somewhat "procedural"
- Complete?
 - With respect to what?

Complete means that the language can express the queries you want to ask

Relational Calculus

User (uid int, name string, age int, pop float)
Group (gid string, name string)
Member (uid int, gid string)

- $\{u.uid \mid u \in User \land \neg(\exists u' \in User: u.pop < u'.pop)\}$, or
- $\{u.uid \mid u \in User \land (\forall u' \in User: u.pop \ge u'.pop)\}$
- Relational algebra = "safe" relational calculus
 - Every query expressible as a safe relational calculus query is also expressible as a relational algebra query
 - And vice versa
- Example of an "unsafe" relational calculus query
 - $\{u.name \mid \neg(u \in User)\}$
 - Cannot evaluate it just by looking at the database

Relational Calculus

- RA and RC form the basis for "real" query languages (e.g. SQL), and for implementation
 - Relational Algebra
 - Queries are expressed by applying a sequence of operations to relations
 - More operational/procedural, very useful for representing how query executes
 - Relational Calculus
 - Let's users describe WHAT they want in first-order logic
 - Rather than HOW to compute it
 - Non-operational, declarative

Codd's Theorem

- Established equivalence in expressivity between:
 - Relational Calculus
 - Relational Algebra
- Why an important result?
 - Connects declarative representation of queries with operational description
 - Constructive: we can "compile" SQL into relational algebra

Limits of relational algebra

- Relational algebra has no recursion
 - Example: given relation Friend(uid1, uid2), who can Bart reach in his social network with any number of hops?
 - Writing this query in r.a. is impossible!
 - So RA is not as powerful as general-purpose languages (not Turing-complete)
- But why not?
 - Optimization becomes undecidable
 - Simplicity is empowering
 - Besides, you can always implement it at the application level (and recursion is added to SQL nevertheless (20))

What's Next

- Next class:
 - Database design in E/R model (recording)
- Sign up Piazza
 - https://piazza.com/sfu.ca/summer2024/cmpt354d100
 - Access code: qc76kdzsz9l
- Assignment 1 releasing soon
- Stay tuned for RA help tools