

Writing relational algebra queries

More general approach

1. Choose relations involved

- Ask yourself which relations need to be involved. Ignore the rest!
- Every time you combine relations, confirm that you specify the names of matching attributes (unless natural join)

Example 1

Person (name, age, gender)

Frequents (name, pizzeria)

Eats (name, pizza)

Serves (pizzeria, pizza, price)

- Find all pizzerias frequented by at least one person under the age of 18.
- We need pizzeria and age, joined on name
- Creating a new relation:

ExtendedFrequents (name, age, gender, pizzeria) = Person \bowtie Frequents

- We then apply the selection to ExtendedFrequents:

$\sigma_{\text{age} < 18}$ ExtendedFrequents

- And finally projection for desirable attributes

$\pi_{\text{pizzeria}}(\sigma_{\text{age} < 18}(\text{Person} \bowtie \text{Frequents}))$

2. Write intermediate relations with attributes and sample data

- Remember that selection checks one tuple at a time.
- If you need info from two different tuples, you **must** make a new relation where all the required info is in one tuple.
- Use assignment to define this intermediate relation.
- To visualize:
 - Draw an example of an intermediate relation with actual data in it.
 - Use good names for new relations.
 - Name the attributes on the LHS each time, so you don't forget what you have in hand.
 - Add a comment explaining exactly what's in the relation.

Example 2

Person (name, age, gender)

Frequents (name, pizzeria)

Eats (name, pizza)

Serves (pizzeria, pizza, price)

- Find all pizzerias that are frequented by only females or only males.

$$\text{PizzeriasFemales_MalesNotExcluded}(\text{pizzeria}) = \pi_{\text{pizzeria}}(\sigma_{\text{gender}='female'}(\text{Person}) \bowtie \text{Frequents})$$

$$\text{PizzeriaMales_FemalesNotExcluded}(\text{pizzeria}) = \pi_{\text{pizzeria}}(\sigma_{\text{gender}='male'}(\text{Person}) \bowtie \text{Frequents})$$

$$\text{PizzeriaFemalesOnly} = \text{PF_M} - \text{PM_F}$$

PF_M	PM_F
pizzeria	pizzeria
A	A
B	D
C	E
D	F

PFemales
pizzeria
B
C

3. Computing Max (min is analogous)

- Do self-product and find those that are not max
- Subtract from all to find the maxes

Example 3: 1/2

Person (name, age, gender)

Frequents (name, pizzeria)

Eats (name, pizza)

Serves (pizzeria, pizza, price)

- Find the pizzeria serving the cheapest pepperoni pizza. In the case of ties, return all of the cheapest-pepperoni pizzerias.

ServesPepperoni1 (pizzeria, price) = $\rho_{\text{ServesPepperoni1}}$

$[\pi_{\text{pizzeria,price}}(\sigma_{\text{pizza='pepperoni'}}\text{Serves})]$

ServesPepperoni2 (pizzeria, price) = $\rho_{\text{ServesPepperoni2}}$

$[\pi_{\text{pizzeria,price}}(\sigma_{\text{pizza='pepperoni'}}\text{Serves})]$

- Pair all tuples in SP1 with all other tuples in SP2 (**Cartesian product**, not a join):

ServesPepperoni1 x ServesPepperoni2

Example 3: 2/2

Person (name, age, gender)

Frequents (name, pizzeria)

Eats (name, pizza)

Serves (pizzeria, pizza, price)

- Select those that are **not min** price, because there are some pairs where $SP1.price > SP2.price$

$PizzeriasNotCheapestPP = \sigma_{SP1.price > SP2.price} (SP1 \times SP2)$

$Result = \pi_{pizzeria} ServesPepperoni1 - \pi_{SP1.pizzeria} (PizzeriasNotCheapestPP)$

4. Queries asking for “every”

- Make all combinations that include both every and some
- Subtract those that make it “not every”. The result is those who failed “every”.
- Subtract the failures from all to get a result

Example 4

Person (name, age, gender)

Frequents (name, pizzeria)

Eats (name, pizza)

Serves (pizzeria, pizza, price)

- Find the names of all people who frequent **every** pizzeria serving at least one pizza they eat.
- First, for each person – all pizzerias that serve pizzas the person eats

PotentialGoodPizzerias (name, pizzeria) = $\pi_{\text{name,pizzeria}}(\text{Eats} \bowtie \text{Serves})$

- Now need to find those people who do **not** frequent **every** good pizzeria, missing some that serve desirable pizzas:

NotEvery = $\pi_{\text{name}}(\text{PotentialGoodPizzerias} - \text{Frequents})$

Frequent**Every**GoodPizzeria = $\pi_{\text{name}}(\text{Person}) - \text{NotEvery}$

5. K or more

- Make k Cartesian products with itself and select rows where all k values are equal

Example 5

Person (name, age, gender)

Frequents (name, pizzeria)

Eats (name, pizza)

Serves (pizzeria, pizza, price)

- Find names of all pizzerias which serve **at least 2** pizzas that Amy can eat
- Pizzerias which serve desirable pizzas (**including** those that serve **only 1 good** pizza)

AllAmyPizzerias = $\pi_{\text{pizzeria, pizza}} (\sigma_{\text{name}='Amy'}(\text{Eats}) \bowtie \text{Serves})$

- Cartesian product of AAP with itself, select only rows where pizzerias are equal – and pizzas are different

$A1 = \rho_{A1}(\text{AllAmyPizzerias}), A2 = \rho_{A2}(\text{AllAmyPizzerias})$

$\text{AtLeast2} = \sigma_{A1.\text{pizza} > A2.\text{pizza} \text{ AND } A1.\text{pizzeria} = A2.\text{pizzeria}} (A1 \times A2)$

$\text{Answer} = \pi_{\text{pizzeria}}(\text{AtLeast2})$

6. Exactly k

- “ k or more” – “ $(k+1)$ or more”

Example 6

Person (name, age, gender)

Frequents (name, pizzeria)

Eats (name, pizza)

Serves (pizzeria, pizza, price)

- Find the names of all pizzerias which serve **exactly 2** pizzas that Amy can eat

$A1 = \rho_{A1}(\text{AllAmyPizzerias})$, $A2 = \rho_{A2}(\text{AllAmyPizzerias})$, $A3 = \rho_{A3}(\text{AllAmyPizzerias})$

- We have AtLeast2
- Compute at least 3:

$\text{AtLeast3} = \sigma_{A1.pizza > A2.pizza \text{ AND } A1.pizzeria = A2.pizzeria}$

$\text{AND } A1.pizza > A3.pizza \text{ AND } A1.pizzeria = A3.pizzeria$ (A1 x A2 x A3)

$\text{Exactly2} = \text{AtLeast2} - \text{AtLeast3}$

$\text{Answer} = \pi_{pizzeria}(\text{Exactly2})$

7. Related pairs

- Do Cartesian product and select the non-equal pairs joined on the desired shared attribute

Example 7

Person (name, age, gender)

Frequents (name, pizzeria)

Eats (name, pizza)

Serves (pizzeria, pizza, price)

- Find all pairs of customers who frequent the same pizzeria

$F1 = \rho_{F1}(\text{Frequents})$

$F2 = \rho_{F2}(\text{Frequents})$

$\sigma_{F1.pizzeria=F2.pizzeria \text{ AND } F1.name < F2.name}(F1 \times F2)$