#### **LECTURE 17**

# SQL II

SQL and Databases: An alternative to Pandas and CSV files.



# Agenda

- Filtering Groups
- EDA in SQL
- Joins
- IMDB Demo



# **SQL Query Structure**

```
SELECT <column expression list>
FROM 
[WHERE <predicate>]
[GROUP BY <column list>]
[ORDER BY <column list>]
[LIMIT <number of rows>]
[OFFSET <number of rows>];
```



# Summary so far



# **Filtering Groups**

- Filtering Groups
- EDA in SQL
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- IMDB Demo



## Filtering Groups With HAVING

What if we only want to keep groups that obey a certain condition?

**HAVING** filters groups by applying some condition across all rows in each group.

How to interpret: "keep only the groups **HAVING** some condition"

```
SELECT columns
FROM table
GROUP BY grouping_column
HAVING condition applied across group;
```



## Filtering Groups With HAVING

```
SELECT type, COUNT(*)
FROM Dish
GROUP BY type
HAVING MAX(cost) < 8;</pre>
```

# type COUNT(\*)

```
appetizer 3
dessert 1
```

```
name
               type cost
   ravioli
                       10
             entree
             entree
                       13
   ramen
     taco
             entree
edamame appetizer
     fries appetizer
potsticker appetizer
ice cream
            dessert
         Dish
```



#### WHERE vs. HAVING

SELECT type, COUNT(\*)
FROM Dish
WHERE cost < 8
GROUP BY type;</pre>

What will happen here?

type cost name ravioli entree 10 13 entree ramen entree taco edamame appetizer 4 fries appetizer 4 potsticker appetizer 4 dessert ice cream Dish

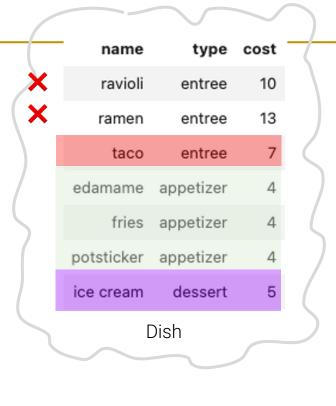
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SELECT type, COUNT(\*)
FROM Dish
WHERE cost < 8
GROUP BY type;</pre>

type	COUNT(*)
appetizer	3
dessert	1
entree	1

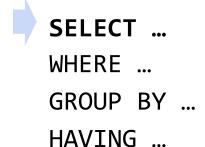




#### To filter:

- Rows, use WHERE.
- Groups, use **HAVING**.

WHERE precedes HAVING.



SELECT \*
FROM Dish
WHERE cost > 4
GROUP BY type
HAVING MAX(cost) < 10;

name	type	cost
ravioli	entree	10
ramen	entree	13
taco	entree	7
edamame	appetizer	4
fries	appetizer	4
potsticker	appetizer	4
ice cream	dessert	5
	Dish	



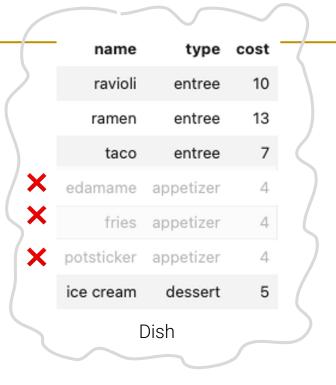
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WHERE precedes HAVING.

SELECT ...
WHERE ...
GROUP BY ...
HAVING ...

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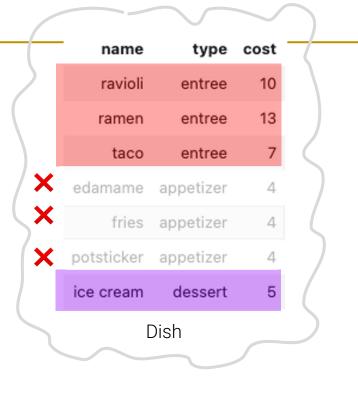
SELECT ...

WHERE ...

GROUP BY ...

HAVING ...

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FROM Dish
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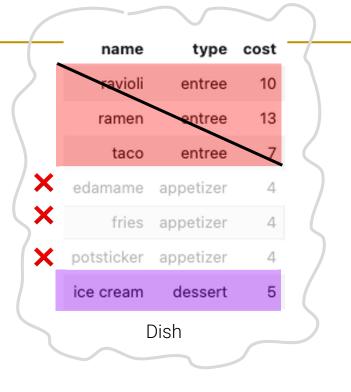
#### To filter:

- Rows, use WHERE.
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WHERE precedes HAVING.

SELECT ...
WHERE ...
GROUP BY ...
HAVING ...

SELECT \*
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HAVING MAX(cost) < 10;</pre>



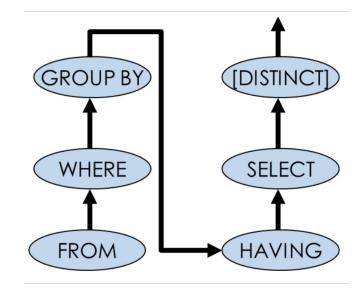


#### **Order of Execution**

A query is **not** evaluated according to Python operator precedence.

Generally, the order of execution of clauses within a statement are:

- 1. FROM: retrieve the relations.
- 2. WHERE: filter the rows.
- 3. GROUP BY: make groups.
- 4. HAVING: filter the groups.
- 5. SELECT: aggregate into rows, get specific columns.
- **6. DISTINCT**: enforce that results must be unique





# **New keywords**

```
SELECT <column expression list>
FROM 
[WHERE <predicate>]
[GROUP BY <column list>]
[HAVING <predicate>]
[ORDER BY <column list>]
[LIMIT <number of rows>]
[OFFSET <number of rows>];
```

- By convention, use all caps for keywords in SQL statements.
- Use newlines to make SQL code more readable.
- AS keyword: rename columns during selection process.
- WHERE: rows; HAVING: groups. WHERE precedes HAVING.



# Summary so far



#### Quick Check: WHERE vs. HAVING

What will be the return relation?

```
SELECT type, MAX(name)
FROM DishDietary
WHERE notes == 'gf'
GROUP BY type
HAVING MAX(cost) <= 7;</pre>
```



type MAX(name) type MAX(name) D. F. So appetizer appetizer fries fries mething else entree taco entree taco dessert ice cream



7



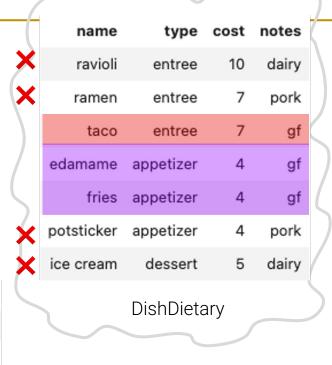
#### Quick Check: WHERE vs. HAVING

What will be the return relation?

SELECT type, MAX(name)
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WHERE notes == 'gf'
GROUP BY type
HAVING MAX(cost) <= 7;</pre>



E.





appetizer fries
entree taco
dessert ice cream

So mething else

F.



# **EDA in SQL**

- Filtering Groups
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- IMDB Demo



#### The IMDB Dataset

IMDB = "Internet Movie Database"

Contains information about movies and actors. For example, the Title table:

tconst	titleType	primaryTitle	originalTitle	isAdult	startYear	endYear	runtimeMinutes	genres
381681	movie	Before Sunset	Before Sunset	0	2004	None	80	Drama,Romance
81846	tvMiniSeries	Cosmos	Cosmos	0	1980	1980	780	Documentary
8526872	movie	Dolemite Is My Name	Dolemite Is My Name	0	2019	None	118	Biography, Comedy, Drama
309593	movie	Final Destination 2	Final Destination 2	0	2003	None	90	Horror,Thriller
882977	movie	Snitch	Snitch	0	2013	None	112	Action,Drama,Thriller
9619798	movie	The Wrong Missy	The Wrong Missy	0	2020	None	90	Comedy,Romance
1815862	movie	After Earth	After Earth	0	2013	None	100	Action,Adventure,Sci-Fi
2800240	movie	Serial (Bad) Weddings	Qu'est-ce qu'on a fait au Bon Dieu?	0	2014	None	97	Comedy
2562232	movie	Birdman or (The Unexpected Virtue of Ignorance)	Birdman or (The Unexpected Virtue of Ignorance)	0	2014	None	119	Comedy, Drama
356910	movie	Mr. & Mrs. Smith	Mr. & Mrs. Smith	0	2005	None	120	Action,Comedy,Crime



#### Working with Text: LIKE

We can perform simple text comparisons in SQL using the LIKE keyword

How to interpret: "look for entries that are LIKE the provided example string"

SELECT titleType, primaryTitle FROM Title WHERE primaryTitle LIKE "%Star Wars%";

titleType	primaryTitle
movie	Star Wars: Episode IV - A New Hope
movie	Star Wars: Episode V - The Empire Strikes Back
movie	Star Wars: Episode VI - Return of the Jedi
movie	Star Wars: Episode I - The Phantom Menace
movie	Star Wars: Episode II - Attack of the Clones
movie	Star Wars: Episode III - Revenge of the Sith

#### Two "wildcard" characters:

- % means "look for any character, any number of times"
- \_ means "look for exactly 1 character"



## **Converting Data Types: CAST**

To convert a column to a different data type, use the CAST keyword as part of the SELECT statement. Returns a *column* of the new data type, which we then SELECT for our output.

# SELECT primaryTitle, CAST(runtimeMinutes AS INT) FROM Title;

primaryTitle	CAST(runtimeMinutes AS INT)
A Trip to the Moon	13
The Birth of a Nation	195
The Cabinet of Dr. Caligari	76
The Kid	68
Nosferatu	94
Sherlock Jr.	45

Creates a copy of the column with all values of converted to the new data type. We then SELECT this column to include it in the output.

Similar to .astype in pandas



# **Applying Conditions: CASE**

We create conditional statements (like a Python if) using CASE

```
CASE WHEN <condition> THEN <value>
WHEN <other condition> THEN <other value>
...
ELSE <yet another value>
END
```

Conceptually, very similar to CAST – the CASE statement creates a new column, which we then SELECT to appear in the output.



# **Applying Conditions: CASE**

We create conditional statements (like a Python if) using CASE

```
SELECT titleType, startYear,

CASE WHEN startYear < 1950 THEN "old"
    WHEN startYear < 2000 THEN "mid-aged"
    ELSE "new"
    END AS movie_age
FROM Title;</pre>
```

All of this occurs within the SELECT statement

titleType	startYear	movie_age
movie	2010	new
movie	2019	new
movie	1998	mid-aged
movie	1989	mid-aged
movie	2017	new
tvSeries	1982	mid-aged
movie	1940	old



# **Joins**

- Filtering Groups
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#### **Multidimensional Data**

To minimize redundant information, databases typically store data across **fact** and **dimension tables** 

**Fact table:** central table, contains raw facts that typically have pure numerical values. It has information to link its entries to records in other dimension tables. Tends to have few columns, many records.

**Dimension table:** contains more detailed information about each type of fact stored in the fact table (each column). Tends to have more columns and fewer records than fact tables.

#### Products | Fact Table

drink_id	topping_id	store_id
3451	a	a236
6724	b	d462
9056	С	k378

#### **Drinks** | Dimension Table

7. SANDELONGE - COMMUNICIPAL COMMUNICATION OF THE PROPERTY OF			
drink_id	name	ice_level	sweetness
3451	Black Milk Tea	75	75
6724	Mango Au Lait	50	100
9056	Matcha Latte	100	100

#### Toppings | Dimension Table

topping_id	name
a	Brown Sugar Pearl
b	Lychee Jelly
С	Custard

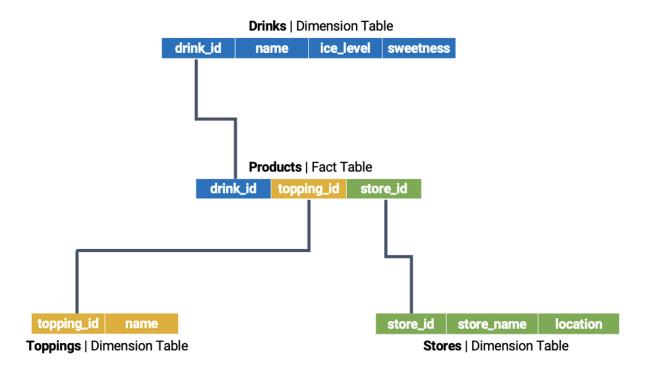
#### Stores | Dimension Table

store_id	store_name	location
a236	Sweetheart	Durant
d462	Feng Cha	Durant
k378	Yi Fang	Bancroft



#### **Multidimensional Data**

A structure that uses fact and dimension tables is called a **star schema** 





# Cats







Persian

Ragdoll

Bengal

id	name
)	Apricot
ı	Boots
2	Cally

Eugene

S



	•
id	breed
1	persian
2	ragdoll
4	bengal
5	persian



Pishi\*



#### **Inner Join**

In an **inner join**, we combine every row from the first table with its matching entry in the second table. If a row in one table does not have a match, it is omitted

S

id	name
0	Apricot
1	Boots
2	Cally
4	Eugene

t

id	breed
1	persian
2	ragdoll
4	bengal
5	persian

Match rows with the same ID across the tables. Exclude rows with no matching ID



#### **Inner Join**

In an **inner join**, we combine every row from the first table with its matching entry in the second table. If a row in one table does not have a match, it is omitted



This is the default behavior of pd.merge



## **JOIN Syntax**

Specify joins between tables as part of the FROM statement

SELECT \*
FROM table1 INNER JOIN table 2
ON table1.key = table2.key

Desired type of join

What columns to use to determine matching entries

	S		t
id	name	id	breed
0	Apricot	· 1	persian
1	Boots	2	ragdoll
2	Cally	4	bengal
4	Eugene	5	persian

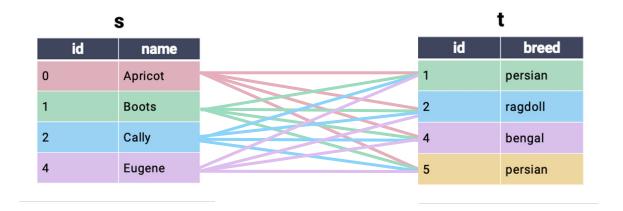
SELECT *
FROM s
INNER JOIN t
ON s.id = t.id

s.id	name	t.id	breed
1	Boots	1	persian
2	Cally	2	ragdoll
4	Eugene	4	bengal



#### **Cross Join**

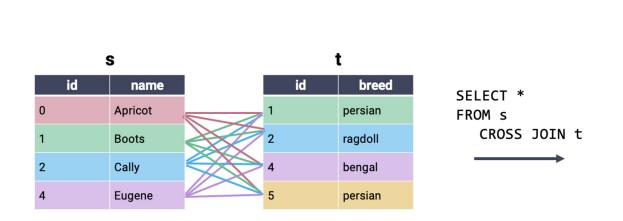
In a **cross join**, we find *every* possible combination of rows across the two tables. A cross join is also called a cartesian product.





#### **Cross Join**

In a **cross join**, we find *every* possible combination of rows across the two tables. A cross join is also called a cartesian product.



Notice that there is no need to specify a matching key (what columns to use for merging)

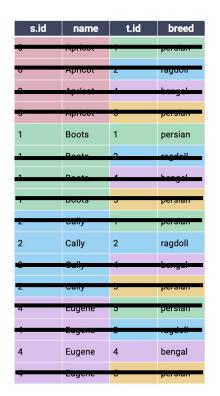
s.id	name	t.id	breed
0	Apricot	1	persian
0	Apricot	2	ragdoll
0	Apricot	4	bengal
0	Apricot	5	persian
1	Boots	1	persian
1	Boots	2	ragdoll
1	Boots	4	bengal
1	Boots	5	persian
2	Cally	1	persian
2	Cally	2	ragdoll
2	Cally	4	bengal
2	Cally	5	persian
4	Eugene	5	persian
4	Eugene	2	ragdoll
4	Eugene	4	bengal
4	Eugene	5	persian

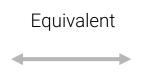


# Inner Join: Cross Join With Filtering

Conceptually, you can imagine an inner join as a cross join filtered to include only matching rows.

SELECT \*
FROM s CROSS JOIN t
WHERE s.id = t.id;





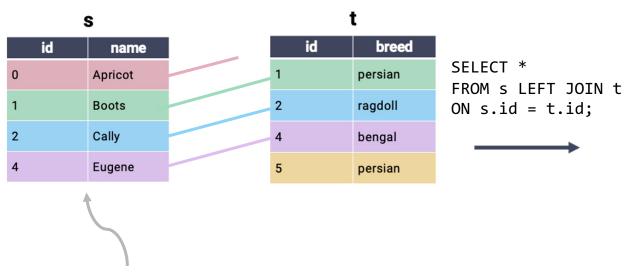
s.id	name	t.id	breed
1	Boots	1	persian
2	Cally	2	ragdoll
4	Eugene	4	bengal

SELECT \*
FROM s INNER JOIN t
ON s.id = t.id;



#### **Left Outer Join**

In a **left outer join** (or just **left join**), keep all rows from the left table and *only matching* rows from the right table. Fill NULL for any missing values.



Fill valu	ues witl	hout	
match	ing entr	ies in	the
right ta	able wit	h NUL	L

4

name

**Apricot** 

Boots

Cally

Eugene

t.id

breed

persian

ragdoll

bengal

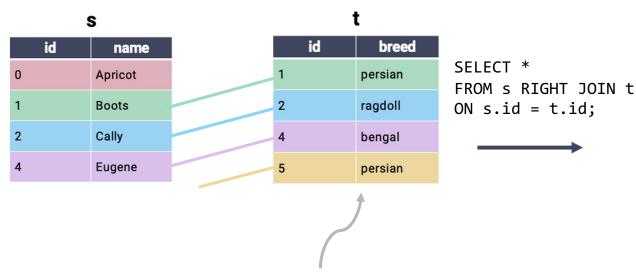
s.id

The "left table" is whichever table is referenced first in the JOIN statement.



## **Right Outer Join**

In a **right outer join** (or just **right join**), keep all rows from the right table and *only matching* rows from the right table. Fill NULL for any missing values.



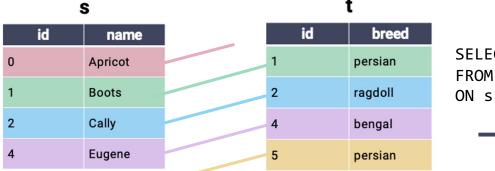
s.id	name	t.id	breed
1	Boots	1	persian
2	Cally	2	ragdoll
4	Eugene	4	bengal
_	_	5	persian

The "right table" is whichever table is referenced second in the JOIN statement.



#### **Full Outer Join**

In a **full outer join**, keep *all rows* from both the left and right tables. Pair any matching rows, then fill missing values with NULL. Conceptually similar to performing both left and right joins.



s.id	name	t.id	breed
0	Apricot	-	-
1	Boots	1	persian
2	Cally	2	ragdoll
4	Eugene	4	bengal
-	-	5	persian



When working with long table names, we often create aliases that are easier to refer to (just as we did with columns yesterday).

SELECT primaryTitle, averageRating
FROM Title AS T INNER JOIN Rating AS R
ON T.tconst = R.tconst;

We can then reference columns using the aliased table names

primaryTitle	averageRating
A Trip to the Moon	8.2
The Birth of a Nation	6.3
The Cabinet of Dr. Caligari	8.1
The Kid	8.3
Nosferatu	7.9
Sherlock Jr.	8.2
Battleship Potemkin	8.0
The Gold Rush	8.2
Metropolis	8.3
The General	8.1



When working with long table names, we often create aliases that are easier to refer to (just as we did with columns yesterday).

	primaryTitle	averageRating
<pre>SELECT primaryTitle, averageRating FROM Title AS T INNER JOIN Rating AS R ON T.tconst = R.tconst;</pre>	A Trip to the Moon	8.2
	The Birth of a Nation	6.3
	The Cabinet of Dr. Caligari	8.1
	The Kid	8.3
The <b>AS</b> is actually optional! We usually include it for clarity.	Nosferatu	7.9
	Sherlock Jr.	8.2
	Battleship Potemkin	8.0
SELECT primaryTitle, averageRating FROM Title T INNER JOIN Rating R ON T.tconst = R.tconst;	The Gold Rush	8.2
	Metropolis	8.3
	The General	8.1

Why bother aliasing?

Referencing columns in the format table\_alias.column\_name avoids any ambiguity if both tables have a column with the same name.

Example: both the Title and Rating tables include a column named tconst

SELECT primaryTitle, averageRating FROM Title INNER JOIN Rating ON tconst = tconst;

Running query in 'sqlite:///data/imdbmini.db' (sqlite3.OperationalError) ambiguous column name: tconst



Example: both the Title and Rating tables include a column named tconst



SELECT primaryTitle, averageRating
FROM Title AS T INNER JOIN Rating AS R
ON tconst = tconst;



Should we look at the **tconst** column from the Title table or from the Rating table?



SELECT primaryTitle, averageRating
FROM Title AS T INNER JOIN Rating AS R
ON T.tconst = R.tconst;



# **IMDB** Demo

- Filtering Groups
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- Joins
- IMDB Demo



# **Demo Slides**

# **Typical Database Workflow**

- Query large amounts of data from a database using SQL. Write SQL queries to perform broad filtering and cleaning of the data
- After querying data, use pandas to perform more detailed analysis (visualization, modeling, etc.)

