INFO 290T

Human-in-the-loop Data Management aka Data Engineering



Today's Agenda

- The essentials
- Bird's eye view of the class material
- Getting to know you
- Classical database systems: an introduction



The Essentials I

- Instructor: Aditya Parameswaran
- Assistant Professor, I School and EECS
- Office: 212 South Hall
- Email: adityagp@berkeley.edu
 - Mention "INFO290" in the title! (will help ensure a response)
- Meeting slots: M/W 9-10.30am @ 210 South Hall
 - Class slides will be posted after class
- Office hours: W 10.30-12pm @ 212 South Hall
 - Also on demand if need be



The Essentials II

- GSI: Doris Lee
- PhD Student, I School
- Email: dorislee@berkeley.edu
 - Mention "INFO290" in the title!
- Office Hours: TBD

• Website: https://info290.github.io



The Essentials: Prerequisites

• "Experience with programming, a basic understanding of computer systems, data structures, and algorithms expected."



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- "Experience with programming"
 - Python (plus ideally Java), web programming
 - Unix/shell scripting, installing and using software packages
- "computer systems"
 - At a high level: OS, disk, memory, processor, networks
- "data structures and algorithms"
 - O(n), sorting, graph searching (DFS/BFS), hashing, dynamic programming



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- That said, I am happy to have people from different backgrounds
 - Talk to me if you're not sure
 - Be prepared to read up on concepts outside of class



What sort of course is this?

- A lot of the content is experimental
 - Not taught in typical database classes
 - Knowledge and tools you pick up by working on industry data systems
 - Rapidly evolving field, so we will try to focus on the timeless concepts
 - A human-centered perspective: emphasizing usability and expressiveness
- Non-traditional evaluation: 5 assignments and a project
 - So no "exams"
- Lecture-oriented, but some of the content will be more "half-baked"



Goals of the Class

- A high-level overview of data systems for various stages of data science, from data preparation to actionable insight
- A modern take on a database class that reflects the rich diversity of data systems for "dealing with data at scale" beyond traditional databases



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- A high-level overview of data systems for various stages of data science, from data preparation to actionable insight
- A modern take on a database class that reflects the rich diversity of data systems for "dealing with data at scale" beyond traditional databases
- Takeaways:
 - "Pick the right tool for the job" when to use which data system.
 - Evaluating data systems: usability, power, scalability
 - The concepts underlying the tools
- This class will **not** teach you how to install or use tool X,Y, or Z
 - Unfortunately, the data systems are constantly evolving, and any such knowledge is likely to be stale in a couple of years
 - Instead, we will emphasize fundamental concepts, which will remain timeless



Relationship to INFO257 (Database Sys.)

- This class will be roughly equivalent to INFO257 (Database Systems)
- If you are a MIMS student, you are free to pick either one to fulfill requirements.
- INFO 257 is a standard, classic, and polished take on database systems;
- INFO 290 is a more modern take, emphasizing user-centered and endto-end lifecycle aspects
 - but beware of hiccups that come with a new class!



Please Beware of Hiccups!

- Three problems:
 - A new class
 - A class on new non-traditional material
 - The first time I'm teaching at Berkeley!

Please Beware of Hiccups!

- Three problems:
 - A new class
 - A class on new non-traditional material
 - The first time I'm teaching at Berkeley!
- So expect problems to arise! (And please be tolerant!)
 - Bugs, lack of clarity, changing instructions, ...
- Please feel free to provide feedback to Doris and/or me at any time
- From our end, we will be as tolerant as we can



Grading Breakdown

- Assignments = 60%
 - 3 Written, each counting for 10%, evaluating roughly a 1/3rd of the class material each
 - 2 Programming-oriented, each counting for 15%, typically evaluating the use of one or more data systems each.
- Project = 35%
 - Implementation or research oriented
- Class participation = 5%



Assignments

- 3 Written, each counting for 10%, evaluating roughly a 1/3rd of the class material each
 - Several questions testing what was covered in class and extending it
 - The primary means of evaluating your understanding of class material
- 2 Programming-oriented, each counting for 15%, typically evaluating the use of one or more data systems each.
 - Hands-on experience
 - To be done in teams of two (but talk to us if you need exceptions)



Late Policy

- Up to four late days, to be used in any way that makes sense, without any explanation.
- Any submission after the deadline will be rounded up to the closest day (i.e., 24 hour), so even one hour late will count as the use of a full day.
- After four late days, students will lose 1% of the grade per day late.
- In case of medical issues, please contact the instructors as soon as you are able.



Project

- Build/design/test something new and cool!
- Can be research or development oriented
 - Research-oriented: doing work that might lead to a paper
 - Should be "original"
 - Dev-oriented: pick a real-world application and address it end-to-end using data systems
- Ideally, teams of two
- Midterm report (5%, due 3/20 midnight)
- Final report + presentation (30%, report due: 05/01 midnight; presentation slides due: 4/27 prior to class)



Project Option I: Research-Oriented

- Goal: have something "publishable" (ish) for a database or HCI venue
- Spectrum of contributions:
 - Mainly algorithmic
 - Mainly system or tool-building oriented
 - Or both
- Important to build on state of the art and extend it
 - Simply reproducing existing work not sufficient
 - Evaluating why it is "better" than prior work is needed
 - Usability, Performance, Accuracy, Expressiveness (or multiple)
 - Getting your hands dirty is a must!



Project Option I: Research-Oriented

- Rough phases:
 - Identify problem (we can help!)
 - Explore related work/tools
 - Prototype
 - Build & Evaluate
 - Write "paper"



Project Option I: Research Oriented

- Build a new scalable graph visualization system
- Design a spreadsheet-computational notebook hybrid
- Develop an extension to databases to support seamless schema changes
- Design a real-time spreadsheet system
- Build an automatic dashboarding tool

•



Project Option 2: Application-Oriented

- Goal: developing a data systems application
 - Extending one or more "data systems" to a real application
 - Applying techniques from class in a "hands-on" way
- Rough phases:
 - Identify problem (we can help!)
 - Mock-up design and requirements
 - Prototype
 - Build & Evaluate
 - Write "report"



Project Option 2: Application-Oriented

- Building a historical news archive querying database
- Developing a new medical record management database
- Designing the database of MIMS capstone projects
- Develop a course-textbook swap portal
- Design an interface to support retrieval of legal documents
- Design a scalable protein-protein interaction visualizer

• ...



Class Participation

- Goals: assess understanding, get feedback, make the class more lively
- Not essential that
 - you ask "good" questions
 - answer questions "correctly"
 - make "intelligent" points
 - you have to attend and be super engaged every class
- Any participation is good participation!
- Hard to come up with a rubric for class participation, but I promise to not be dogmatic about this, nor grade strictly... the goal is to just ensure that you are engaged with the class material



Any Questions?

• Next, an overview of class content

Data is Central to Our Everyday Lives



Data has the potential to inform decisions and improve lives across sectors

- Medicine
- Government
- Science
- Technology
- Commerce
- Manufacturing
- Education
- Journalism
- Social Sciences

• ...



Data is Central to Our Everyday Lives: But...



There's a lot of it!

Every minute there's

- 0.5M tweets tweeted
- 4M You Tube videos watched
- 120 new LinkedIn profiles created
- 16M text messages sent
- 600 new page edits to Wikipedia
- 18M new requests to The Weather Channel
- 50K new Uber rides



Data is Central to Our Everyday Lives: As a result...



"Organizations are drowning in data, but starving for insight."

[Forrester report "Digital Insights: The New Currency of Business", 2016]

• "An organization with data but no one to analyze it cannot take advantage of it."

[Hal Varian "Al, Economics, & Industrial Organization", 2018]

• "40% of majors need to be data-savvy, up from 10%." [McKinsey Institute report "Age of Analytics: Competing in a Data-driven World", 2017]

The Goal of this Class: Becoming "Data Savvy"!

- This class is meant to teach you the basics of data engineering, i.e., how to store, manage, structure, clean, query, and make sense of data at scale
- A "systems/engineering"-oriented complement to your more algorithmic machine learning classes (which is also essential for data savviness)

Algorithms (ML/Stats)

Systems/Engineering



Another Goal: Emphasizing "Human-in-the-loop"

Why the fuss about humans?

- Humans are the ones analyzing data
- Reasoning about them "in the loop" of data science is crucial
- Traditional database research and practice ignores the human aspects!



The Goal of this Class: Becoming "Data Savvy"!

- We start with classical database systems
- And then move to non-traditional data engineering
 - Streaming systems
 - Graph processing systems
 - Column stores
 - Visualization systems
 - Dataframe systems
 - •

Algorithms (ML/Stats)

Systems/Engineering



Classical Database Systems? What are those?

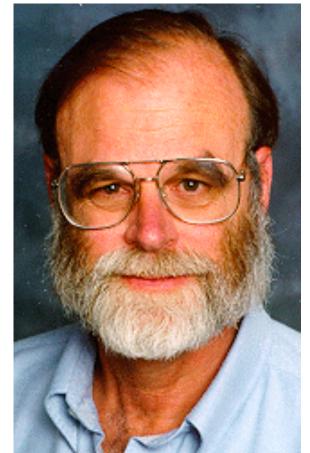
- The database systems field began in the 60s...
- Really hit its stride in the 70s thanks to a clean new "data model" a way of representing and thinking about data
 - The "relational" model: essentially a model of relations (or sets)
 - Led to the field of relational databases, with several commercial and open-source database systems
- Led to several software systems "powered" by relational databases
 - Banking systems
 - Flight reservation systems
 - Payment systems
- Relational databases are still at the "core" of the technology stack of most companies today.



Classical Database Systems: A Brief Berkeley Side Note...

The last two Turing awards for databases came from Berkeley!

- Jim Gray ('98) was the first CS PhD student from Berkeley; pioneer of transaction processing
- Mike Stonebraker ('14) was faculty for many years at Berkeley; influenced or developed many popular database systems





Classical Database Systems: Are we done?

Most modern data scientists work outside of (classical) relational databases.

Most use a combination of files + scripts + Excel + Python/R

Discussion Question: Why is that?

(Since many of you have experience doing "data science" in industry)

And if you're encountering databases for the first time, don't worry, you'll know this first-hand later on in the class!



Classical Database Systems: Are we done?

Most modern data scientists work outside of relational databases.

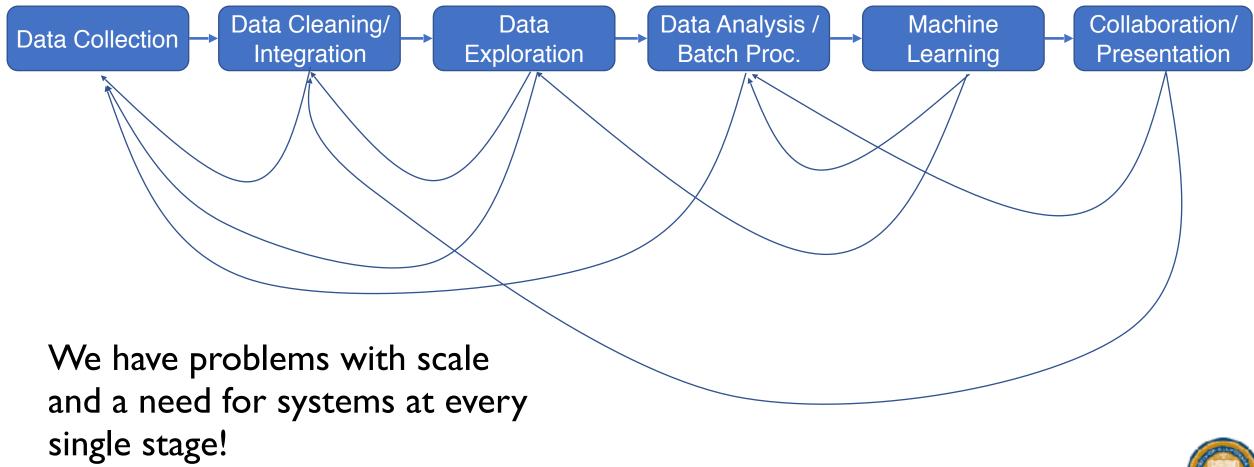
Discussion Question: Why is that?

- Hard to use/learn
- Does not scale
- Not easy to do quick and dirty data analysis
- Does not deal well with ill-formatted, noisy, or unstructured data
- Loading times are high
- Cannot do machine learning or data analysis easily
- Does not do well with special data formats

At a high-level, they are overly rigid, and do not support other stages of the data science lifecycle well.



The Modern Data Science Lifecycle



Beyond Classical Database Systems

The second half of the class will cover a range of non-traditional data systems

Data preparation: dealing with noisy data

Semi-structured and unstructured data

Data cleaning and integration

Data exploration: performing "quick-and-dirty" analysis

Data frames and Spreadsheets

Summarization, OLAP, and visualization

Batch analysis: doing ad-hoc analysis at scale

Map-reduce and parallel data processing

Column stores and data compression

Real-time analysis: dealing with data that is evolving or too large

Streaming systems

Sketching and approximation

Specialized analysis requirements

Graph processing and machine learning

Security and privacy



Recap

- Data is central to our lives
- Data-savviness is the future!
 - Classical (relational) database systems
 - Non-traditional database systems and the data life-cycle
- **Next**: why relational databases



Questions?

• Let's start with Classical (Relational) Databases!



First, what is a Database System?

DBMS (Data Base Management System)

= Database System = Database

System to manage, maintain, query, interact with, transact with data.

More loosely, database systems are used for "data management"



Task: Build a Banking "Data Management" System from Scratch w/o a Database

Goal: Manage customers, accounts, joint accounts, transfers, transactions, interest rates.

Let's say I implement this system using C++/Java/Python, without using a database system

Think like a designer: what aspects do we need to worry about?

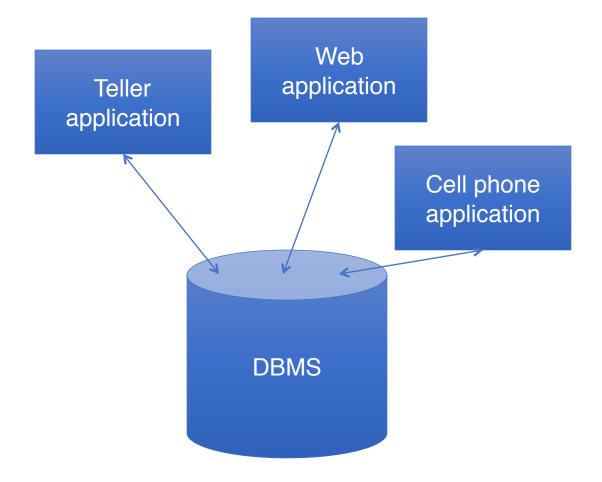
Aspects to worry about

- Deal with lots of data
- Be fast
- Don't lose information
- Allow multiple users
- Stay consistent
- Easy to use



The Database Approach: Abstraction

- Abstract out all of the data management functionality into a separate layer
- Many applications can access it
- Turns out this "separate layer" keeps turning up in many many many scenarios
- Makes sense to abstract it out





One possible definition:

System for providing EFFICIENT, CONVENIENT, and SAFE, MULTI-USER storage of and access to MASSIVE amounts of PERSISTENT data



System for providing EFFICIENT, CONVENIENT, and SAFE, MULTI-USER storage of and access to MASSIVE amounts of PERSISTENT data

Data: information on accounts, customers, balances, current interest rates, transaction histories, etc.

MASSIVE: many TBs at a minimum for big banks; more if we keep history of all transactions; even more if we keep images of checks!



System for providing EFFICIENT, CONVENIENT, and SAFE, MULTI-USER storage of and access to MASSIVE amounts of PERSISTENT data

PERSISTENT: data lives on, beyond programs that operate on it, even on system shutdown and power failure.

→ Can't store data in memory, we have to rely on stable storage (disk, flash)



System for providing EFFICIENT, CONVENIENT, and SAFE, MULTI-USER storage of and access to MASSIVE amounts of PERSISTENT data

MULTI-USER: many people/programs accessing same database, or even same data, simultaneously → Need controls

- Bob @ ATM2: withdraw \$50 from account #002 get balance from database; if balance >= 50 then balance := balance 50; // dispense cash update balance in database;
- Initial balance = 100.What's the ideal case? What could go wrong?



System for providing EFFICIENT, CONVENIENT, and SAFE, MULTI-USER storage of and access to MASSIVE amounts of PERSISTENT data

SAFE:

• from system failures. E.g., money should not disappear or appear from the account, due to a power failure!

```
Bob @ ATM2: withdraw $50 from account #002

get balance from database;

if balance >= 50

then balance := balance - 50; // dispense cash failure right here

update balance in database;
```

from malicious users

System for providing EFFICIENT, CONVENIENT, and SAFE, MULTI-USER storage of and access to MASSIVE amounts of PERSISTENT data

CONVENIENT:

- simple commands to debit account, get balance, write statement, transfer funds, etc.
- also unpredicted queries should be easy
- shouldn't require complex 100s of lines of code

EFFICIENT:

 don't search all files in order to get balance of one account, get all accounts with low balances, get large transactions, etc.



Why Direct Implementation is Super Hard/Won't Work

- Early DBMS evolved from file systems
- Provided storage of MASSIVE amounts of PERSISTENT data, to some extent

• SAFE?

 when system crashes, no guarantees on how program may behave: we may lose data

EFFICIENT?

• Does not intrinsically support fast access to data whose location in file is not known: will need to write custom code



Why Direct Implementation is Super Hard/Won't Work

- Early DBMS evolved from file systems
- Provided storage of MASSIVE amounts of PERSISTENT data, to some extent

CONVENIENT?

- need to write a new C++/Java program for every new query
- small changes to structure entails changing file formats; need to rewrite virtually all applications

MULTI-USER ACCESS?

- limited protection
- need to worry about interfering with other users



This is why Database Systems were Invented!

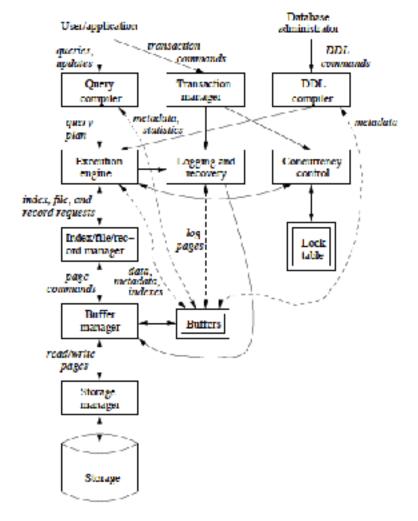
- Buy, install, set up for particular application or applications
- Major relational database vendors:
 - Oracle
 - Microsoft SQL Server
 - IBM DB2
- Open source relational databases:
 - Postgres
 - MySQL
 - Sqlite



A Complex Piece of Software

 We will cover "bird's eye view" of database systems, emphasizing the user perspective

• For the "database internals" perspective, consider taking CS186/286 and beyond!





Recap

- Data is central to our lives
- Data-savviness is the future!
 - Classical database systems: today's focus!
 - Non-traditional database systems and the data life-cycle
- Notion of a DBMS
- Requirements for a DBMS
- Example DBMSs, and their complexity
- **Next**: the relational data model



Fundamentals: The Relational Data Model

- A DBMS stores many databases
- Each database has many relations
- A relation = a set of tuples, each w/ a predefined set of attributes

Table/Relation (an instance of)

Attributes/Columns

Name	Price	Category	Manufacturer
Gizmo	\$19.99	gadgets	GizmoWorks
Power Gizmo	\$29.99	gadgets	GizmoWorks
SnapPea Camera	\$149.99	photography	Canon
Hi Phone	\$700.99	phone	Snapple

Tuples/ Rows/ Records



Fundamentals: The Relational Data Model

- Each relation is defined by a set of attributes
 - Each attribute has a type, called domain
 - The domain must be atomic: string, integer, ... (but many DBs violate this)
- Each relation contains a set of tuples or records
 - Each tuple has values for each attribute

Table/Relation (an instance of)

Attributes/Columns

tance oi)	Name	Price	Category	Manufacturer
	Gizmo	\$19.99	gadgets	GizmoWorks
Tuples/	Power Gizmo	\$29.99	gadgets	GizmoWorks
Rows/ Records	SnapPea Camera	\$149.99	photography	Canon
	Hi Phone	\$700.99	phone	Snapple



Equivalent Representations

• Since a relation has a set of attributes and a set of tuples, we can reorder both and not change the relation

Name	Price	Category	Manuf.
Hi Phone	\$700.99	phone	Snapple
SnapPea Camera	\$149.99	photography	Canon
Power Gizmo	\$29.99	gadgets	GizmoWorks
Gizmo	\$19.99	gadgets	GizmoWorks

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Relational Schema & Instance

- Schema: The structure, format, or scaffolding
 - Schema for a Relation:
 - Relation names plus attribute names & domains
 - Product (Name String, Price Float, Category String, Manuf. String)
 - Schema for a Database:
 - Schemas for many relations
 - Product (...)
 - Vendor (Name String, Address String, Phone Integer)
- Instance: Specific instantiation of the Relation or Database
 - A Relation or Database with "values filled in"

Name	Price	Category	Manuf.
Hi Phone	\$700.99	phone	Snapple
SnapPea Camera	\$149.99	photography	Canon



Recap: The Relational Data Model

- Data is stored in relations
- Relations are defined by a set of attributes/columns w/ specific domains
 - Together this defines the "schema" of a relation
- At any time, relations comprise of a set of tuples/rows/records
 - This constitutes the "instance" of a relation
- Databases are collections of relations
- Next, we will define the Relational Algebra: the operations that manipulate relations (the operands)
 - Parallels to linear algebra or set algebra



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- Notion of a DBMS
- Requirements for a DBMS
- Example DBMSs, and their complexity
- The relational data model
- **Next**: relational algebra



Set Operations: Union and Difference

- Union & Difference
 - Binary operations from set algebra
 - Recall that relations are sets
- Union
- Notation: $R1 \cup R2$
 - RI and R2 must have same schema
- Output schema: same as input
- Example: CurrentStudents (Name, Addr) U Alumni (Name, Addr)
- Difference is similar R1-R2 : tuples in R1 and not in R2



Unary Op.: Selection (Cross out rows)

- Returns all tuples that satisfy a condition (also called filter)
- Notation $\sigma_{c}(R)$
 - Condition C can involve attributes in R with =, >, <, AND, OR, etc.
- Output schema: same as input
- Find all employees with salary greater than \$40,000 from an Employee relation



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$$\sigma_{Salary>40000}(Employee)$$



Another Selection Example

SSN	Name	DepID	Salary
839930303	Avi	1	100000
182902033	Sarah	2	90000
383930032	Farida	2	120000
123233849	Jacob	1	70000

• Find employees with Salary greater than 100000 and Department ID 2.



Another Selection Example

SSN	Name	DepID	Salary
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Find employees with Salary greater than 100000 and Department ID
2.

$$\sigma_{Salary>100000\ AND\ DepID=2}(Employee)$$



Unary Op.: Projection (Cross Out Columns)

- Notation $\Pi_{A_1,A_2,...,A_n}(R)$

• Where
$$R(B_1,B_2,\ldots,B_m)$$

$$A_1, A_2, \dots, A_n \subseteq B_1, B_2, \dots, B_m$$

- That is, you can only sub select from columns that are present in R
- Output Schema (A_1, A_2, \ldots, A_m)
- Example: project out only the names from the Employee (SSN, Name, DepID, Salary)



Unary Op.: Projection (Cross Out Columns)

- Notation $\prod_{A_1,A_2,...,A_n} (R)$

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- That is, you can only sub select from columns that are present in R
- Output Schema (A_1,A_2,\ldots,A_m)
- Example: project out only the names from the Employee (SSN, Name, DepID, Salary) $\Pi_{Name}(Employee)$
- Note: Deletes any duplicate rows to return a set!



Another Projection (+ Selection) Example

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839930303	Avi	1	100000
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• Find names and salaries of Employees whose Department ID is 2.



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Name	Salary
Sarah	90000
Farida	120000

• Find names and salaries of Employees whose Department ID is 2.

$$\Pi_{Name,Salary}(\sigma_{DepID=2}(Employee))$$



Other Operators... Next week!

