#### DATABASE FUNDAMENTALS (RECALL/CRASH COURSE)

**Database functionalities** 

#### **Database Management Systems**

- Functionality provided
  - What kind of data can I put in? Relations/documents/pairs...
  - How can I get data out of it?
     query languages/API
  - How does it handle concurrent access?

ACID (or less)

– How long does a given operation take?

**Query execution, optimization** 

• Implementation (internals)

How does it cope with scale?
 for reads?
 Smart storage

for reads? Smart storage and indexing structures for writes? Concurrency control

#### Relational Database Management Systems

- Functionality provided
  - What kind of data can I put in?
  - How can I get data out of it?
     SQL query language
  - How does it handle concurrent access?

ACID (or less)

Relations

– How long does a given operation take?

#### **Query optimization**

- Implementation (internals)
  - How does it cope with scale?

for reads? Smart storage and indexing structures for writes? Concurrency control

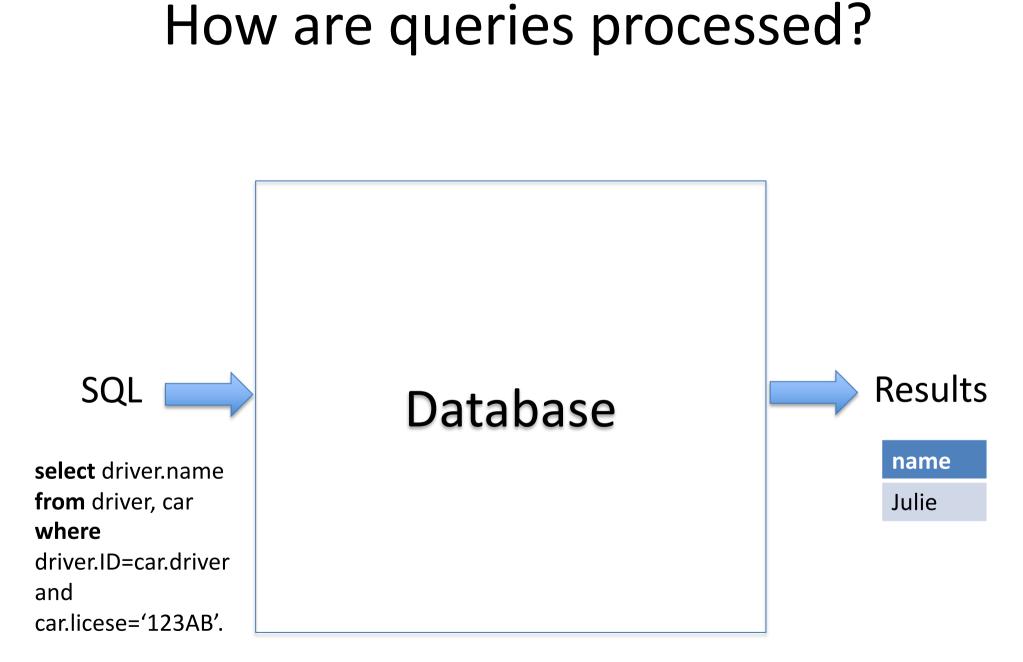
#### Fundamental database features

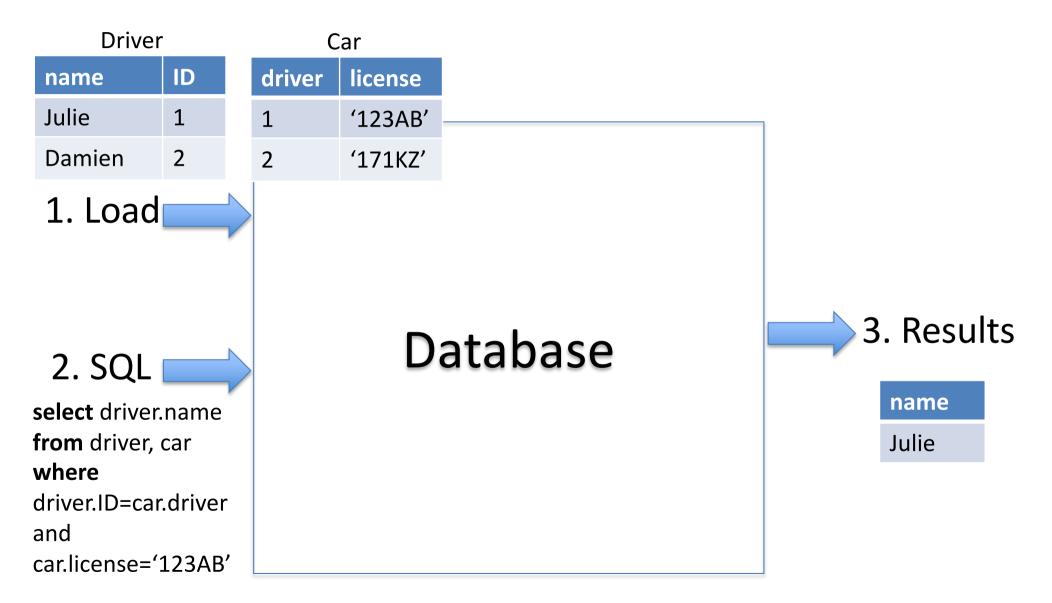
#### 1. Data storage

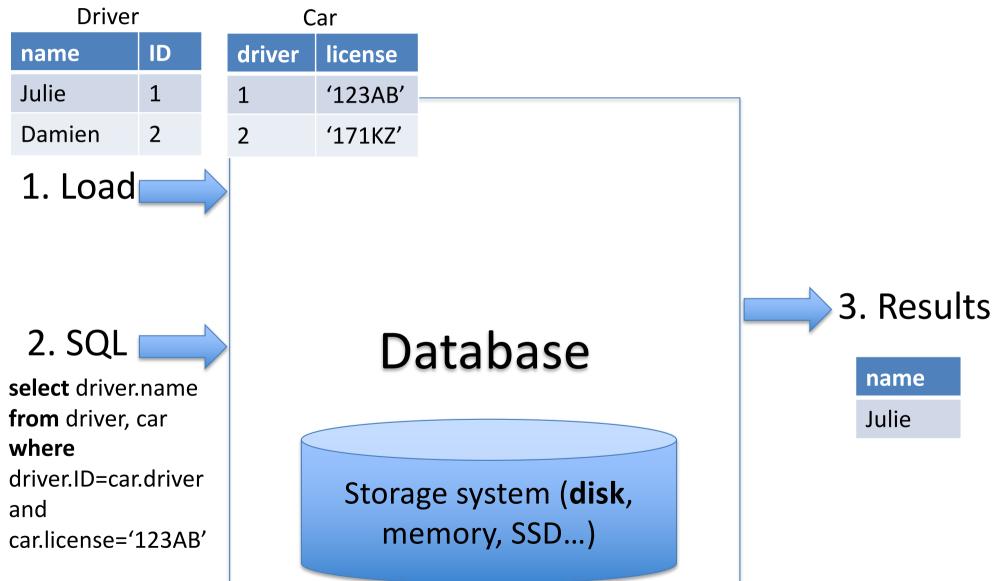
- Protection against unauthorized access, data loss
- 2. Ability to at least **add** to and **remove** data to the database
  - Also: updates; active behavior upon update (triggers)
- 3. Support for **accessing** the data
  - Declarative query languages: say what data you need, not how to find it

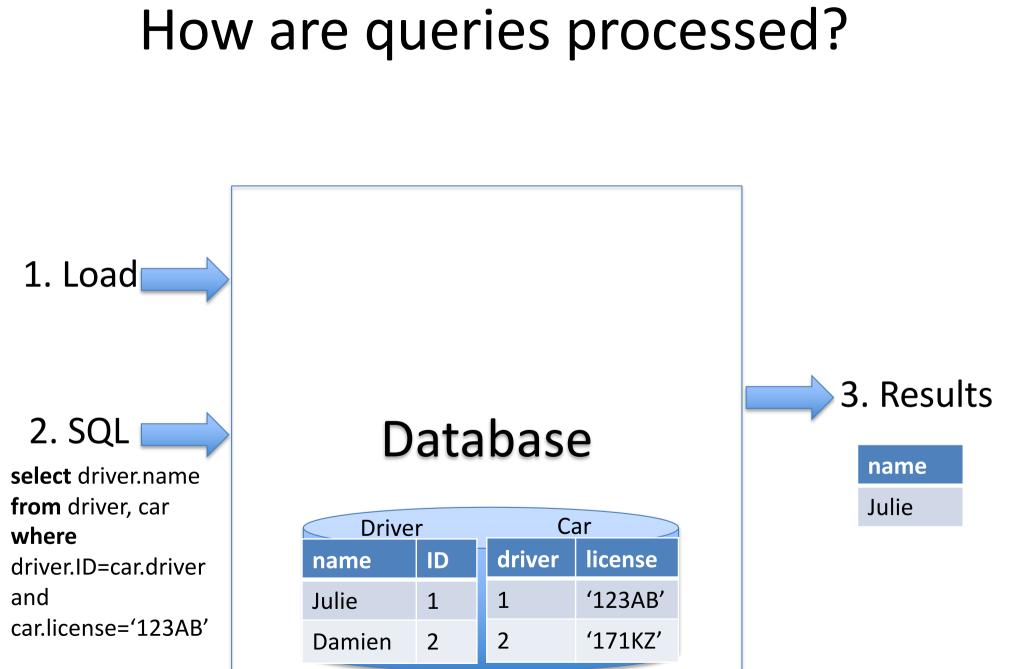
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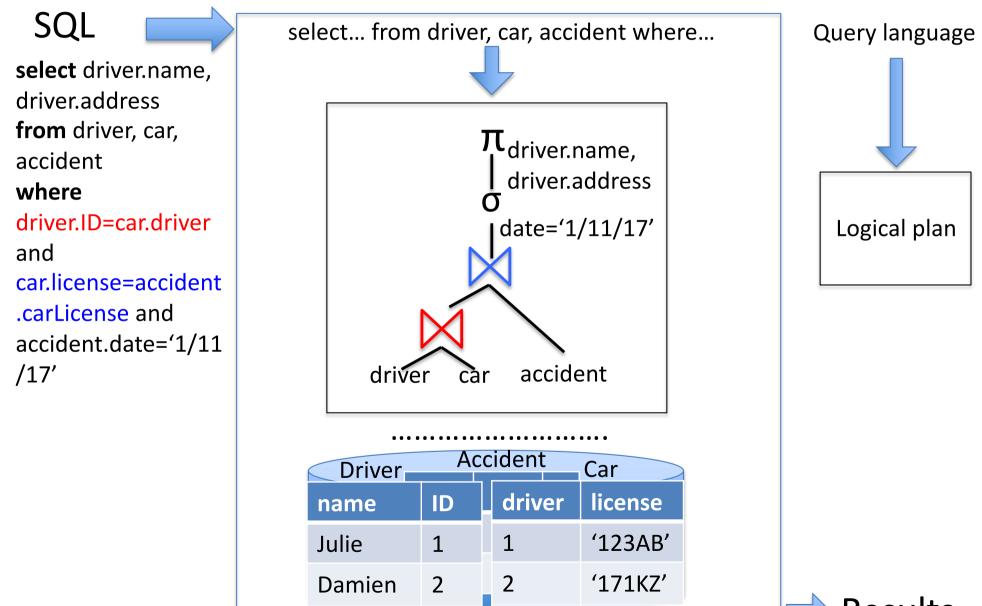
**Query processing** 







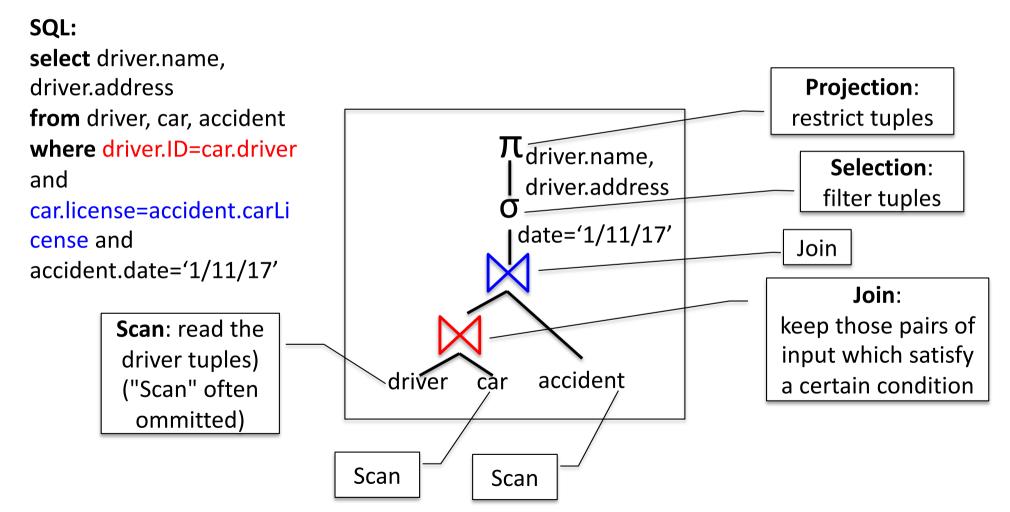






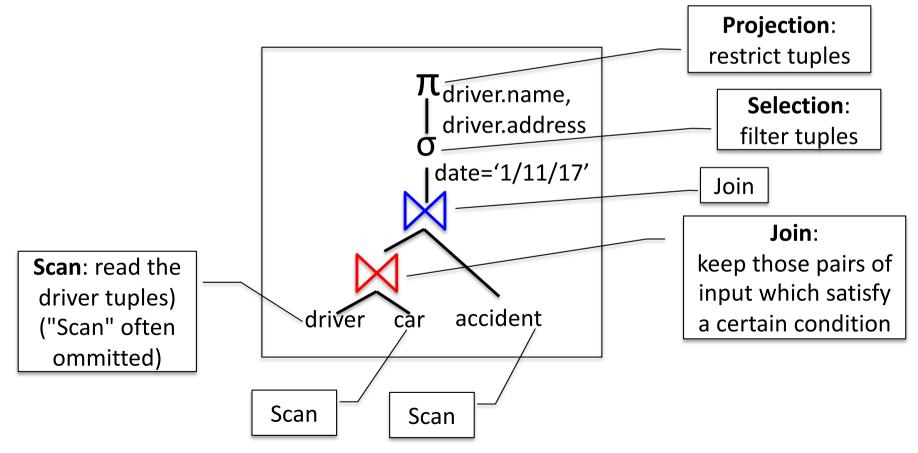
#### Logical query plans

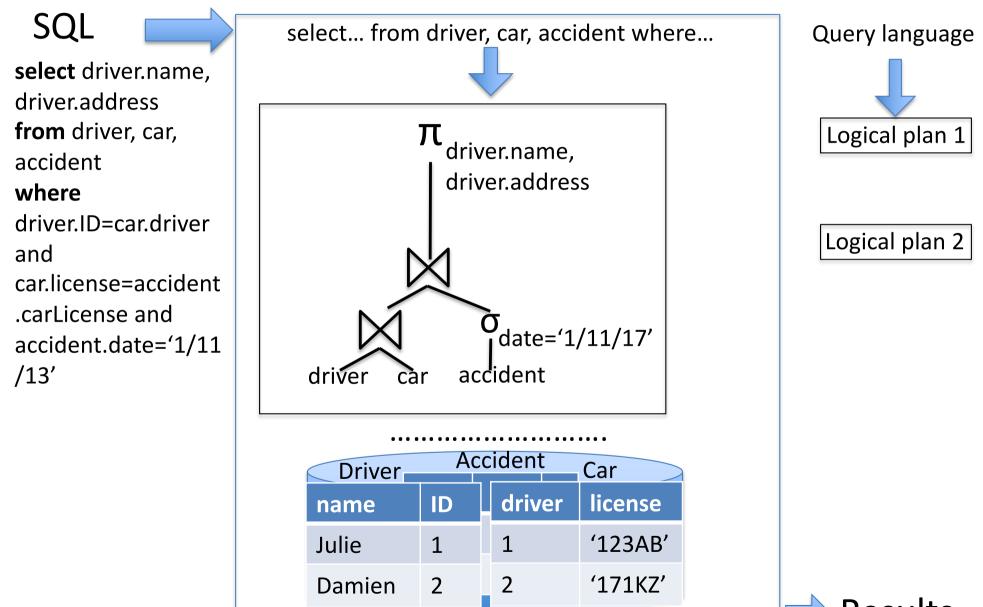
• Trees made of logical operators, each of which specializes in a certain task



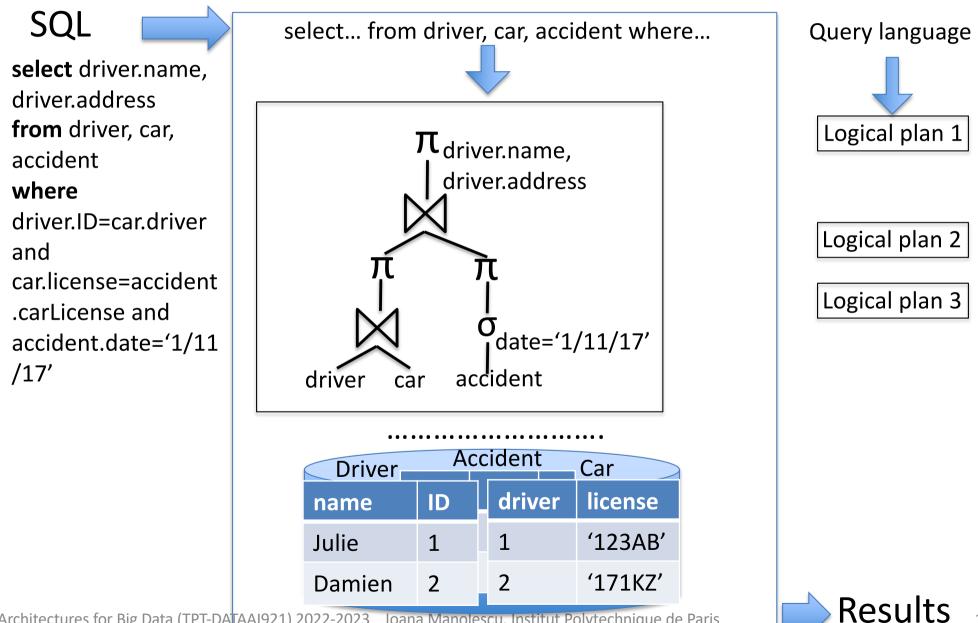
#### Logical query plans

- Trees made of logical operators, each of which specializes in a certain task
- Logical operators: they are defined by their result, not by an algorithm
- Physical operators (see next) implement actual algorithms

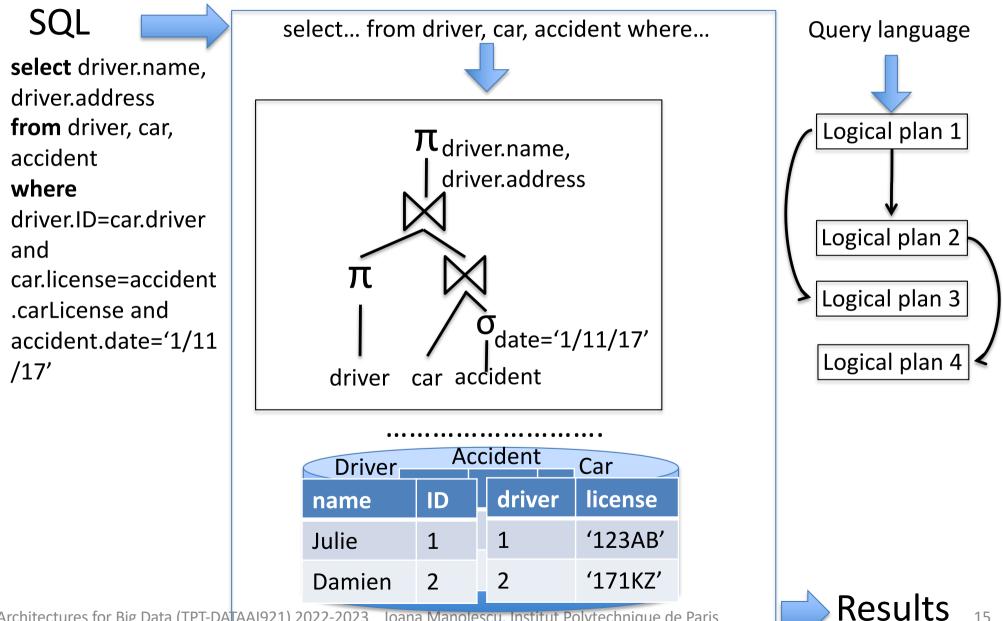












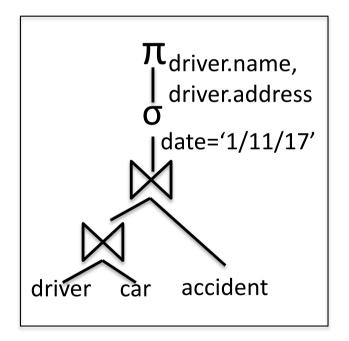


## Logical query optimization

- Enumerates logical plans
- All logical plans compute the query result
  - They are equivalent
- Some are (much) more **efficient** than others
- Logical optimization: moving from a plan to a more efficient one
  - Pushing selections
  - Pushing projections
  - Join reordering: most important source of optimizations

1.000.000 cars, 1.000.000 drivers, 1.000 accidents, 2 cars per accident, 10 accidents on 1/11/17

« Name and address of drivers in accidents on 1/11/2017? »

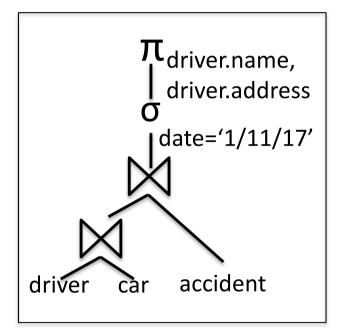


Cost of an operator: depends on the number of tuples (or tuple pairs) which it must process e.g. c\_disk x number of tuples read from disk e.g. c\_cpu x number of tuples compared Cardinality of an operator's output: how many tuples result from this operator

The cardinality of one operator's output determines the cost of its parent operator Plan **cost** = the sum of the costs of all operators in a plan

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« Name and address of drivers in accidents on 1/11/2017? »



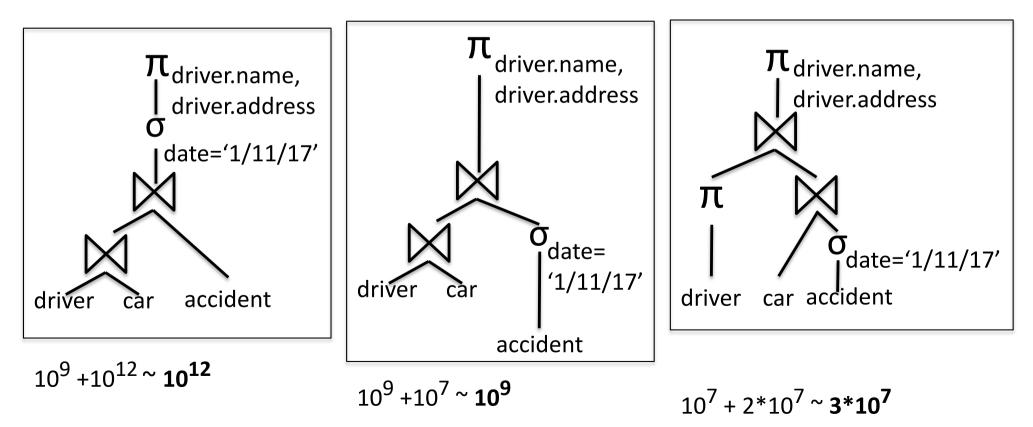
cs, cj, cf constant

Pessimistic (worst-Scan costs: cs x  $(10^6 + 10^6 + 10^3)$ case) estim. Scan cardinality estimations:  $10^6$ ,  $10^6$ ,  $10^3$ Driver-car join cost estimation: cj x  $(10^6 \times 10^6 = 10^{12})$ Driver-car join cardinality estimation: 10<sup>6</sup> Driver-car-accident join cost estim.: cj x  $(10^6 \times 10^3 = 10^9)$ Driver-car-accident join cardinality estimation:  $2 \times 10^3$ Selection cost estimation: cf x  $(2 \times 10^3)$ Selection cardinality estimation: 10 Projection (similar), negligible Total cost estimation: cs x  $(2x10^{6}+10^{3})$ + cf x 2x  $10^{3}$ + cj x  $(10^{12} + 2x10^3)$  ~ cj x  $10^{12}$  ~ **10**<sup>12</sup>

1.000.000 cars, 1.000.000 drivers, 1.000 accidents, 2 cars per accident, 10 accidents on 1/11/17

« Name and address of drivers in accidents on 1/11/2017? »

Three plans, same scan costs (neglected below); join costs dominant



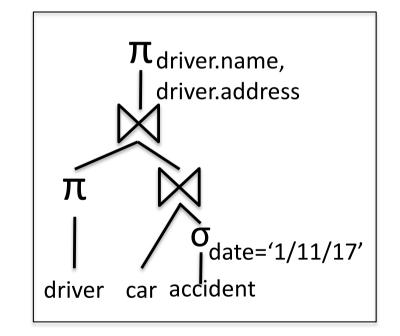
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Three plans, same scan costs (neglected below); join costs dominant

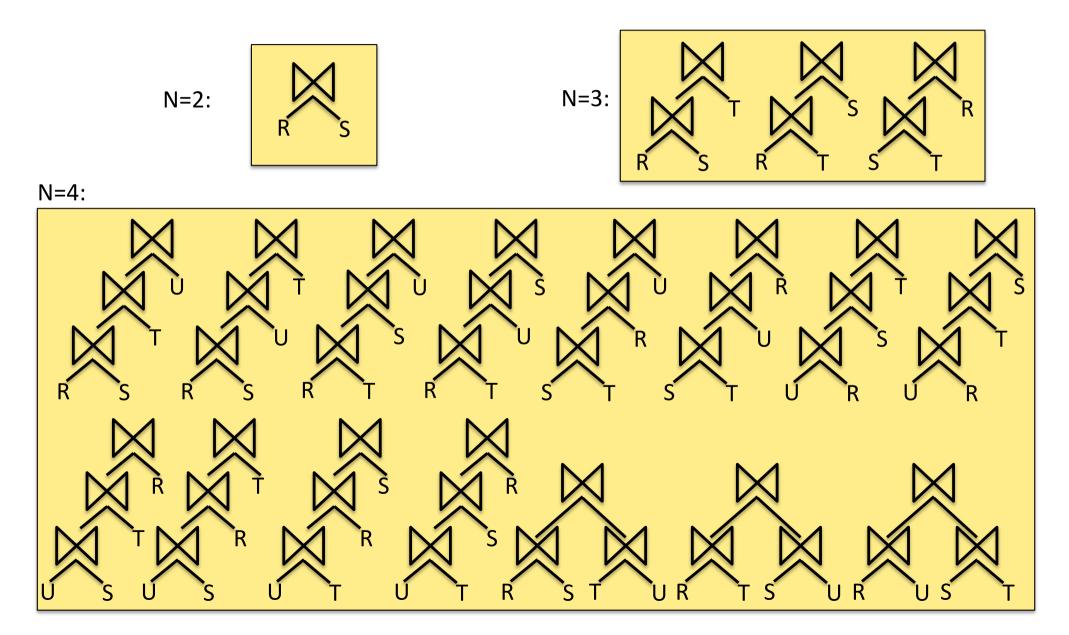
The best plan reads only the accidents that have to be consulted

- Selective data access
- Typically supported by an index
  - Auxiliary data structure, built on top of the data collection
  - Allows to access directly objects satisfying a certain condition



$$10^7 + 2*10^7 \sim 3*10^7$$

Join ordering is the main problem in logical query optimization



# Join ordering is the main problem in logical query optimization

Plans(n+1) = (n+1) \* Plans(n) +  $\frac{1}{2}$  \*  $\Sigma_{i=1}$  (n/2) Plans(i)\*Plans(n+1-i)

High (exponential) complexity  $\rightarrow$  many heuristics

• Exploring only left-linear plans etc.

N=4:

#### Logical query optimization needs statistics

**Exact** statistics (on base data):

- 1.000.000 cars, 1.000.000 drivers, 1.000 accidents
- **Approximate** / estimated statistics (on intermediary results)
  - "1.75 cars involved in every accident"

Statistics are gathered

- When **loading** the data: take advantage of the scan •
- **Periodically** or upon **request** (e.g. analyze in the Postgres RDBMS) •
- At **runtime**: modern systems may do this to change the data layout •

Statistics on the base data vs. on results of operations not evaluated (yet):

- « On average 2 cars per accident »
- For each column R.a, store: •

|R|, |R.a| (number of distinct values), min{R.a}, max{R.a}

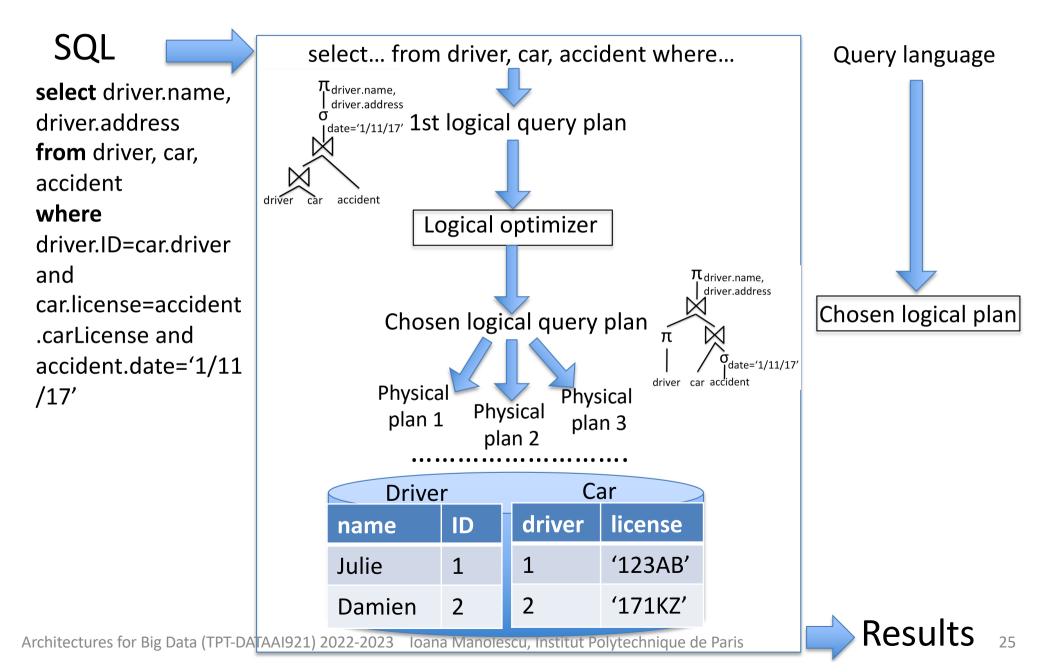
- Assume **uniform distribution** in R.a
- Assume independent distribution
  - of values in R.a vs values in R.b; of values in R.a vs values in S.c
- + simple probability computations

#### More on statistics

- For each column R.a, store: •
  - |R|, |R.a| (number of distinct values), min{R.a}, max{R.a}
- Assume **uniform distribution** in R.a.
- Assume independent distribution •
  - of values in R.a vs values in R.b; of values in R.a vs values in S.c

- The **uniform distribution** assumption is **frequently wrong** •
  - Real-world distribution are skewed (popular/frequent values)
- The **independent distribution** assumption is **sometimes wrong** •
  - « Total » counter-example: *functional dependency*
  - Partial but strong enough to ruin optimizer decisions: correlation
- Actual optimizers use more sophisticated statistic informations •
  - **Histograms**: equi-width, equi-depth
  - Trade-offs: size vs. maintenance cost vs. control over estimation error

# Database internal: query optimizer



#### Physical query plans

#### Made up of **physical operators** =

algorithms for implementing logical operators

Example: equi-join (R.a=S.b)

Nested loops join:

foreach t1 in R{
 foreach t2 in S {
 if t1.a = t2.b then output (t1 || t2)

Merge join: // requires sorted inputs
repeat{
 while (!aligned) { advance R or S };
 while (aligned) { copy R into topR, S into topS };
 output topR x topS;
} until (endOf(R) or endOf(S));

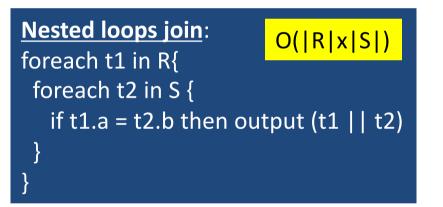
Hash join: // builds a hash table in memoryWhile (!endOf(R)) {  $t_R \leftarrow R.next$ ; put(hash( $t_R.a$ ),  $t_R$ ); }While (!endOf(S)) {  $t_S \leftarrow S.next$ ;<br/>matchingR = get(hash( $t_S.b$ ));<br/>output(matchingR x  $t_S$ );

#### Physical query plans

#### Made up of **physical operators** =

algorithms for implementing logical operators

Example: equi-join (R.a=S.b)



Merge join: // requires sorted inputs				
repeat{	O( R + S )			
while (!aligned) { advance R or S };				
while (aligned) { copy R into topR, S into topS };				
output topR x topS;				
<pre>} until (endOf(R) or endOf(S));</pre>				

Hash join: // builds a hash table in memoryAWhile (!endOf(R)) {  $t_R \leftarrow R.next; put(hash(t_R.a), t_R); }BIWhile (!endOf(S)) { <math>t_S \leftarrow S.next;$ InmatchingR = get(hash(t\_S.b));HO(|R|+|S|)Output(matchingR x t\_S);

Also:

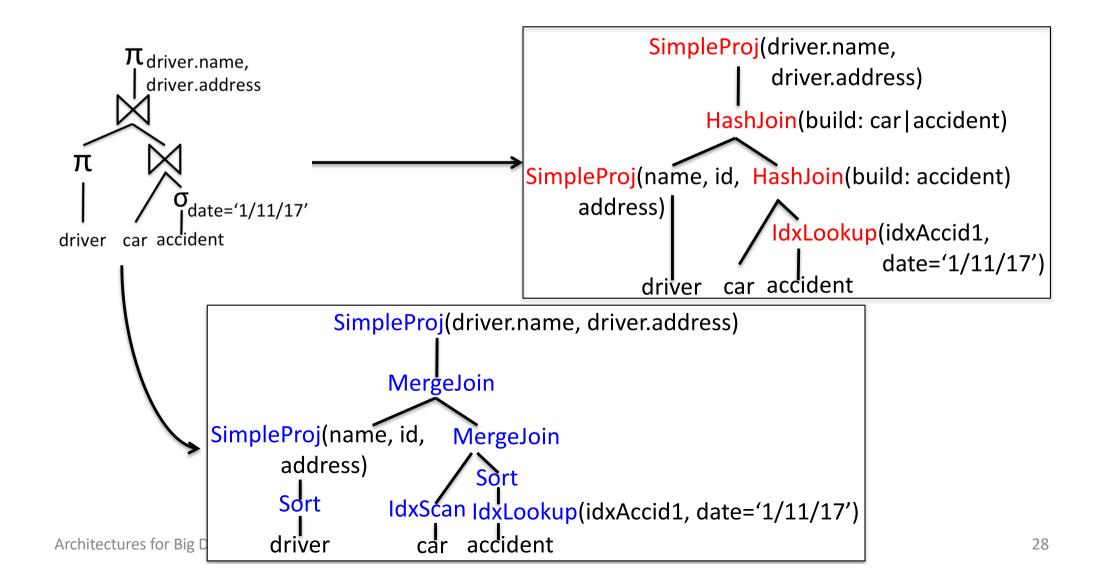
...

Block nested loops join Index nested loops join Hybrid hash join Hash groups / teams

Architectures for Big Dat

# Physical optimization

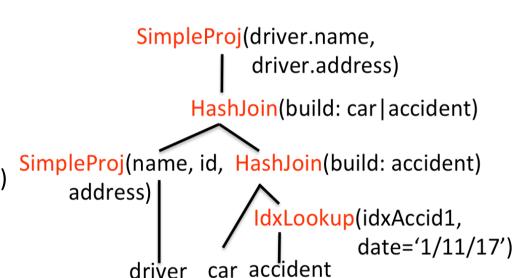
Possible physical plans produced by physical optimization for our sample logical plan:



# Physical plan performance

Metrics characterizing a physical plan

- **Response time**: between the time the query starts running to the we know it's end of results
- Work (resource consumption)
  - How many I/O calls (blocks read)
    - Scan, IdxScan, IdxAccess; Sort; HybridHash (or spilling HashJoin)
  - How much CPU
    - All operators
  - Distributed plans: network traffic
- Total work: work made by all operators



# Query optimizers in action

Most database management systems have an « explain » functionality  $\rightarrow$  physical plans. Below sample Postgres output:

EXPLAIN SELECT \* FROM tenk1;

QUERY PLAN

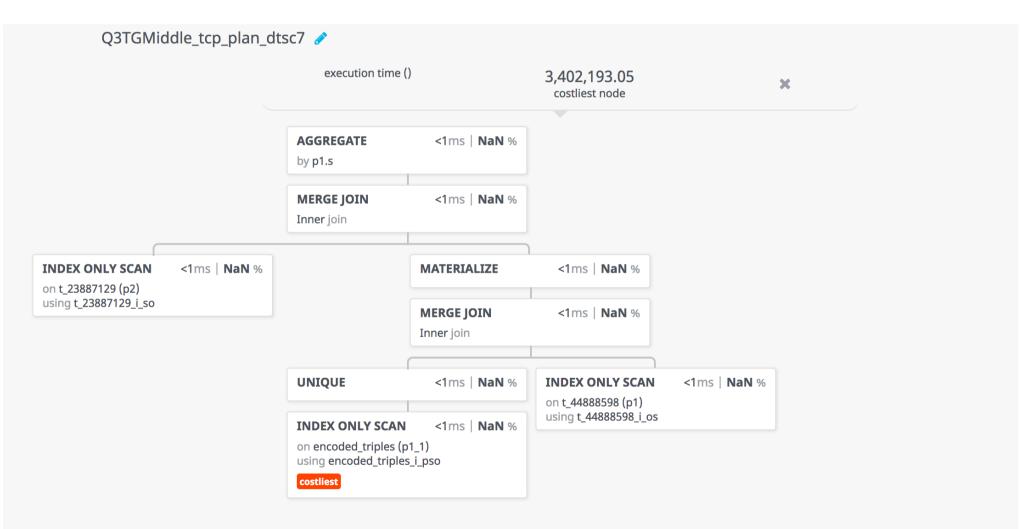
Seq Scan on tenk1 (cost=0.00..458.00 rows=10000 width=244)

```
EXPLAIN SELECT * FROM tenk1 t1, tenk2 t2
WHERE t1.unique1 < 100 AND t1.unique2 = t2.unique2;
QUERY PLAN
```

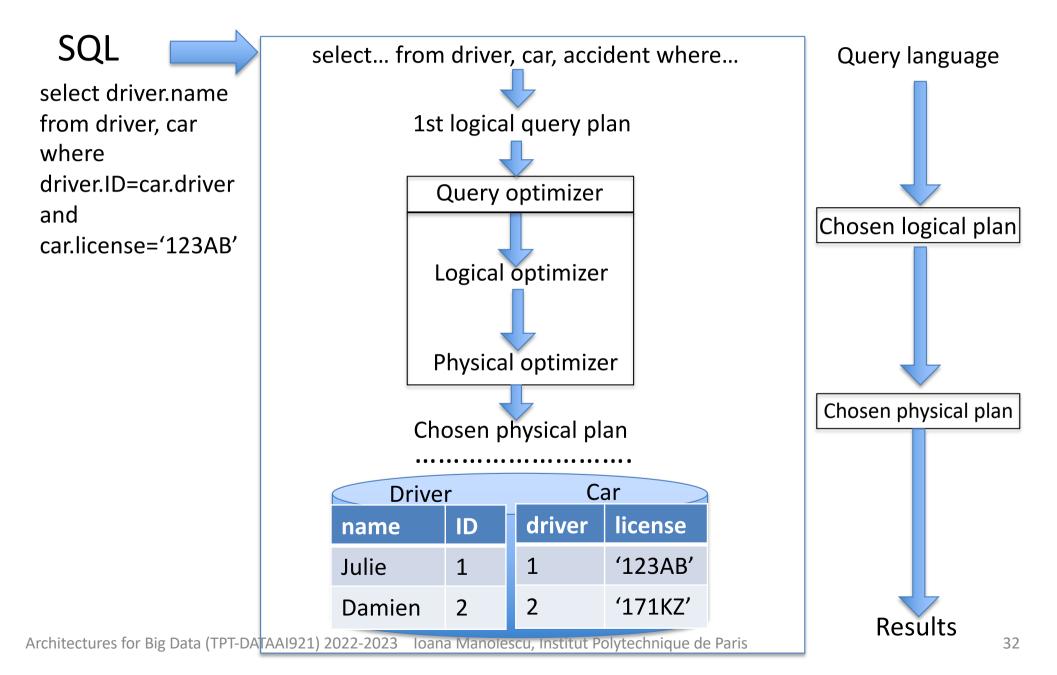
```
Hash Join (cost=232.61..741.67 rows=106 width=488)
Hash Cond: ("outer".unique2 = "inner".unique2)
-> Seq Scan on tenk2 t2 (cost=0.00..458.00 rows=10000 width=244)
-> Hash (cost=232.35..232.35 rows=106 width=244)
-> Bitmap Heap Scan on tenk1 t1 (cost=2.37..232.35 rows=106 width=244)
Recheck Cond: (unique1 < 100)
-> Bitmap Index Scan on tenk1_unique1 (cost=0.00..2.37 rows=106 width=0)
Index Cond: (unique1 < 100)</pre>
```

#### Inspecting query plans

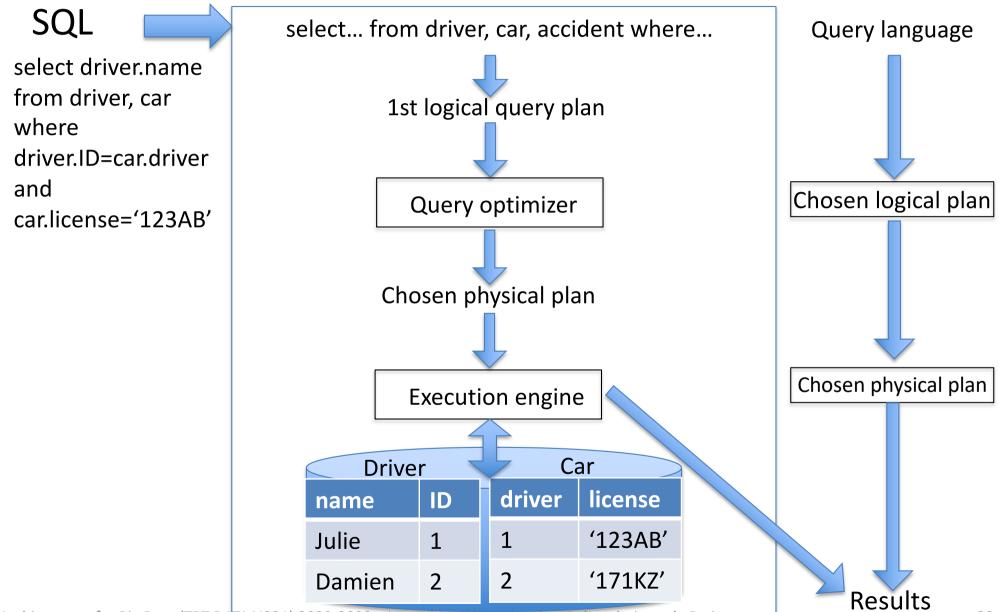
Here using <u>https://tatiyants.com/pev/#/plans</u>



## Database internal: physical plan

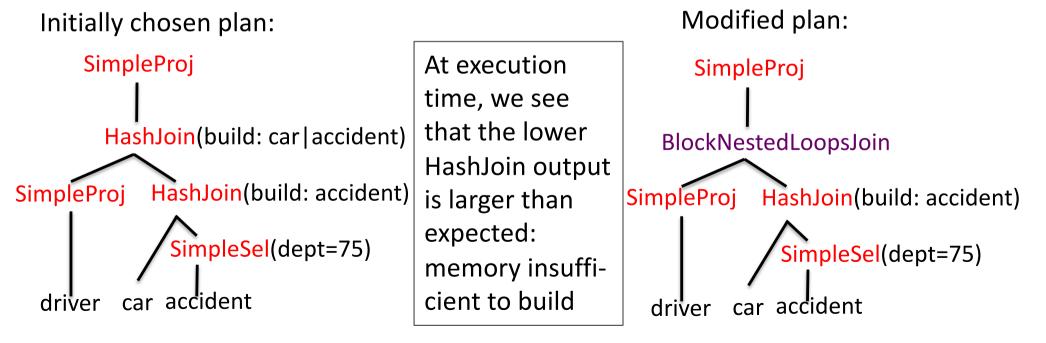


#### Database internals: query processing pipeline



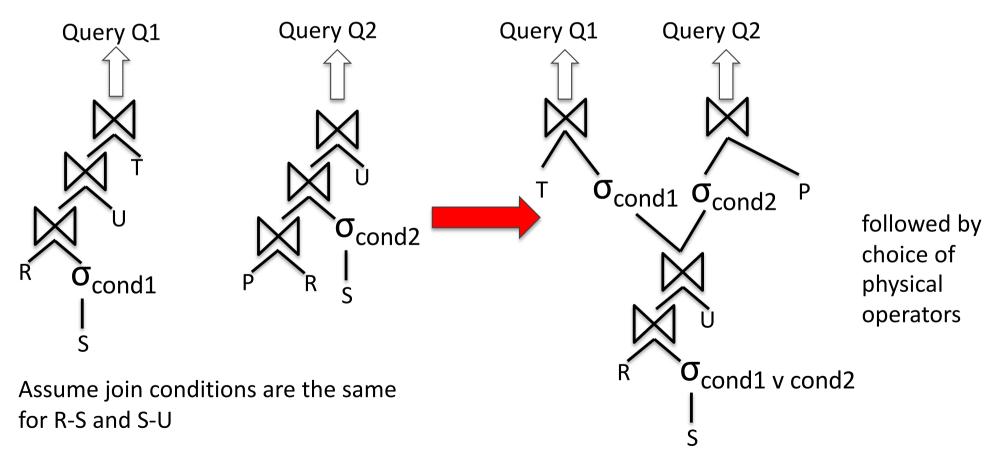
### Advanced query optimization techniques: Dynamic Query Optimization

- Sizes (cardinalities) of intermediary results are estimated, which may lead to estimation errors
- A cardinality estimation error may lead to chosing a logical plan and a set of physical operators that perform significantly different from expectation (especially for the worse)



#### Advanced query optimization techniques: Multi-Query Optimization

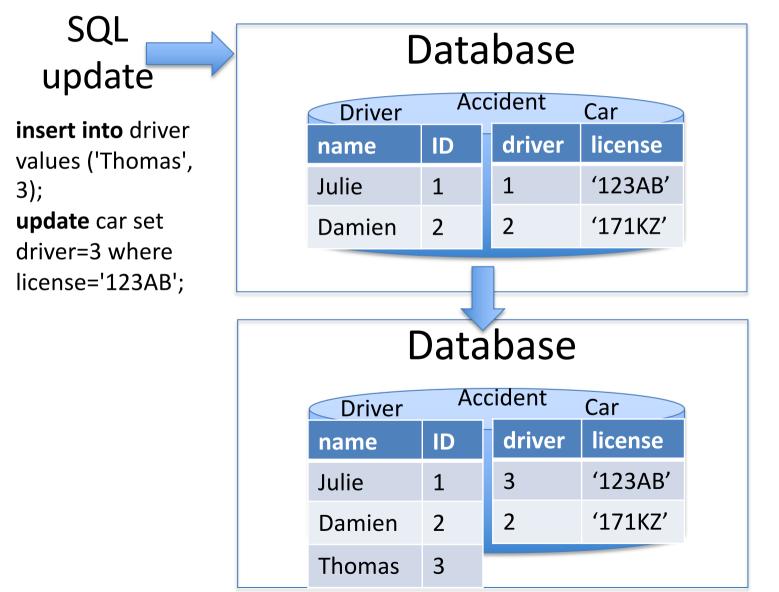
Multiple queries sharing sub-expressions can be optimized together into a single plan with **shared subexpressions** 



#### DATABASE FUNDAMENTALS (RECALL/CRASH COURSE)

**Updating the database** 

### What's in a database?



## Database updates

- A set of operations atomically executed (either all, or none) is called a **transaction**
- There may be some **dependencies** between the operations of a transaction
  - First read the bank account balance
  - Then write that value reduced by 50€
- A total order over the operations of several concurrent transaction is called a **scheduling**
- The DB component that receives all incoming transactions and decides what operation will be executed when (i.e., <u>global order</u> over the operations of all transactions) is the **scheduler**

#### Database updates

• The **scheduler** is in charge of ordering all operations so that they will appear executed one after the other (serially)

T1: BEGIN A=A+100, B=B-100 END T2: BEGIN A=1.06\*A, B=1.06\*B END

T1: A=A+100, B=B-100, T2: A=1.06\*A, B=1.06\*B

T1: A=A+1	00,	B=B-100	
T2:	A=1.06*A,	B=1.0	)6*B

# Ensuring consistency of concurrent updates

- **Scheduling** is implemented through specific algorithms and with the help of protocols
- A **protocol** is a rule that holds on the order in which a transaction performs its operations
  - E.g., "once a trasaction has released a lock, the transaction will never take another lock"
- <u>If</u> all transactions follow a given protocol, <u>then</u>, regardless on the order in which they are executed, certain **good properties** are guaranteed
  - E.g., "there is no deadlock" or "the result is the same as if the transactions had been executed one after the other"

#### Fundamental database features

#### 1. Data storage

- Protection against unauthorized access, data loss
- 2. Ability to at least **add** to and **remove** data to the database
  - Also: updates; active behavior upon update (triggers)
- 3. Support for **accessing** the data
  - Declarative query languages: say what data you need, not how to find it

# Fundamental properties of database stores: ACID

- Atomicity: either all operations involved in a transactions are done, or none of them is
  - E.g. bank payment
- Consistency: application-dependent constraint
   E.g. every client has a single birthdate
- **Isolation**: concurrent operations on the database are executed as if each ran alone on the system
  - E.g. if a debit and a credit operation run concurrently, the final result is still correct
- **Durability**: data will not be lost nor corrupted even in the presence of system failure during operation execution

Jim Gray, ACM Turing Award 1998 for « fundamental contributions to databases and transaction management »

## **ACID** properties

- **Atomicity**: per transaction (cf. boundaries)
- **Consistency**: difference in the expressive power of the constraints
- Illustrated below for relational databases, create table statement:

```
CREATE TABLE tbl_name (create_definition,...) [table_options] [partition_options]
```

CHECK (expr)

*column\_definition: data\_type* [**NOT NULL** | **NULL**] [DEFAULT *default\_value*] [AUTO\_INCREMENT] [UNIQUE [KEY] | [PRIMARY] KEY] (...)

# **ACID** properties

Consistency (continued)

• <u>SQL</u> constraint syntax (within create table):

[CONSTRAINT [symbol]] FOREIGN KEY [index\_name]

(index\_col\_name, ...)

**REFERENCES** *tbl\_name* (*index\_col\_name*,...)

[**ON DELETE** *reference\_option*]

[**ON UPDATE** *reference\_option*]

reference\_option: RESTRICT | CASCADE | SET NULL | NO ACTION

- Key-value store: <u>REDIS</u>
  - a data item can have only one value for a given property
- Key-value store: <u>DynamoDB</u>
  - The value of a data item can be constrained to be unique, or allowed to be a set
- Hadoop File System (<u>HDFS</u>): no constraints

## ACID properties

- **Isolation**: concurrent operations on the database are executed as if each ran alone on the system
  - Watch out for: read-write (RW) or write-write (WW) conflicts
  - Conflict granularity depends on the data model
- An example of advanced isolation support: SQL
  - E.g. SQL

Isolation Level	Dirty Read	Non Repeatable Read	Phantom
Read uncommitted	Yes	Yes	Yes
Read committed	No	Yes	Yes
Repeatable read	No	Νο	Yes
Snapshot	No	Νο	No
Serializable	No	Νο	No

- High isolation conflicts with high transaction throughput
- E.g. HDFS: a file is never modified (written only once and integrally)

#### DATABASE FUNDAMENTALS (RECALL/CRASH COURSE)

**Takeaway** 

Main principles behind correct and scalable data management...

... pioneered in database management systems:

- **1. Declarative query language** allows users to just state what they want
- 2. For one query there are several **logical plans**; for each, several **physical plans** 
  - Optimizer picks **best plan**
- **3.** ACID properties crucial for "faith in the system" ("my salary, payments, and social security are within a reliable system")