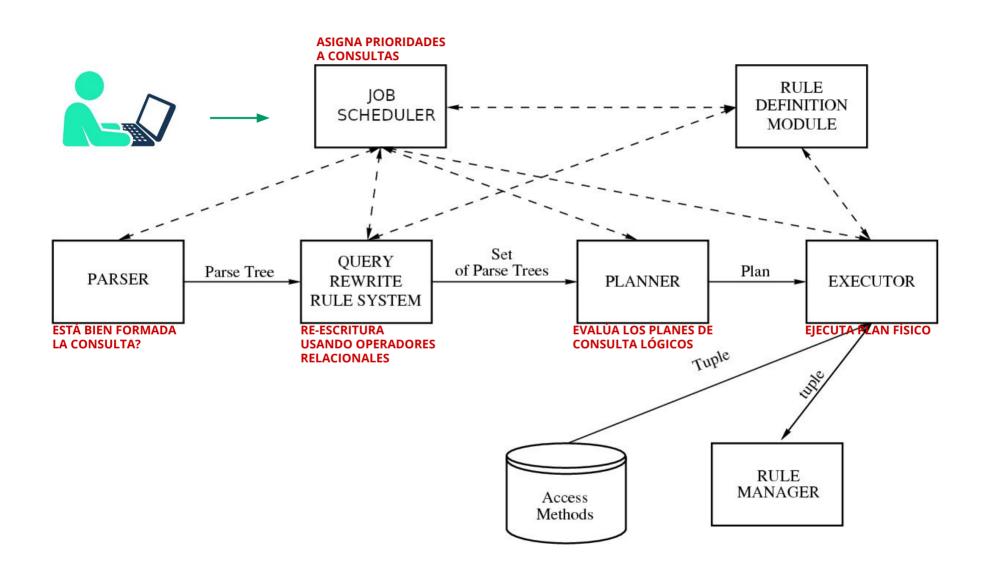
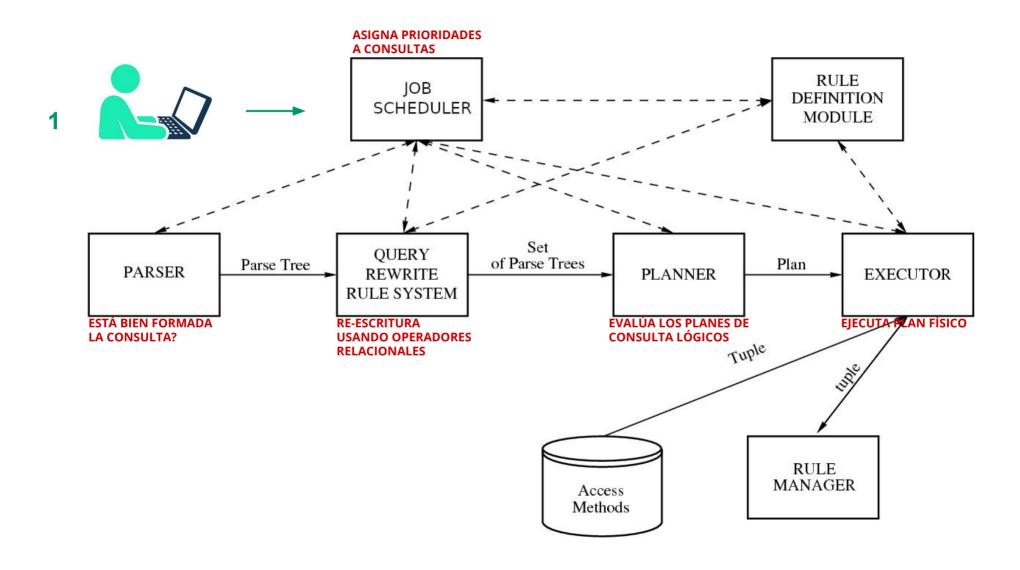
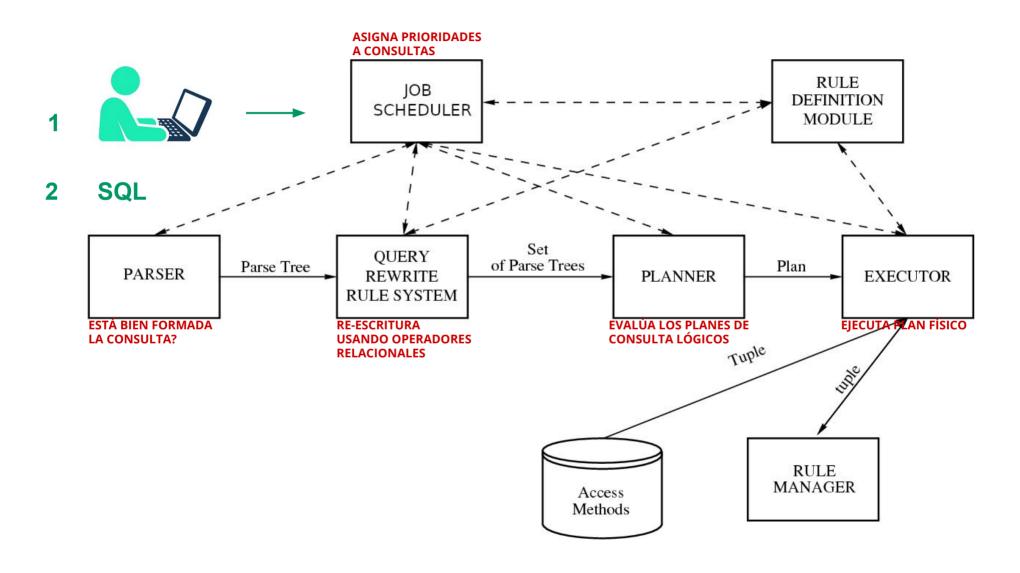
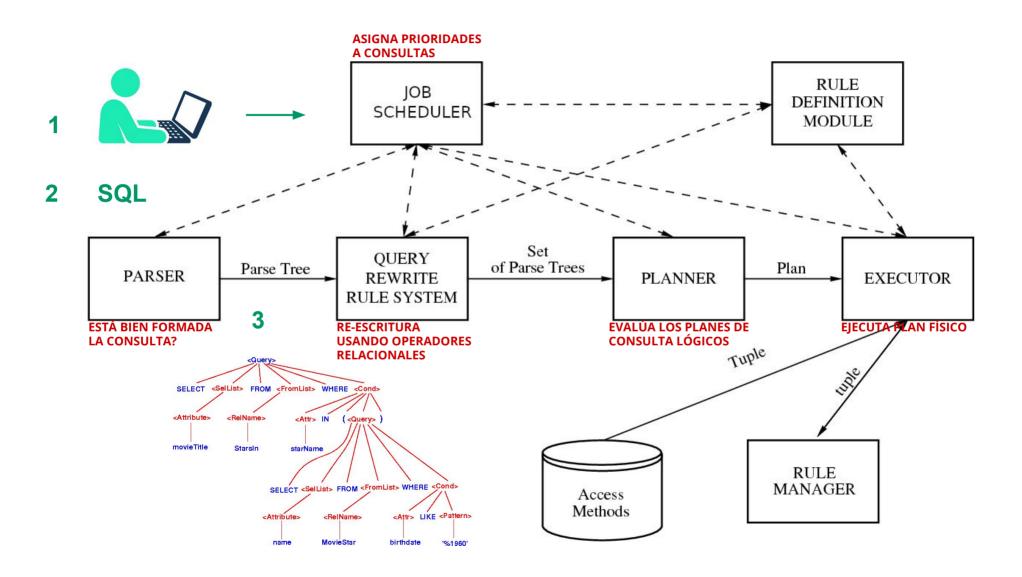
Bases de Datos

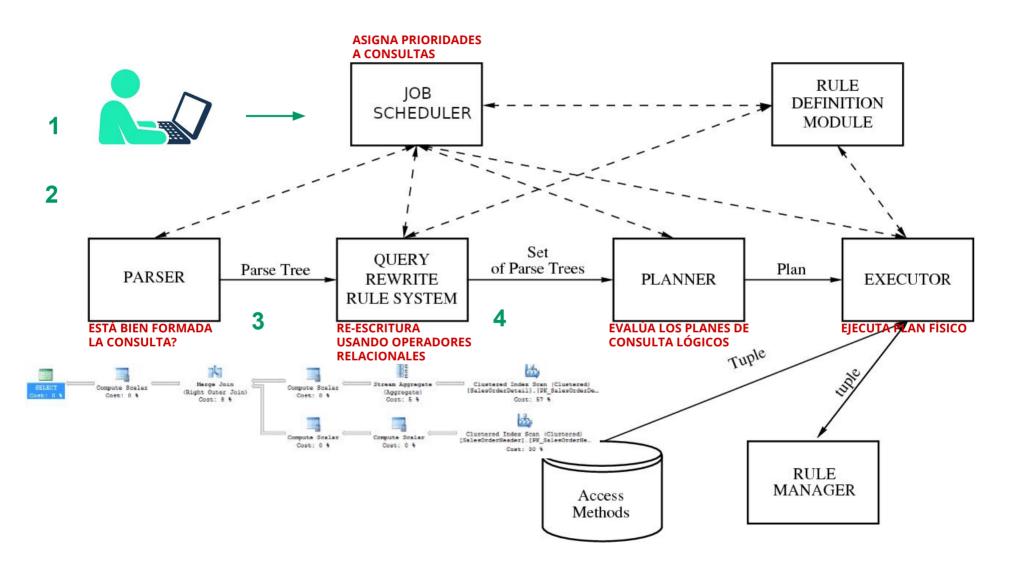
Ejecución de consultas

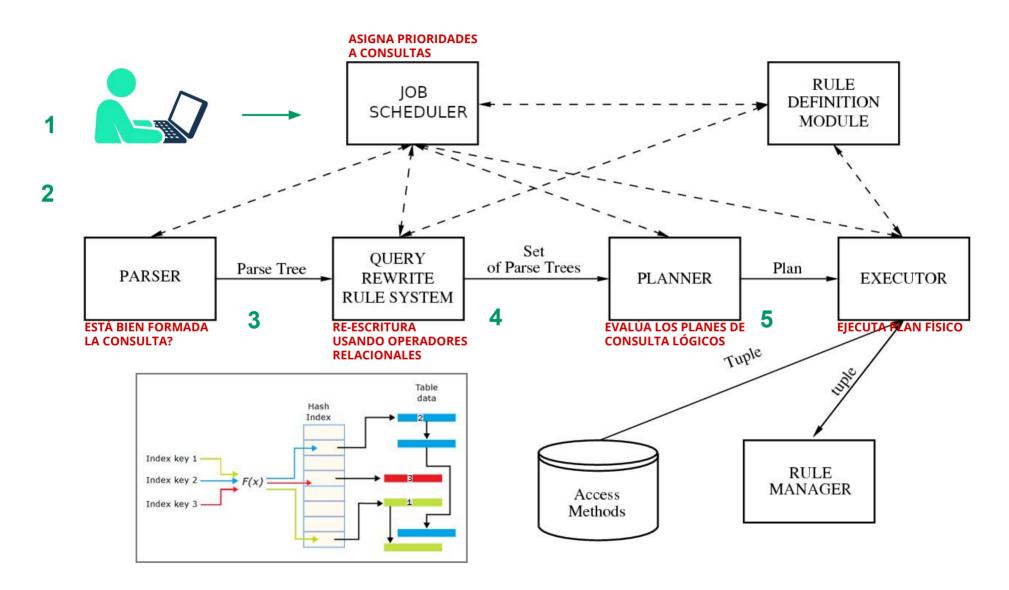


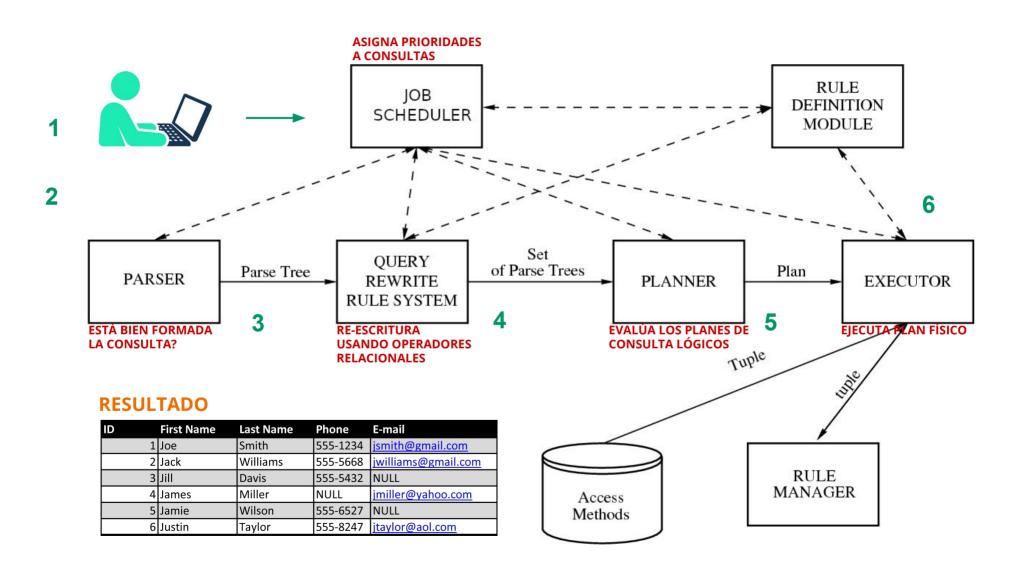




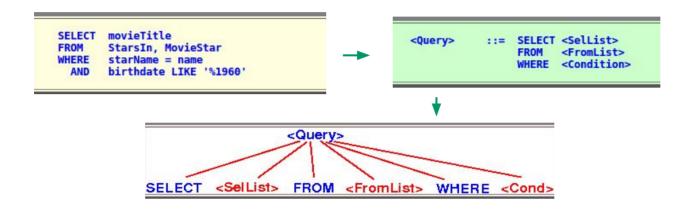


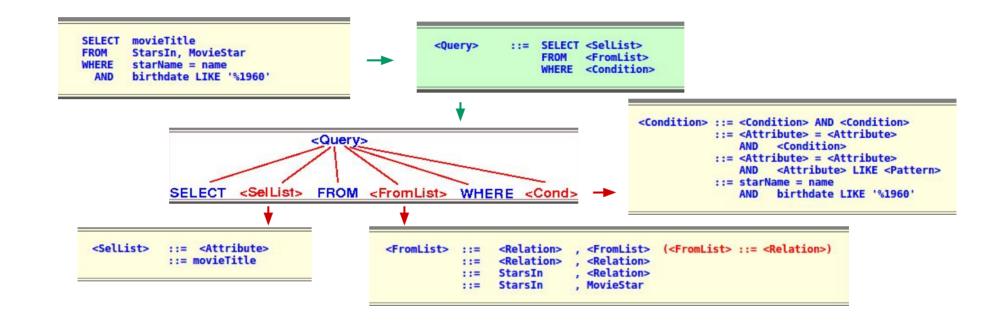


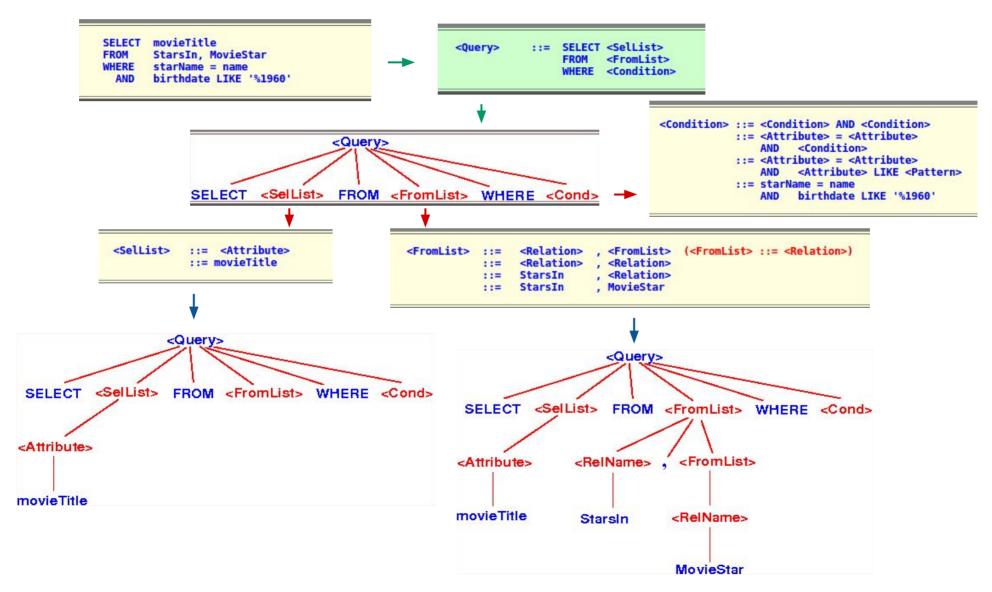


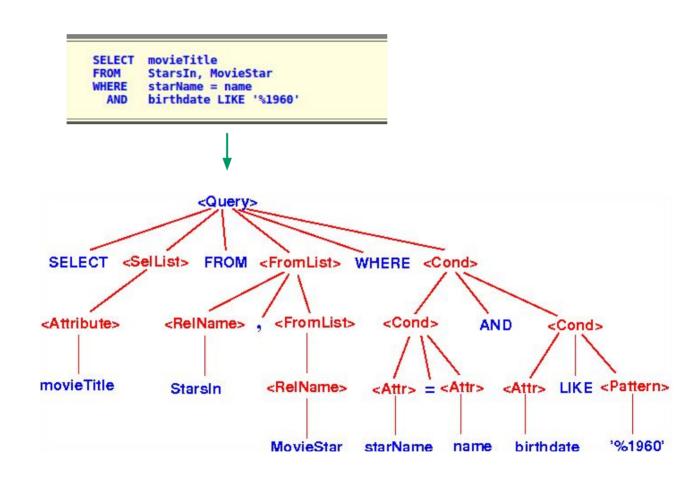


Gramática SQL

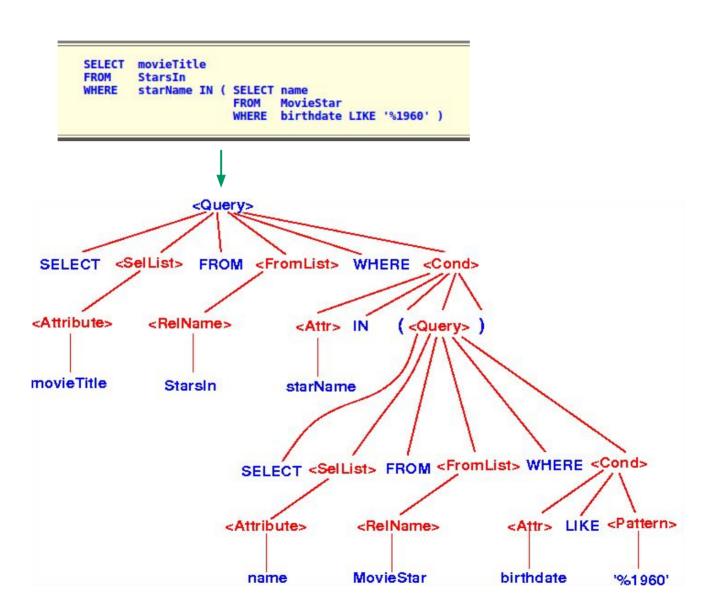








```
SELECT movieTitle
FROM StarsIn
WHERE starName IN ( SELECT name
FROM MovieStar
WHERE birthdate LIKE '%1960' )
```

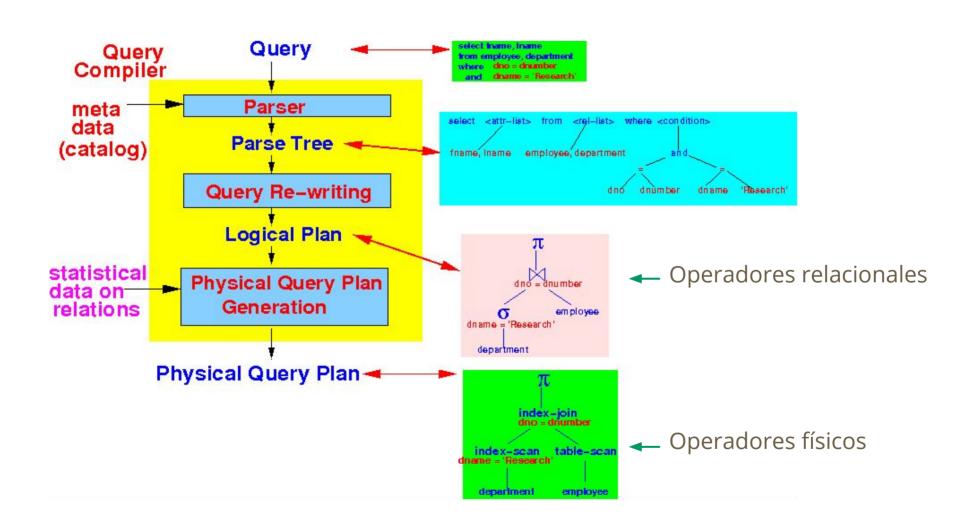


Los planes de ejecución lógicos se expresan usando operadores relacionales

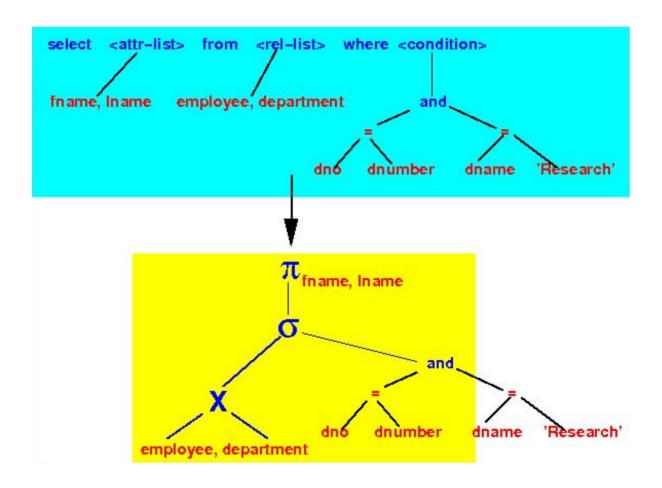
Symbol	Name of operator	Operation	
table- scan(R)	Table scan	Reads blocks of the input relation R in - one block at a time.	
index- scan(R)	Index scan	Reads blocks of the input relation R in using some index on the relation R	
$\sigma_{\rm cond}$	Selection	Selects the tuples that satisfies the condition cond	
Π _{attrs}	Projection	Extracts the attributes in attrs from a tuple Note: result of π is a bag (possible duplicates !!!)	
\bowtie_{cond}	Theta Join	Joins 2 relations on the condition cond	
M	Natural Join	Equi-joins 2 relations on the attributes than have the same name	
×	Product	Computes the (cartesian) product of 2 relations	
δ	Duplicate Elmination	Removes the duplicate tuples from a bag of tuples (the result is a set)	
$\gamma_{\rm L}$	Grouping	Form gorups of tuples on common attribute values and apply the function L on each group	
Us	Set Union	Computes the set union of 2 sets (no duplicates)	
UB	Bag Union	Computes the bag union of 2 sets (allows duplicates)	
n _s	Set Intersection	Computes the set intersection of 2 sets (no duplicates)	
∩ _B	Bag Intersection	Computes the bag intersection of 2 sets (allows duplicates)	
-s	Set Difference	Computes the set difference of 2 sets (no duplicates)	
- _в	Bag Difference	Computes the bag difference of 2 sets (allows duplicates)	

Los planes de ejecución lógicos se expresan usando operadores relacionales

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—	⋈cond	Theta Join	Joins 2 relations on the condition cond
—	M	Natural Join	Equi-joins 2 relations on the attributes than have the same name
CROSS JOIN	×	Product	Computes the (cartesian) product of 2 relations
	δ	Duplicate Elmination	Removes the duplicate tuples from a bag of tuples (the result is a set)
GROUP BY	$\gamma_{\rm L}$	Grouping	Form gorups of tuples on common attribute values and apply the function L on each group
	Us	Set Union	Computes the set union of 2 sets (no duplicates)
	UB	Bag Union	Computes the bag union of 2 sets (allows duplicates)
	ns	Set Intersection	Computes the set intersection of 2 sets (no duplicates)
	∩ _B	Bag Intersection	Computes the bag intersection of 2 sets (allows duplicates)
	-s	Set Difference	Computes the set difference of 2 sets (no duplicates)
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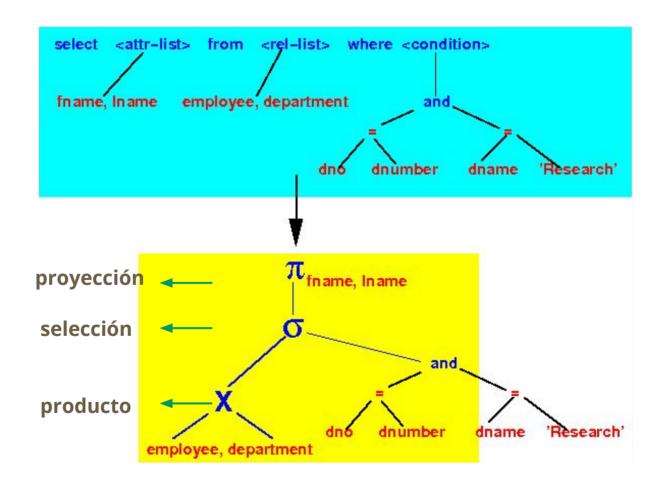


Parse-tree → plan lógico de la consulta



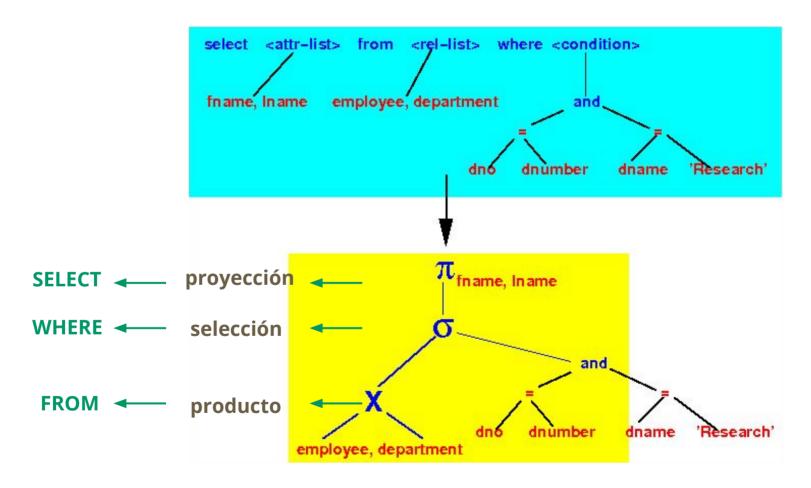
Partir desde el nivel más bajo hacia los niveles altos

Parse-tree → plan lógico de la consulta



Partir desde el nivel más bajo hacia los niveles altos

Parse-tree → plan lógico de la consulta



Partir desde el nivel más bajo hacia los niveles altos

Query Planner

Plan lógico de la consulta → plan optimizado

Criterios de optimización:

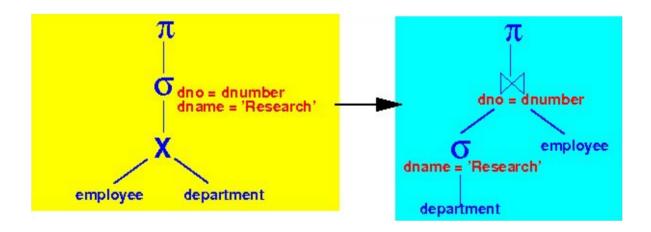
- 1. **Selection pushdown:** empujar la selección hacia las hojas
- 2. Join reordering: permutar las relaciones en un join

Query Planner

Plan lógico de la consulta → plan optimizado

Criterios de optimización:

- 1. Selection pushdown: empujar la selección hacia las hojas
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Query Planner

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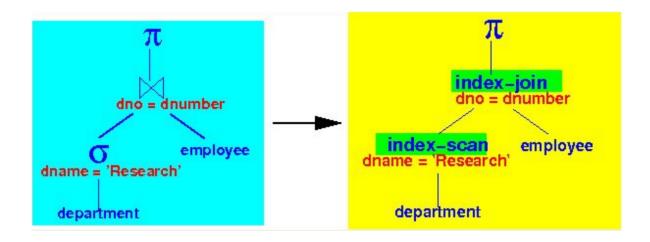


Executor

Selecciona el mejor algoritmo para ejecutar el plan lógico optimizado

El mejor algoritmo depende de:

- Índices disponibles
- Memoria disponible para procesar la consulta

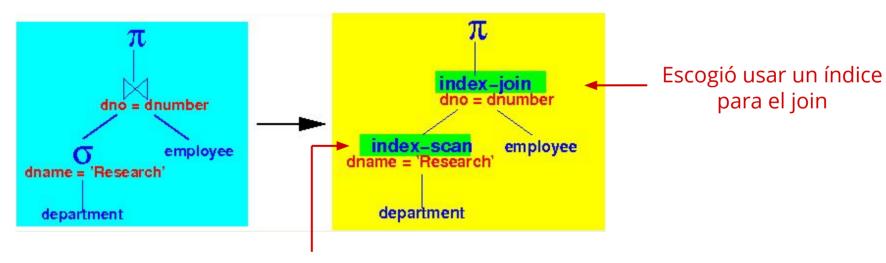


Executor

Selecciona el mejor algoritmo para ejecutar el plan lógico optimizado

El mejor algoritmo depende de:

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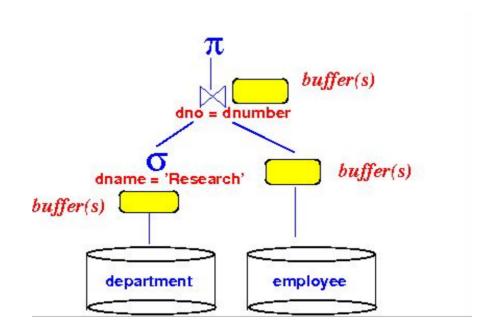
Escogió usar un índice para la selección

Costo de una consulta

¿Qué variables afectan al costo de ejecución?

Costo de una consulta

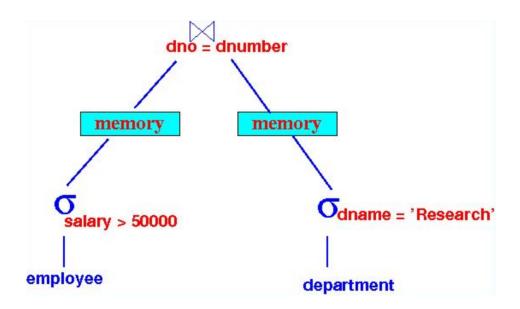
Una consulta requiere de *buffers* de memoria para poder computarse.



- La ejecución del plan físico requiere de varios *buffers* para alojar bloques de discos
- El costo de la consulta está determinado por el número de bloques de disco accesados

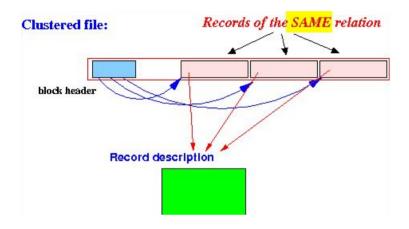
Costo de una consulta

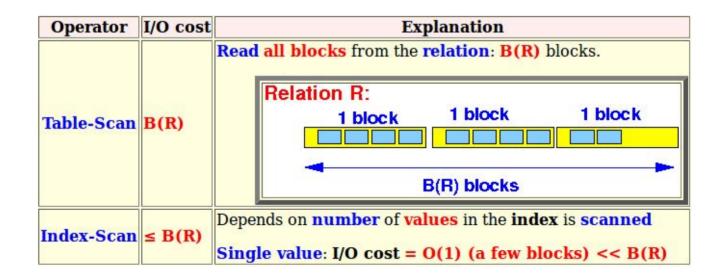
Asumimos que los resultados intermedios pueden serializarse en un pipeline, es decir, el operador que recibe los datos los lee desde un buffer:



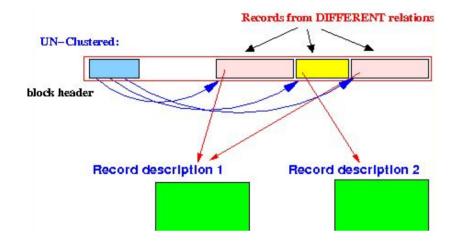
- Los costos pueden contabilizarse como el número de buffers necesarios para transferir los resultados en el plan de la consulta
- Contabilizamos los costos en buffers de entrada y en memoria

Archivos clusterizados

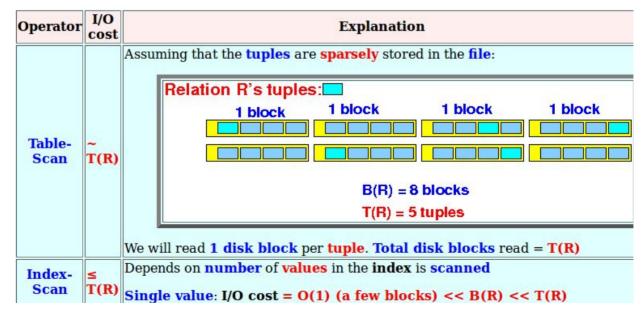




Archivos no clusterizados



Record header contains information of the relation to which the record belong



Operador de selección

Ejemplo:

Algoritmo de una pasada (*one-pass*)

```
Gcond ( R )

R = { ('John', 30000), ('Jane', 50000) }

Gsalary > 40000 ( employee ) = { ('Jane', 50000) }

while ( R has more data blocks )
{
  read data blocks to fill available buffers;
  check condition for each tuple in the buffers;
  Move qualifying tuples to output
}
```

Operador de selección

Ejemplo:

Algoritmo de una pasada (*one-pass*)

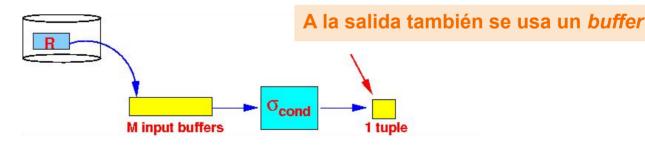
Uso del buffer

```
Gcond (R)

R = { ('John', 30000), ('Jane', 50000) }

Gsalary > 40000 ( employee ) = { ('Jane', 50000) }

while (R has more data blocks )
{
  read data blocks to fill available buffers;
  check condition for each tuple in the buffers;
  Move qualifying tuples to output
}
```



Costos

B(R) --- if the relation R is clustered
 T(R) --- if the relation R is UNclustered

Operador de proyección

Ejemplo:

Algoritmo de una pasada (*one-pass*)

```
Tattrs ( R )

R = { ('John', 30000), ('Jane', 50000) }

Tfiname ( employee ) = { ('John'), ('Jane') }

while ( R has more data blocks )
{
  read data blocks to fill available buffers;
  Extract the projection attrs for each tuple in buffers;
```

Move projected tuples to output

Operador de proyección

Ejemplo:

Algoritmo de una pasada (*one-pass*)

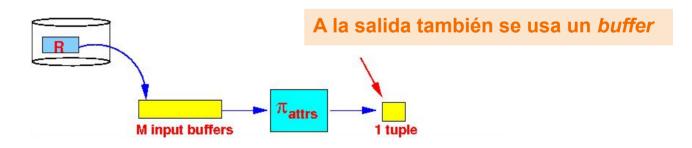
Uso del buffer

```
π<sub>attrs</sub> ( R )

R = { ('John', 30000), ('Jane', 50000) }

π<sub>fname</sub> ( employee ) = { ('John'), ('Jane') }
```

```
while ( R has more data blocks )
{
   read data blocks to fill available buffers;
   Extract the projection attrs for each tuple in buffers;
   Move projected tuples to output
}
```



Costos

B(R) --- if the relation R is clustered
 T(R) --- if the relation R is UNclustered

Operador de agrupamiento:

YL (R)

Ejemplo:

```
R = (x,y)
R = \{ (a,2), (a,3), (b,1), (b,3) \}
Yavg(y) (R) = \{ (a, 2.5), (b, 2.0) \}
```

Algoritmo de una pasada (one-pass):

```
initialize a search structure H on grouping attributes of y;
   Process the statistics for each group
while ( R has more data blocks )
   read 1 data block in buffer b;
   for ( each tuple t ∈ b )
        We need a search structure H to implement the test
        t ∈ H efficiently !!!
        We can use hash table or some bin, search tree
      if (t \in H)
        Update the statistics for group(t);
      else
        insert t in H;
        Initialize the statistics for group(t);
   Now we can output the aggregate function for each group
for ( each group ∈ H )
   Output group search key + statistics;
```

- 1 *buffer* para leer las tuplas
- M 1 *buffers* para almacenar el índice

Algoritmo de una pasada (one-pass):

```
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     else
        insert t in H;
        Initialize the statistics for group(t);
  Now we can output the aggregate function for each group
for ( each group ∈ H )
  Output group search key + statistics;
```

- 1 *buffer* para leer las tuplas
- M 1 *buffers* para almacenar el índice

Costo:

```
    B(R) -- if the relation R is clustered
    T(R) -- if the relation R is UNclustered
```

Memoria:

```
M = O( V(R, [A<sub>1</sub>, A<sub>2</sub>, ..., A<sub>k</sub>]) )

donde:

■ V(R, [A<sub>1</sub>, A<sub>2</sub>, ..., A<sub>k</sub>]) = #valores que [A<sub>1</sub>,...,A<sub>k</sub>] puede tomar
```

Consultas?

Recuerden!

- Marcelo Mendoza: <u>mmendoza@inf.utfsm.cl</u>
- Margarita Bugueño: <u>margarita.bugueno@usm.cl</u>

Bases de Datos

Ejecución de consultas