Query Processing:

Query Plan Operators

Parke Godfrey

Query Processing

Q → Query Plan

Query Processing

Q → Query Plan

Focus: Relational System

- Others?

Example

Select B,D
From R,S
Where R.A = “c” $\wedge$ S.E = 2 $\wedge$ R.C=S.C

R

A	B	C
a	1	10
b	1	20
c	2	10
d	2	35
e	3	45

S

C	D	E
10	x	2
20	y	2
30	z	2
40	x	1
50	y	3

R

A	B	C
a	1	10
b	1	20
c	2	10
d	2	35
e	3	45

S

C	D	E
10	x	2
20	y	2
30	z	2
40	x	1
50	y	3

Answer

B	D
2	x

- How do we execute query?

One idea

- Do Cartesian product
- Select tuples
- Do projection

RXS

R.A	R.B	R.C	S.C	S.D	S.E
a	1	10	10	x	2
a	1	10	20	y	2
.					
.					
C	2	10	10	x	2
.					
.					

RXS

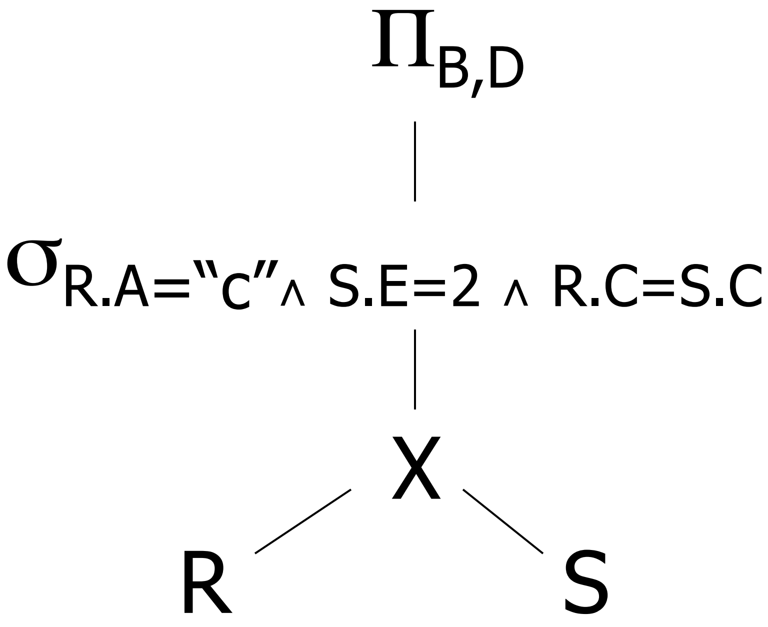
R.A	R.B	R.C	S.C	S.D	S.E
a	1	10	10	x	2
a	1	10	20	y	2
.					
.					
C	2	10	10	x	2
.					
.					

Bingo!

Got one...

Relational Algebra - can be used to describe plans...

Ex: Plan I



$\Pi_{B,D}$

$\sigma_{R.A=\text{“c”} \wedge S.E=2 \wedge R.C=S.C}$

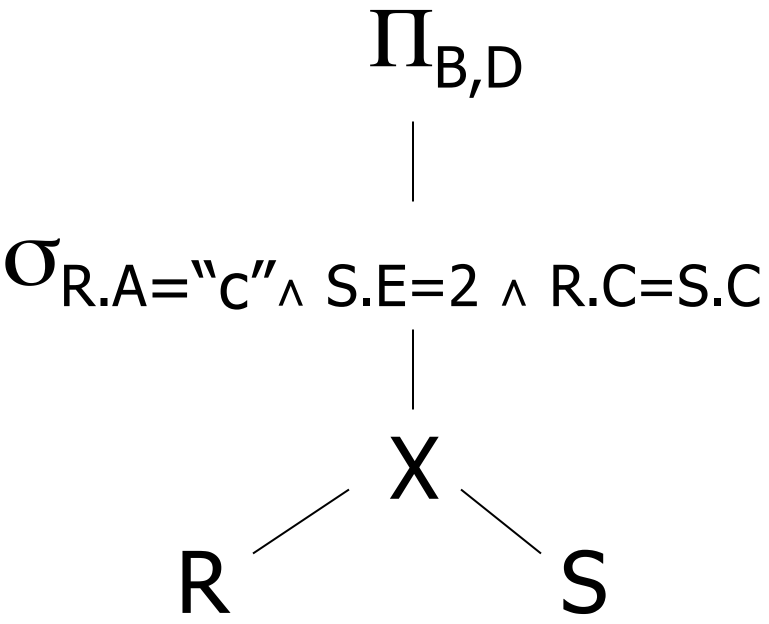
X

R

S

Relational Algebra - can be used to describe plans...

Ex: Plan I



$\Pi_{B,D}$

$\sigma_{R.A=\text{“c”} \wedge S.E=2 \wedge R.C=S.C}$

X

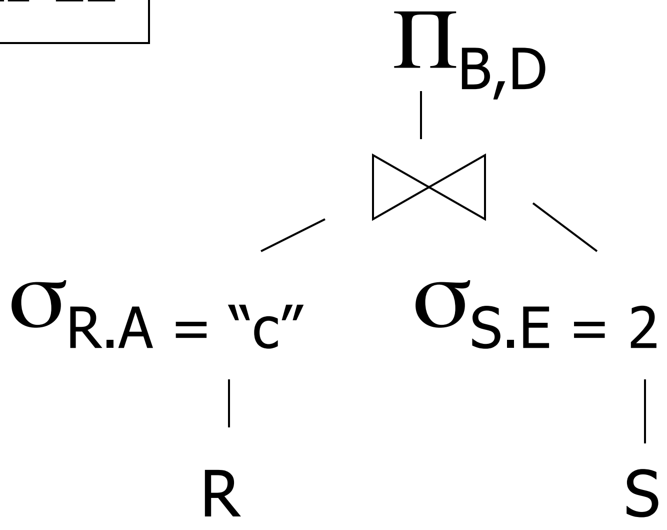
R

S

OR: $\Pi_{B,D}\left[\sigma_{R.A=\text{“c”} \wedge S.E=2 \wedge R.C = S.C}(RXS)\right]$

Another idea:

Plan II



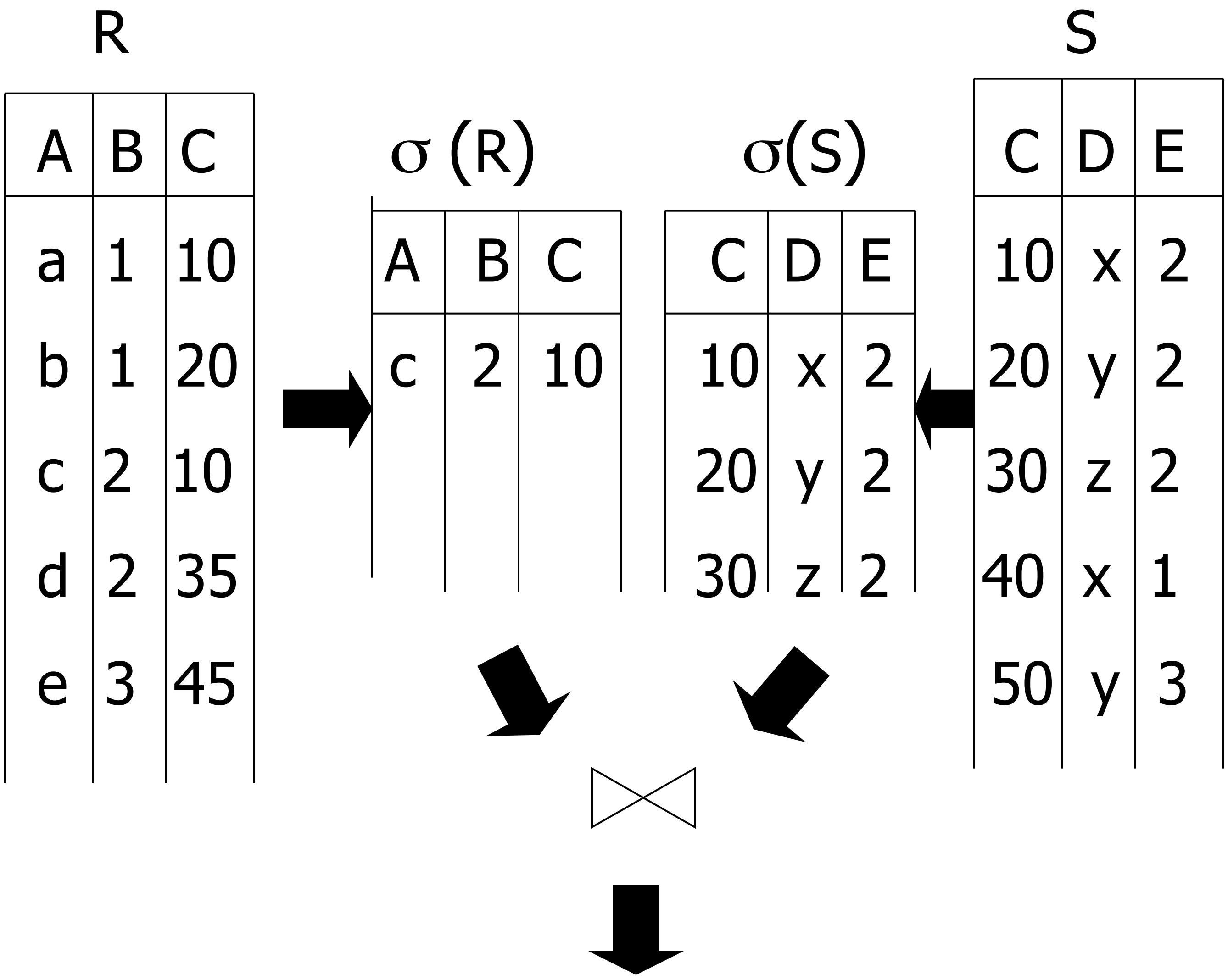
$\Pi_{B,D}$

$\bowtie$

$\sigma_{R.A = “c”}$ $\sigma_{S.E = 2}$

R S

$\bowtie$ natural join



R

A	B	C
a	1	10
b	1	20
c	2	10
d	2	35
e	3	45

σ (R)

A	B	C
c	2	10

σ(S)

C	D	E
10	x	2
20	y	2
30	z	2

S

C	D	E
10	x	2
20	y	2
30	z	2
40	x	1
50	y	3

$\bowtie$

Plan III

Use R.A and S.C Indexes

(1) Use R.A index to select R tuples with R.A = “c”

(2) For each R.C value found, use S.C index to find matching tuples

Plan III

Use R.A and S.C Indexes

(1) Use R.A index to select R tuples with R.A = “c”

(2) For each R.C value found, use S.C index to find matching tuples

(3) Eliminate S tuples S.E ≠ 2

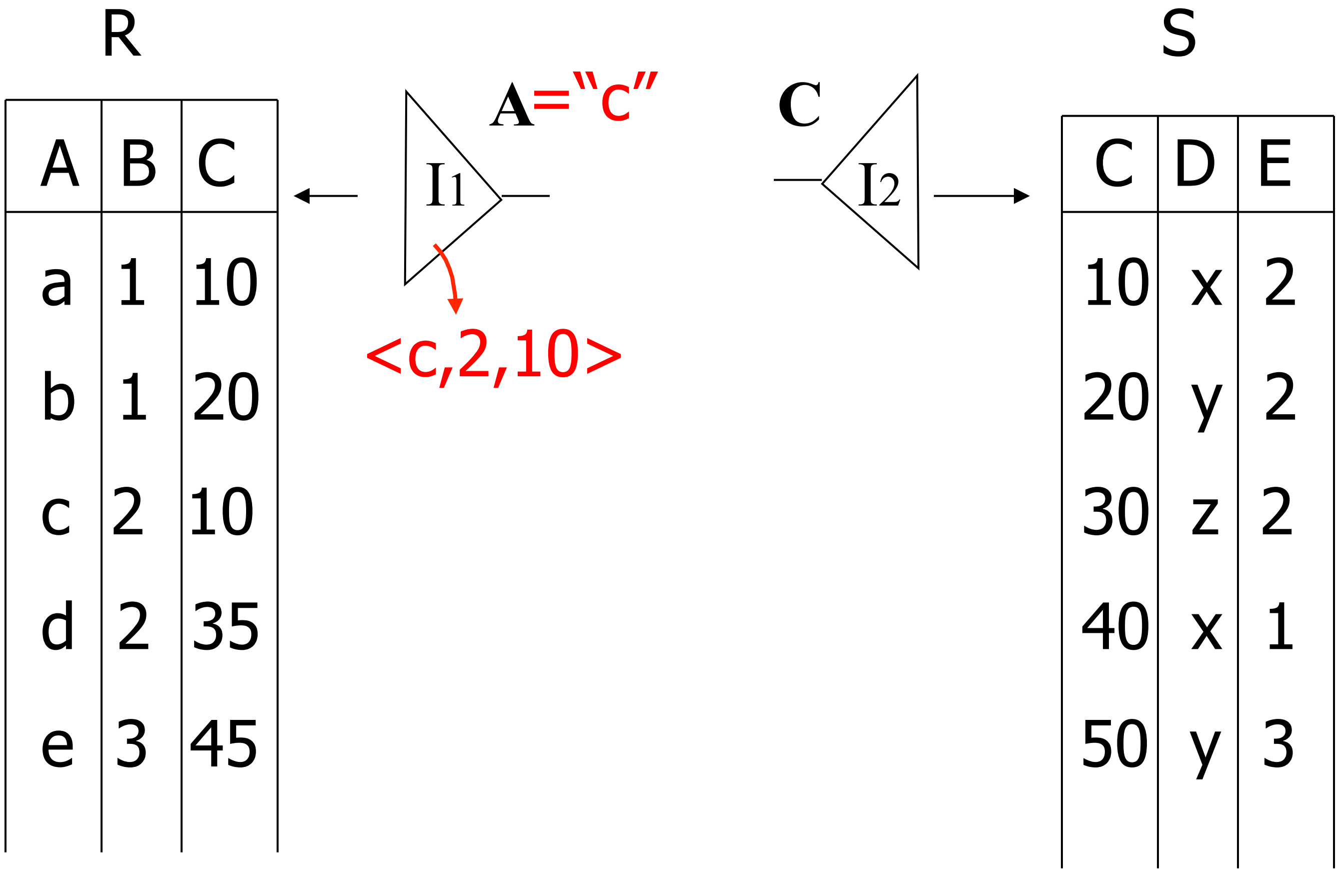
(4) Join matching R,S tuples, project B,D attributes and place in result

R

A	B	C
a	1	10
b	1	20
c	2	10
d	2	35
e	3	45

S

C	D	E
10	x	2
20	y	2
30	z	2
40	x	1
50	y	3



R

A	B	C
a	1	10
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C	D	E
10	x	2
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40	x	1
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R

A	B	C
a	1	10
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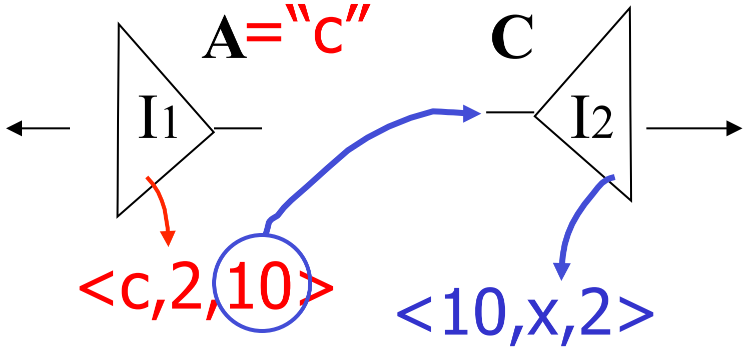
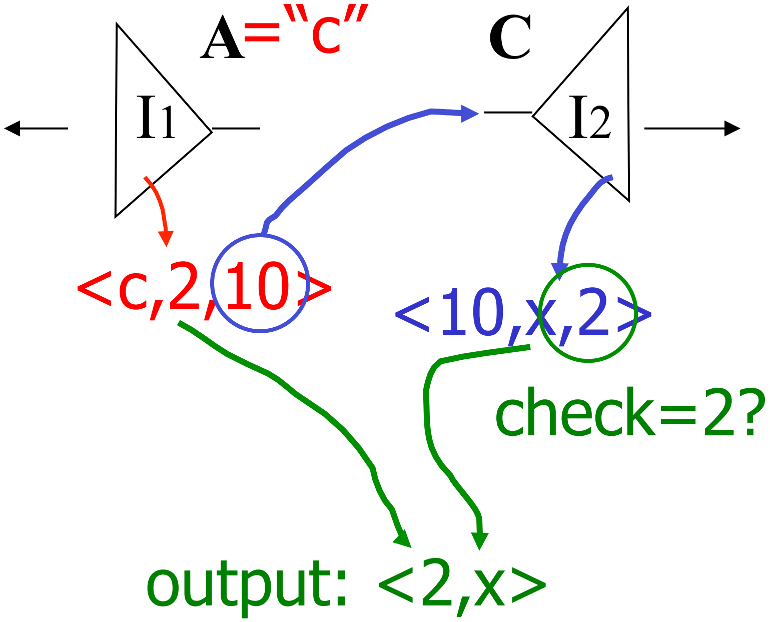


S

C	D	E
10	x	2
20	y	2
30	z	2
40	x	1
50	y	3

R

A	B	C
a	1	10
b	1	20
c	2	10
d	2	35
e	3	45



A="c"

C

<c,2,10>

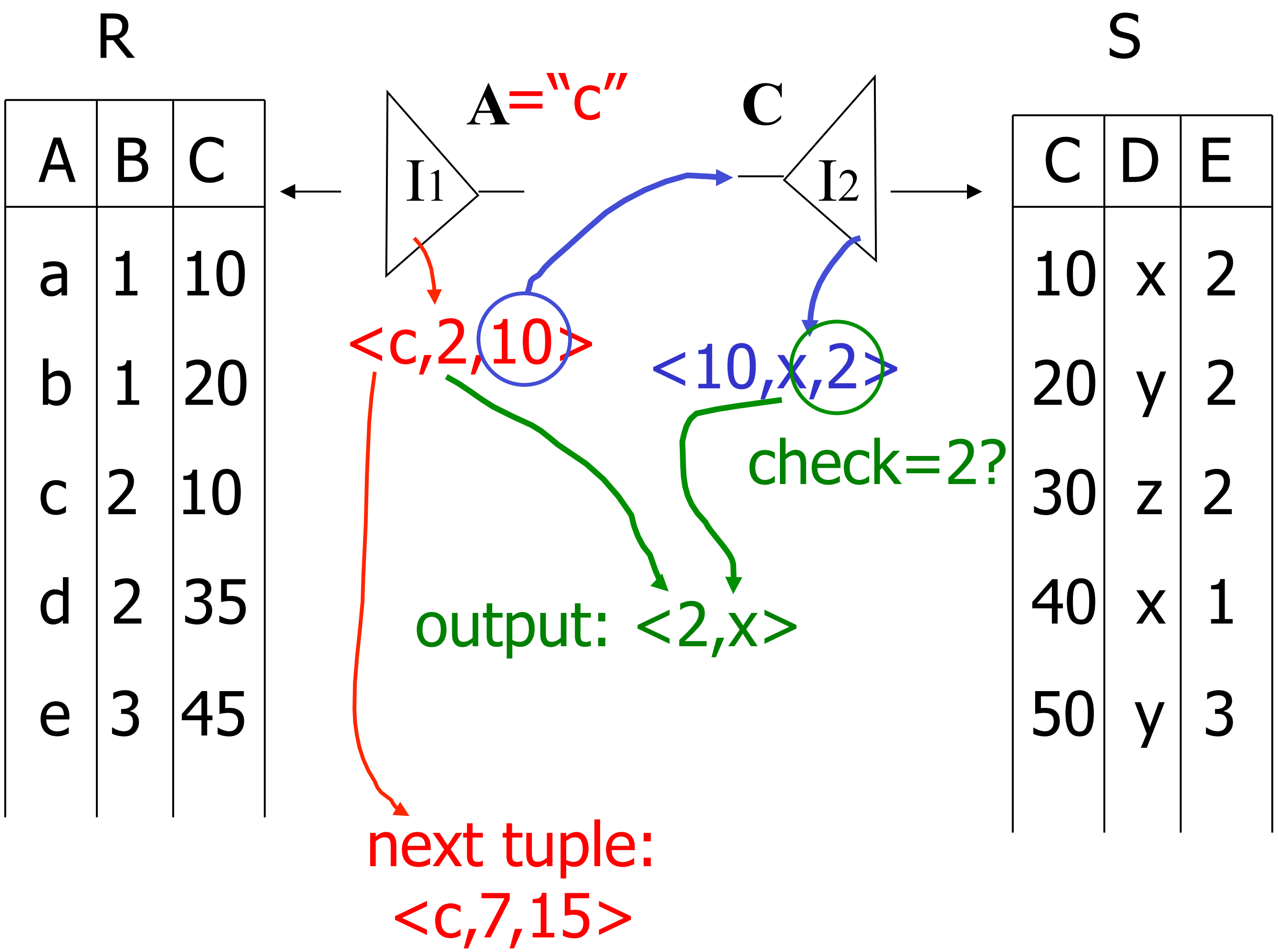
<10,x,2>

check=2?

output: <2,x>

S

C	D	E
10	x	2
20	y	2
30	z	2
40	x	1
50	y	3



R

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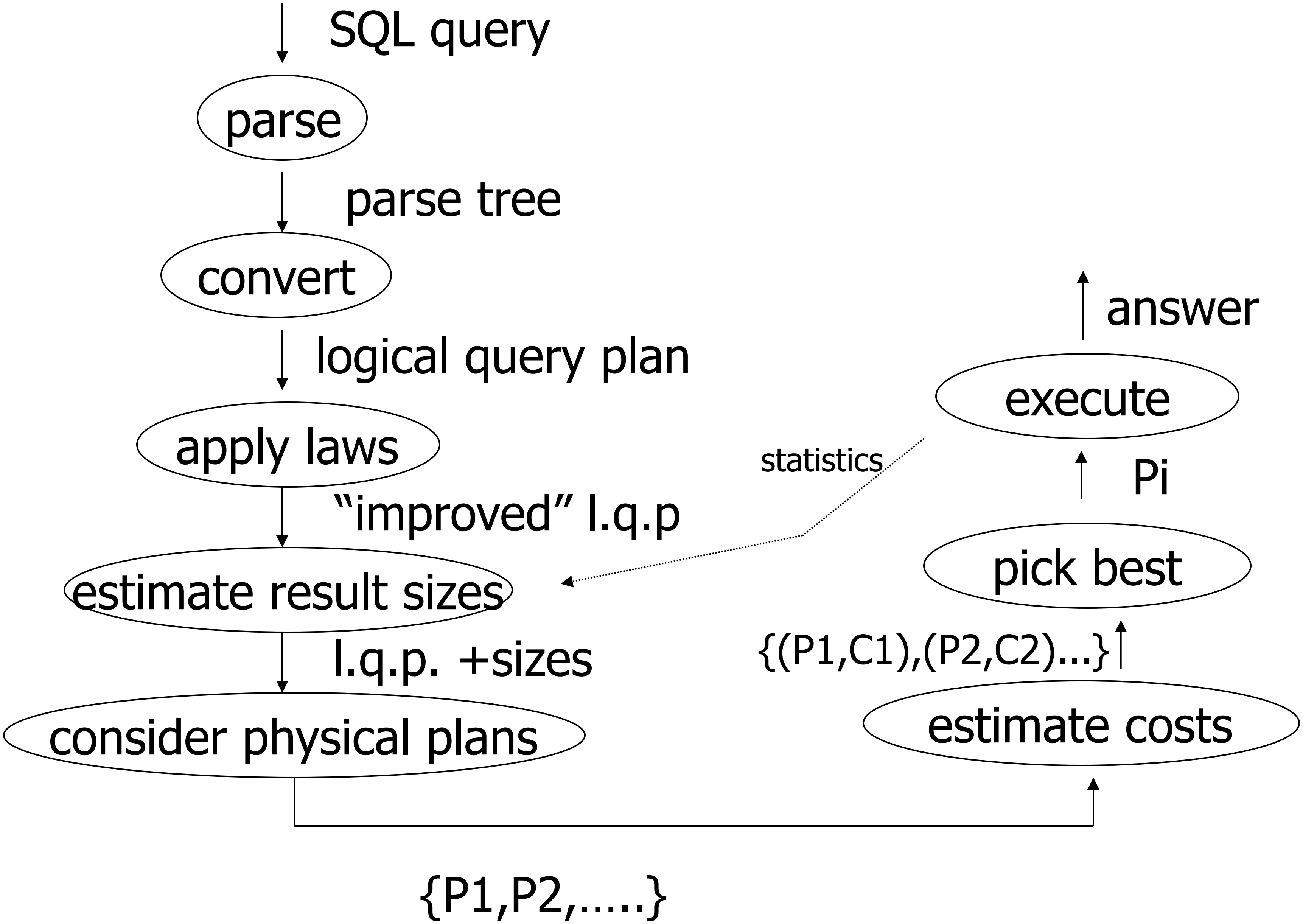
Overview of Query Optimization



SQL query

parse

parse tree

convert

logical query plan

apply laws

“improved” l.q.p

estimate result sizes

l.q.p. +sizes

consider physical plans

{P1,P2,…..}

estimate costs

{(P1,C1),(P2,C2)...}

pick best

Pi

execute

answer

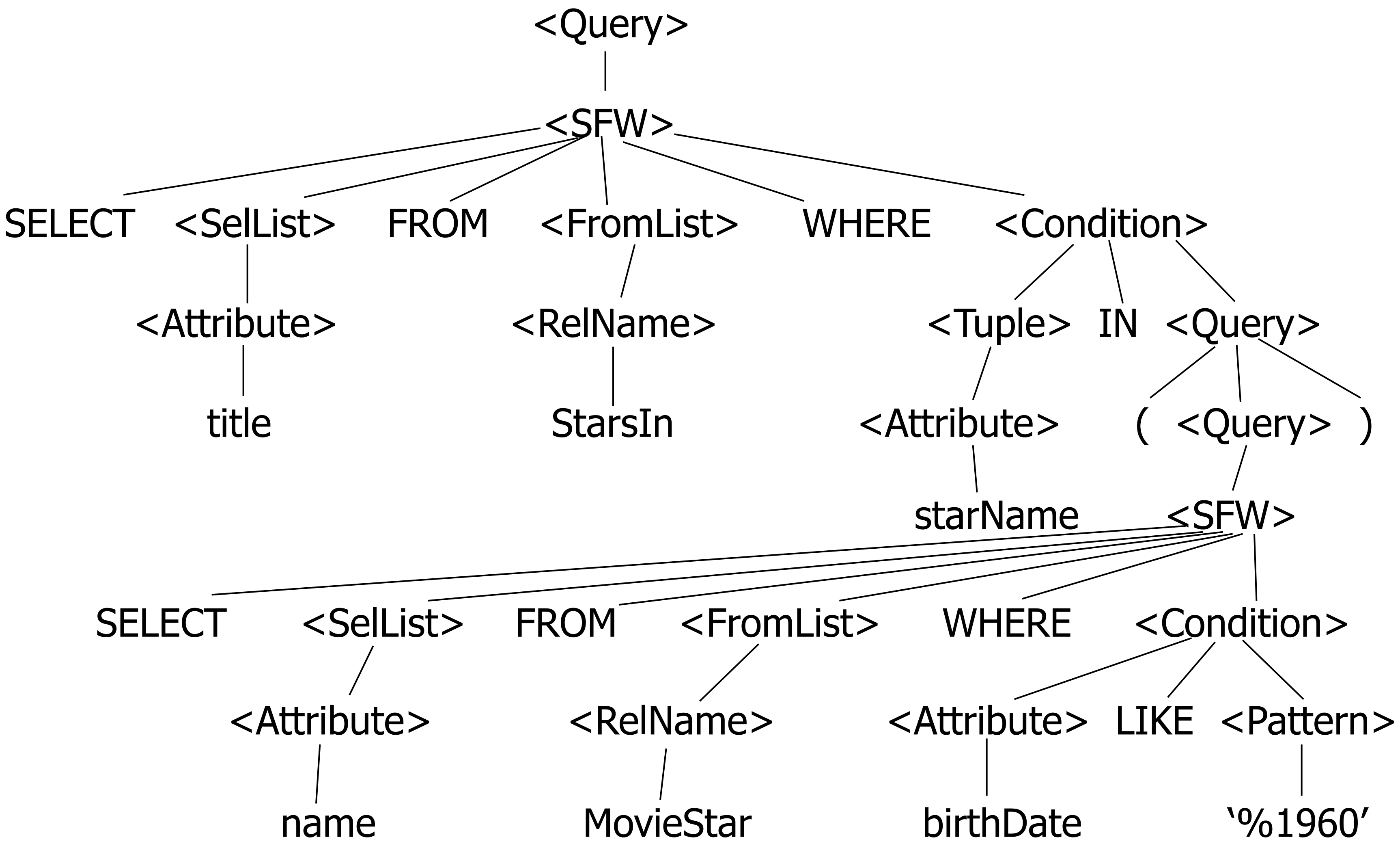
statistics

Example: SQL query

```
SELECT title
FROM StarsIn
WHERE starName IN (
        SELECT name
        FROM MovieStar
        WHERE birthdate LIKE ‘%1960’
);
```

(Find the movies with stars born in 1960)

Example: Parse Tree



<Query>

<SFW>

SELECT <SelList> FROM <FromList> WHERE <Condition>

<Attribute> <RelName> <Tuple> IN <Query>

title StarsIn <Attribute> (<Query>)

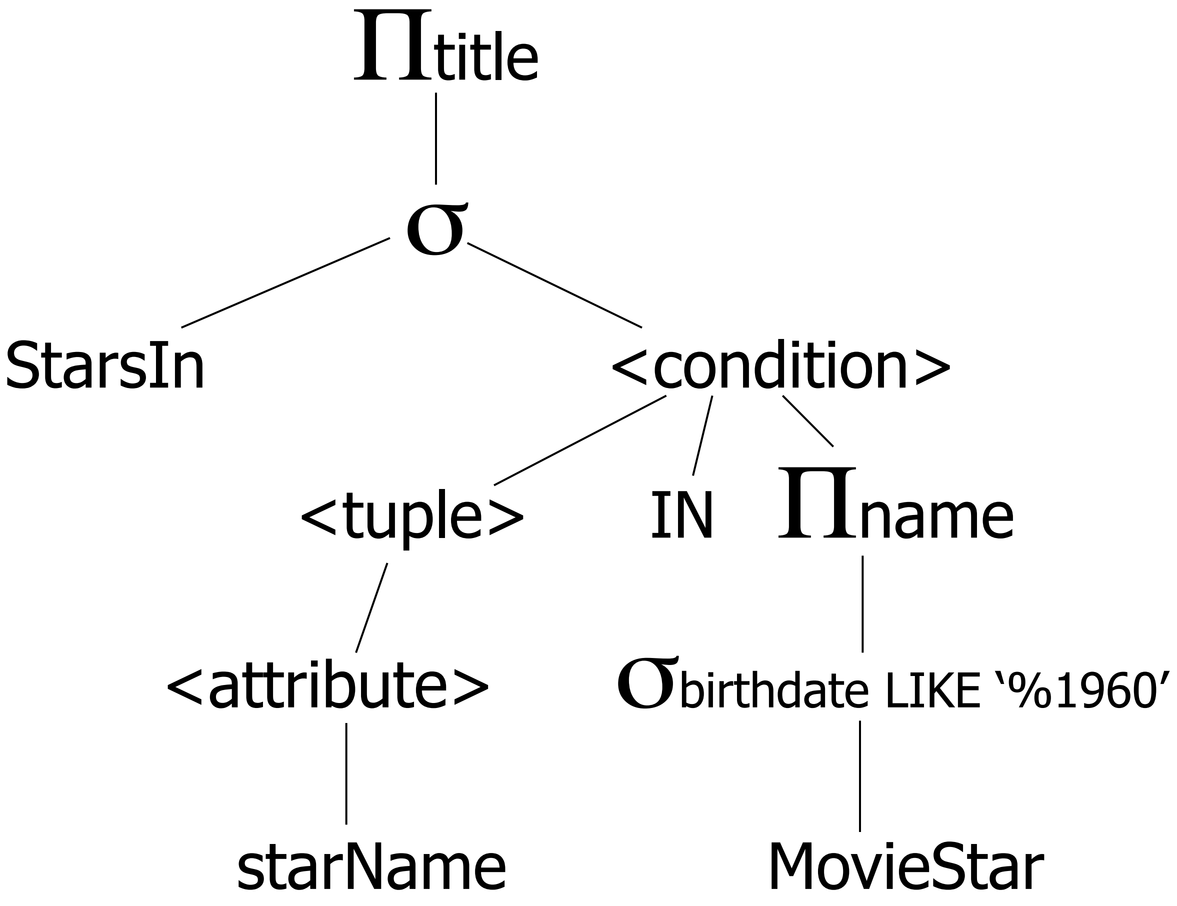
starName <SFW>

SELECT <SelList> FROM <FromList> WHERE <Condition>

<Attribute> <RelName> <Attribute> LIKE <Pattern>

name MovieStar birthDate '%1960'

Example: Generating Relational Algebra



Π_{title}

σ

StarsIn

<condition>

<tuple>

IN

Π_{name}

<attribute>

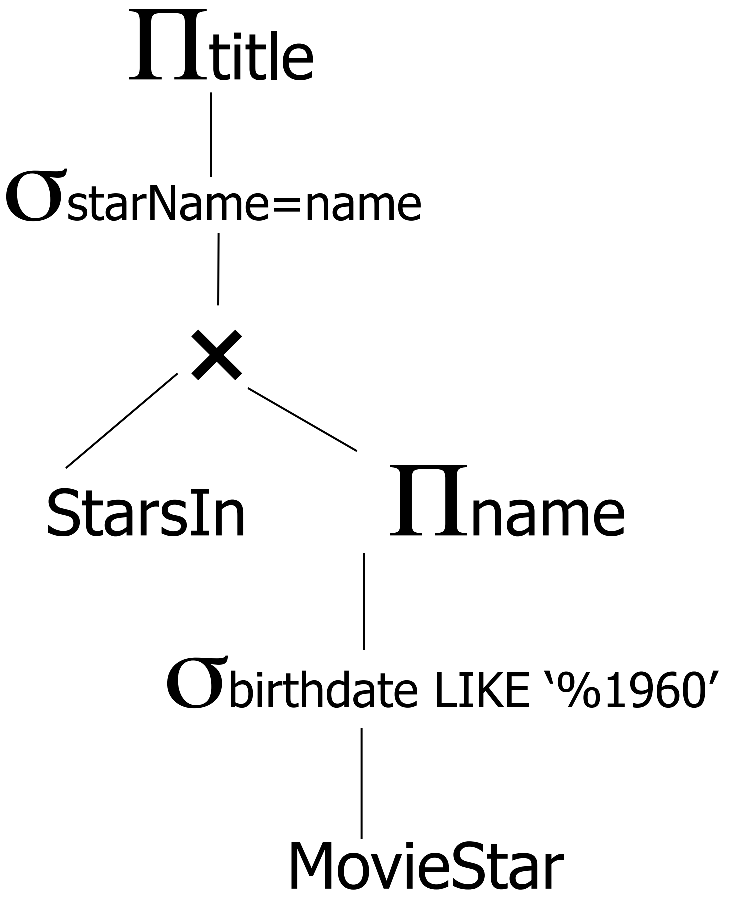
$\sigma_{birthdate\ LIKE\ `\%1960'}$

starName

MovieStar

Fig. 7.15: An expression using a two-argument σ, midway between a parse tree and relational algebra

Example: Logical Query Plan



Π_{title}

$\sigma_{\text{starName=name}}$

$\times$

StarsIn

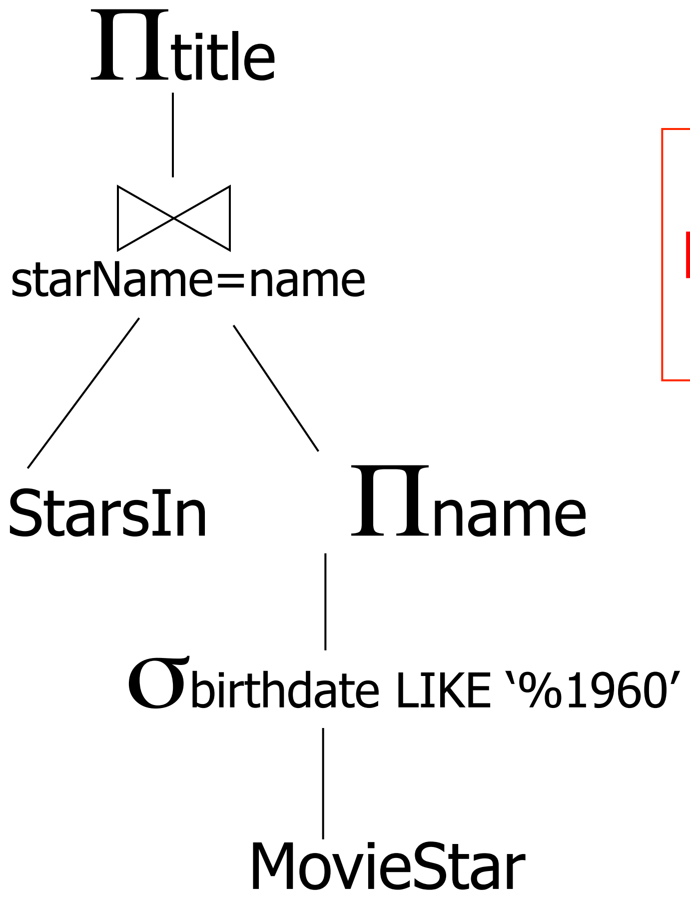
Π_{name}

$\sigma_{\text{birthdate LIKE ‘\%1960’}}$

MovieStar

Fig. 7.18: Applying the rule for IN conditions

Example: Improved Logical Query Plan



Π_{title}

$\bowtie_{\text{starName=name}}$

StarsIn

Π_{name}

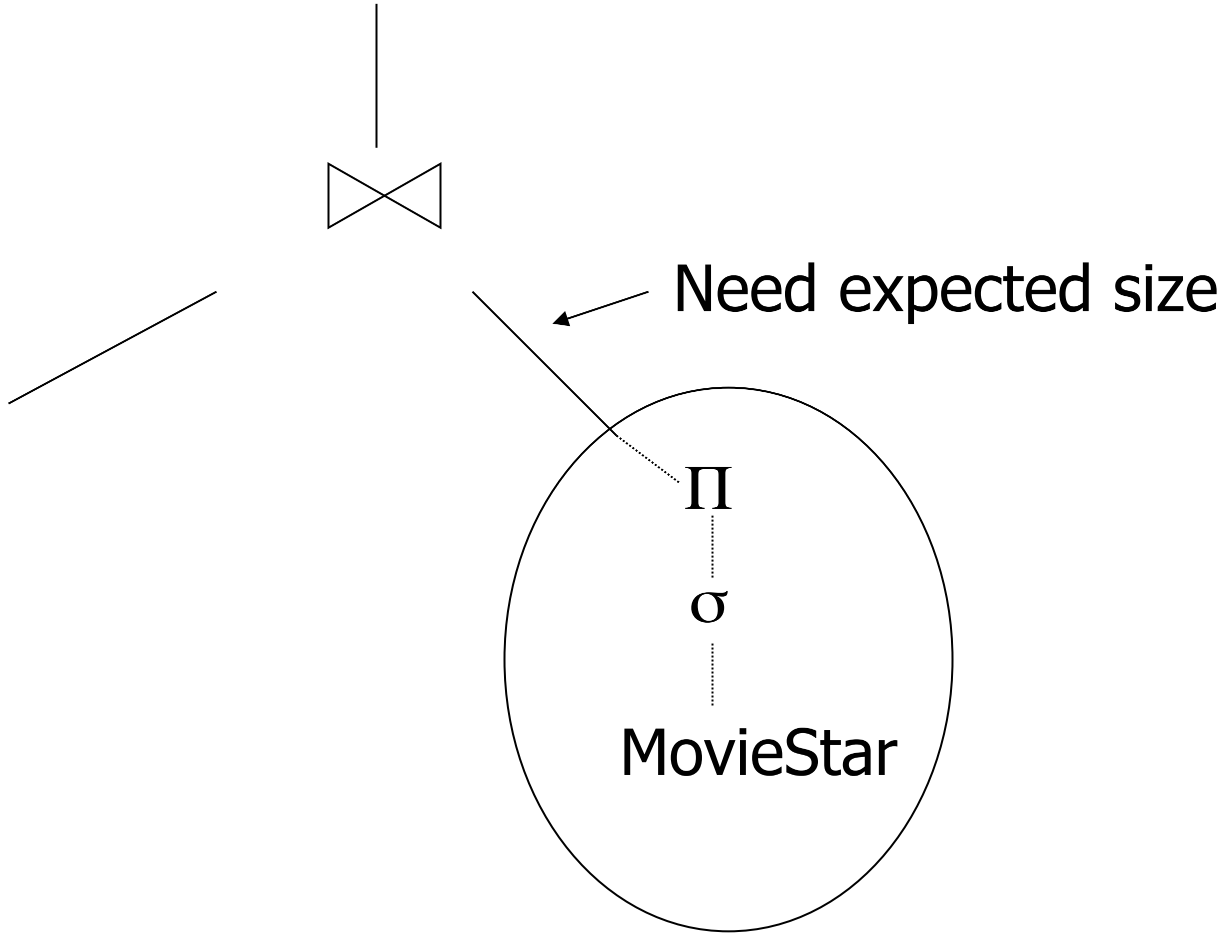
$\sigma_{\text{birthdate LIKE '\%1960'}}$

MovieStar

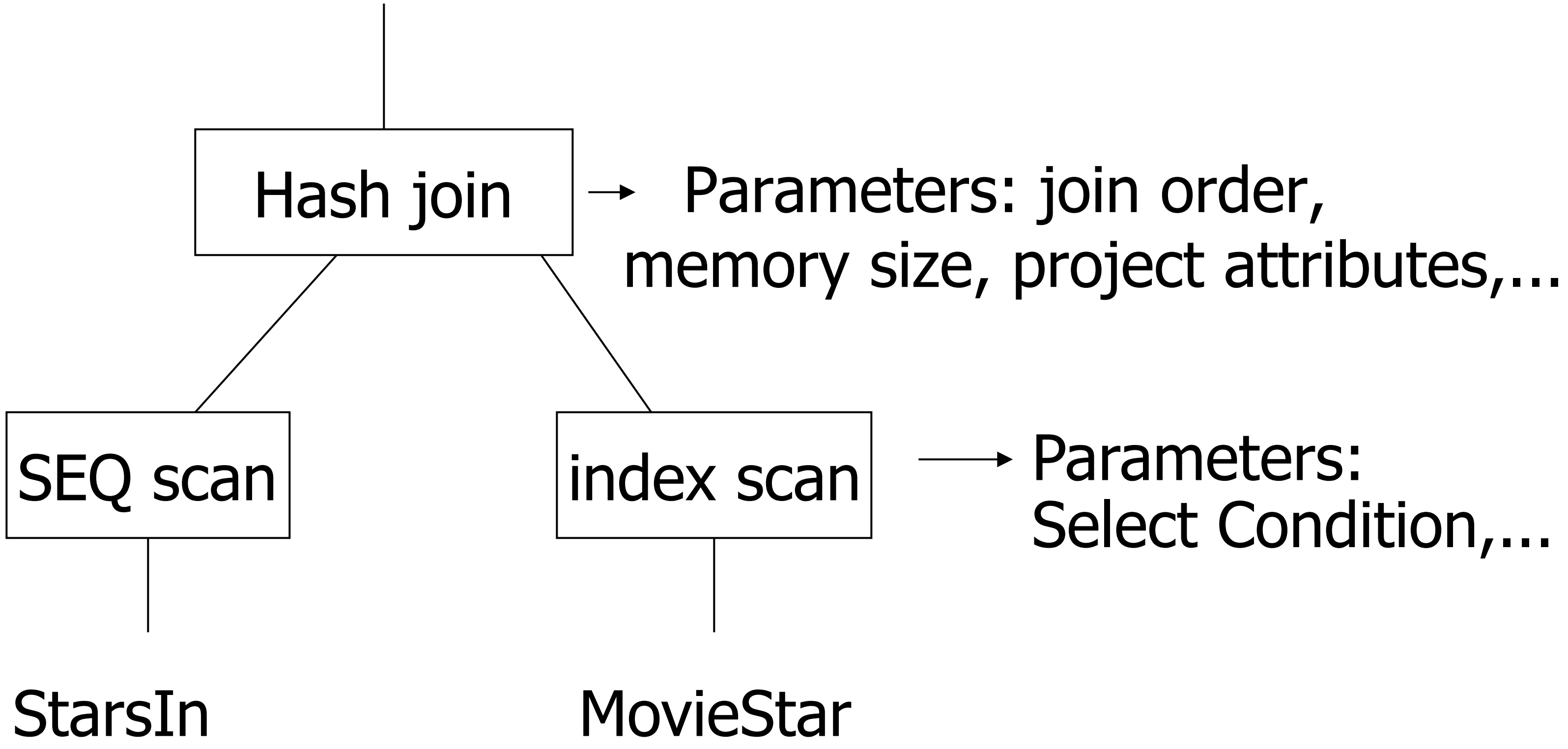
Question:
Push project to
StarsIn?

Fig. 7.20: An improvement on fig. 7.18.

Example: Estimate Result Sizes



Example: One Physical Plan



Hash join → Parameters: join order, memory size, project attributes,...

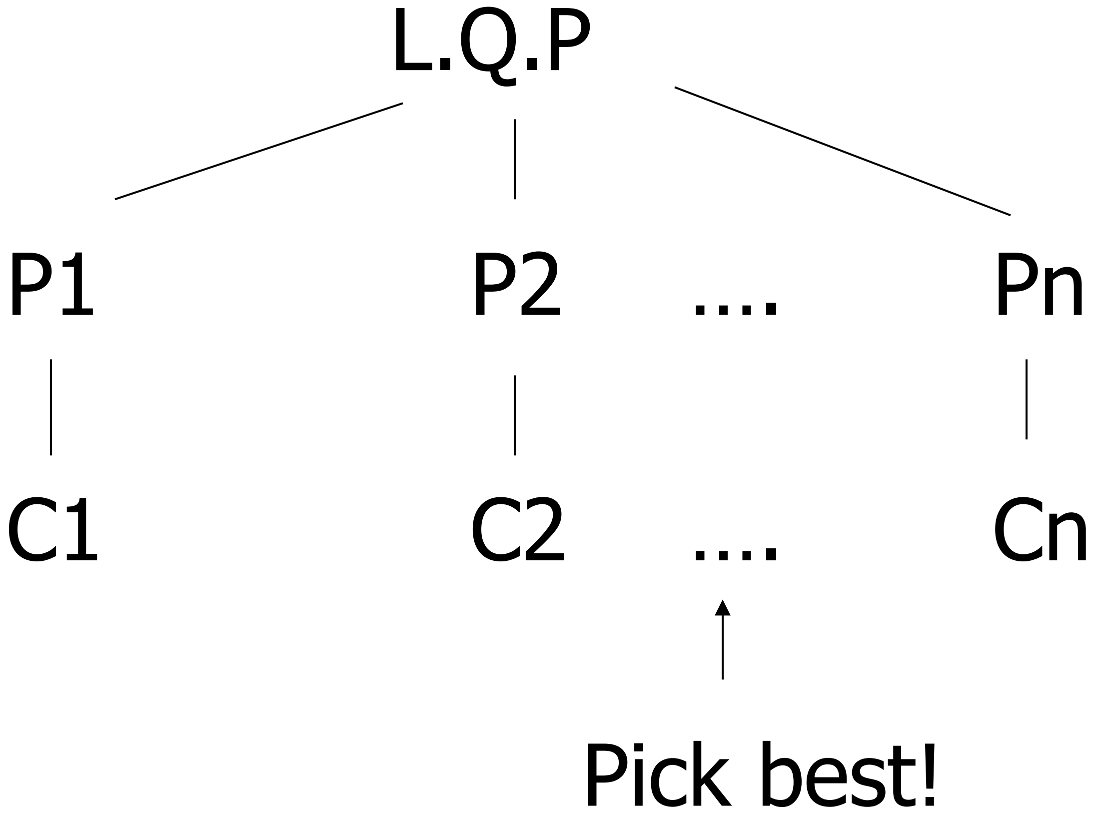
SEQ scan

index scan → Parameters: Select Condition,...

StarsIn

MovieStar

Example: Estimate costs



L.Q.P

P1 P2 …. Pn

C1 C2 …. Cn

Pick best!

Textbook outline

Chapter 15

5 Algebra for queries [bags vs sets]

[Ch 5] - Select, project, join, …. [project list a,a+b->x,…]

- Duplicate elimination, grouping, sorting

15.1 Physical operators

[15.1] - Scan,sort, …

15.2 - 15.6 Implementing operators + estimating their cost

[15.2-15.6]

Chapter 16

16.1[16.1]	Parsing
16.2[16.2]	Algebraic laws
16.3[16.3]	Parse tree -> logical query plan
16.4[16.4]	Estimating result sizes
16.5-7[16.5-7]	Cost based optimization

<u>Reading textbook</u> - Chapters 15, 16

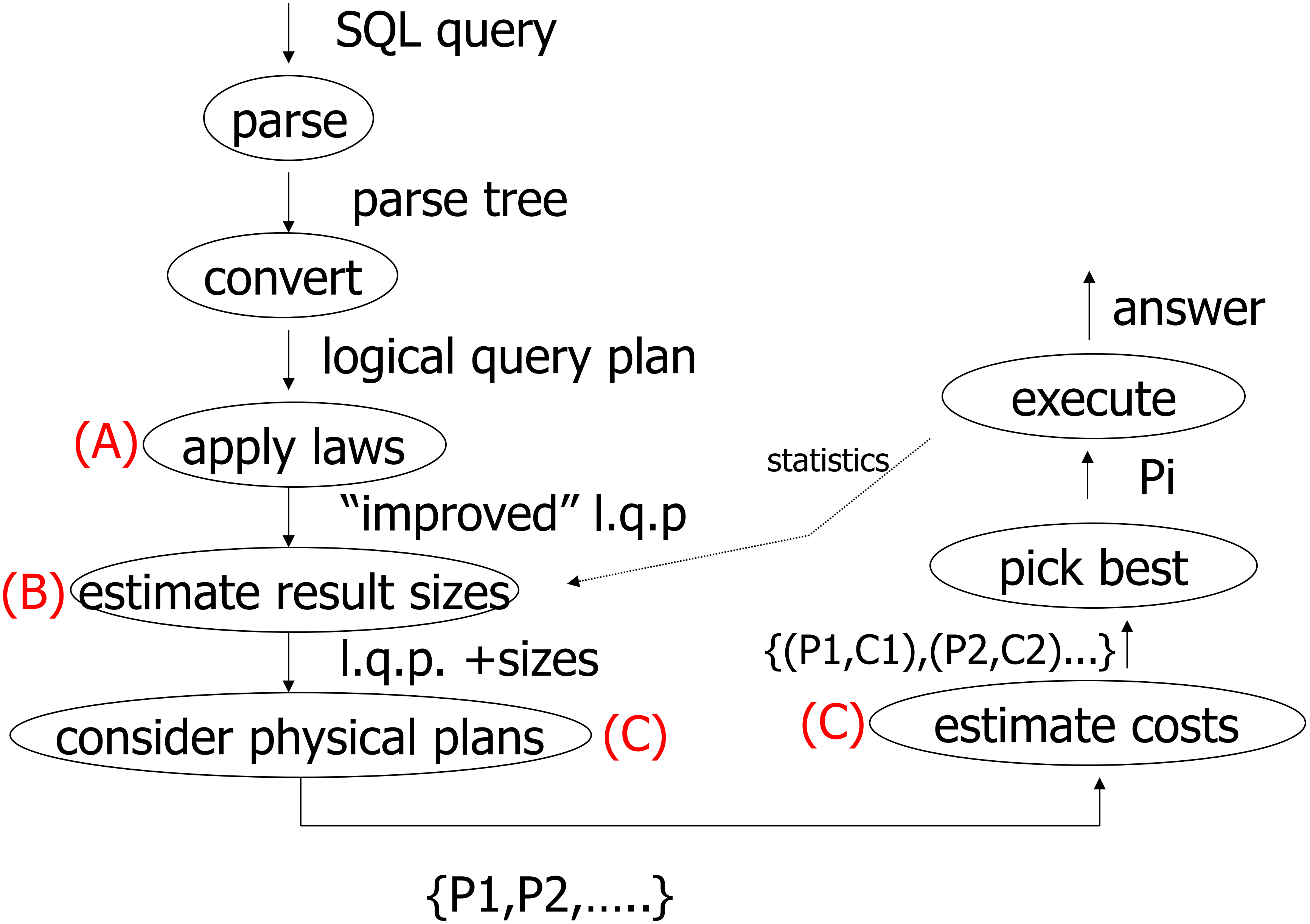
Optional:

- Sections 15.7, 15.8, 15.9 [15.7, 15.8]
- Sections 16.6, 16.7 [16.6, 16.7]

Optional: Duplicate elimination operator
grouping, aggregation operators

Query Optimization - In class order

- Relational algebra level (A)
- Detailed query plan level
  - Estimate Costs (B)
    - without indexes
    - with indexes
  - Generate and compare plans (C)



SQL query

parse

parse tree

convert

logical query plan

(A) apply laws

"improved" l.q.p

(B) estimate result sizes

l.q.p. +sizes

consider physical plans (C)

{P1,P2,.....}

(C) estimate costs

{(P1,C1),(P2,C2)...}

pick best

Pi

execute

answer

statistics

Relational algebra optimization

- Transformation rules
  (preserve equivalence)
- What are good transformations?

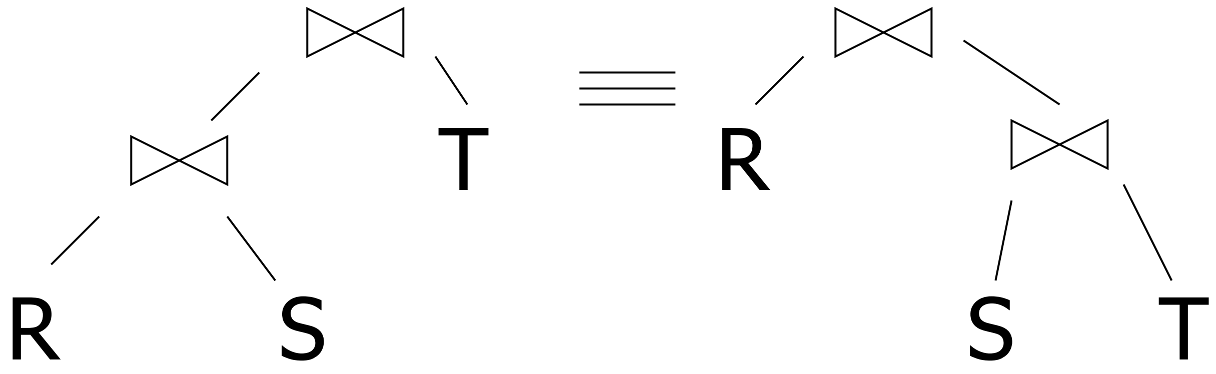
Rules: Natural joins & cross products & union

$R \bowtie S = S \bowtie R$

$(R \bowtie S) \bowtie T = R \bowtie (S \bowtie T)$

Note:

- Carry attribute names in results, so order is not important
- Can also write as trees, e.g.:



Rules: Natural joins & cross products & union

$R \bowtie S = S \bowtie R$

$(R \bowtie S) \bowtie T = R \bowtie (S \bowtie T)$

$R \times S = S \times R$

$(R \times S) \times T = R \times (S \times T)$

$R \cup S = S \cup R$

$R \cup (S \cup T) = (R \cup S) \cup T$

Rules: Selects

$\sigma_{p1 \wedge p2}(R) =$

$\sigma_{p1 \vee p2}(R) =$

Rules: Selects

$$\sigma_{p1 \wedge p2}(R) = \sigma_{p1}\ [\ \sigma_{p2}\ (R)]$$

$$\sigma_{p1 \vee p2}(R) = [\ \sigma_{p1}\ (R)] \cup [\ \sigma_{p2}\ (R)]$$

Bags vs. Sets

R = {a,a,b,b,b,c}

S = {b,b,c,c,d}

RUS = ?

Bags vs. Sets

R = {a,a,b,b,b,c}

S = {b,b,c,c,d}

RUS = ?

- Option 1 SUM

  RUS = {a,a,b,b,b,b,b,c,c,c,d}
- Option 2 MAX

  RUS = {a,a,b,b,b,c,c,d}

Option 2 (MAX) makes this rule work:

$\sigma_{p1 \vee p2}$ (R) = σ_{p1}(R) U σ_{p2}(R)

Example: R={a,a,b,b,b,c}

P1 satisfied by a,b; P2 satisfied by b,c

Option 2 (MAX) makes this rule work:

$\sigma_{p1 \vee p2}$ (R) = σ_{p1}(R) U σ_{p2}(R)

Example: R={a,a,b,b,b,c}

P1 satisfied by a,b; P2 satisfied by b,c

$\sigma_{p1 \vee p2}$ (R) = {a,a,b,b,b,c}

σ_{p1}(R) = {a,a,b,b,b}

σ_{p2}(R) = {b,b,b,c}

σ_{p1}(R) U σ_{p2} (R) = {a,a,b,b,b,c}

“Sum” option makes more sense:

Senators (……) Rep (……)

T1 = $\pi_{yr,state}$ Senators; T2 = $\pi_{yr,state}$ Reps

T1

Yr	State
14	CA
16	CA
15	AZ

T2

Yr	State
16	CA
16	CA
15	CA

Union?

Executive Decision

-> Use “SUM” option for bag unions

-> Some rules cannot be used for bags

Rules: Project

Let: X = set of attributes

Y = set of attributes

XY = X U Y

$\pi_{xy}(R) =$

Rules: Project

Let: X = set of attributes

Y = set of attributes

XY = X U Y

$$\pi_{xy}(R) = \pi_x[\pi_y(R)]$$

Rules: Project

Let: X = set of attributes

Y = set of attributes

XY = X U Y

$$\pi_{xy}(R) = \xcancel{\pi_{x}[\pi_{y}(R)]}$$

Rules: $\sigma + \bowtie$ combined

Let p = predicate with only R attribs
q = predicate with only S attribs
m = predicate with only R,S attribs

σ_p (R $\bowtie$ S) =

σ_q (R $\bowtie$ S) =

Rules: $\sigma + \bowtie$ combined

Let p = predicate with only R attribs

q = predicate with only S attribs

m = predicate with only R,S attribs

$$\sigma_p (R \bowtie S) = [\sigma_p (R)] \bowtie S$$

$$\sigma_q (R \bowtie S) = R \bowtie [\sigma_q (S)]$$

Rules: $\sigma + \bowtie$ combined (continued)

Some Rules can be Derived:

$\sigma_{p \wedge q} (R \bowtie S) =$

$\sigma_{p \wedge q \wedge m} (R \bowtie S) =$

$\sigma_{p \vee q} (R \bowtie S) =$

Do one, others for homework:

$$\sigma_{p \wedge q} (R \bowtie S) = [\sigma_p (R)] \bowtie [\sigma_q (S)]$$

$$\sigma_{p \wedge q \wedge m} (R \bowtie S) =$$

$$\sigma_m \left[(\sigma_p R) \bowtie (\sigma_q S)\right]$$

$$\sigma_{p \vee q} (R \bowtie S) =$$

$$\left[(\sigma_p R) \bowtie S\right] \cup \left[R \bowtie (\sigma_q S)\right]$$

--> Derivation for first one:

$\sigma_{p \wedge q} (R \bowtie S) =$

--> Derivation for first one:

$$\sigma_{p \wedge q} (R \bowtie S) =$$

$$\sigma_p \left[\sigma_q (R \bowtie S) \right] =$$

$$\sigma_p \left[R \bowtie \sigma_q (S) \right] =$$

$$[\sigma_p (R)] \bowtie [\sigma_q (S)]$$

Rules: π,σ combined

Let x = subset of R attributes

z = attributes in predicate P

(subset of R attributes)

$\pi_x[\sigma_p (R)] =$

Rules: π,σ combined

Let x = subset of R attributes
z = attributes in predicate P
(subset of R attributes)

$$\pi_x[\sigma_p(R)] = \{\sigma_p[\pi_x(R)]\}$$

Rules: π, σ combined

Let x = subset of R attributes

z = attributes in predicate P

(subset of R attributes)

$$\pi_x[\sigma_p(R)] = \pi_x\{\sigma_p[\overset{\pi_{xz}}{\cancel{\pi_x}}(R)]\}$$

Rules: π, $\bowtie$ combined

Let x = subset of R attributes
y = subset of S attributes
z = intersection of R,S attributes

$\pi_{xy}(R \bowtie S) =$

Rules: π, $\bowtie$ combined

Let x = subset of R attributes

y = subset of S attributes

z = intersection of R,S attributes

$$\pi_{xy}(R \bowtie S) =$$

$$\pi_{xy}\{[\pi_{xz}(R)] \bowtie [\pi_{yz}(S)]\}$$

$$\pi_{xy}\left\{\sigma_{p}\left(R \bowtie S\right)\right\} =$$

$$\pi_{xy}\left\{\sigma_p\,(R \bowtie S)\right\} \; =$$

$$\pi_{xy}\left\{\sigma_p\,[\pi_{xz'}(R) \bowtie \pi_{yz'}(S)]\right\}$$

$$z' = z \cup \{\text{attributes used in P}\}$$

Rules for σ, π combined with X

similar...

e.g., $\sigma_p(R \times S) = ?$

Rules σ, U combined:

$\sigma_p(R \cup S) = \sigma_p(R) \cup \sigma_p(S)$

$\sigma_p(R - S) = \sigma_p(R) - S = \sigma_p(R) - \sigma_p(S)$

Which are "good" transformations?

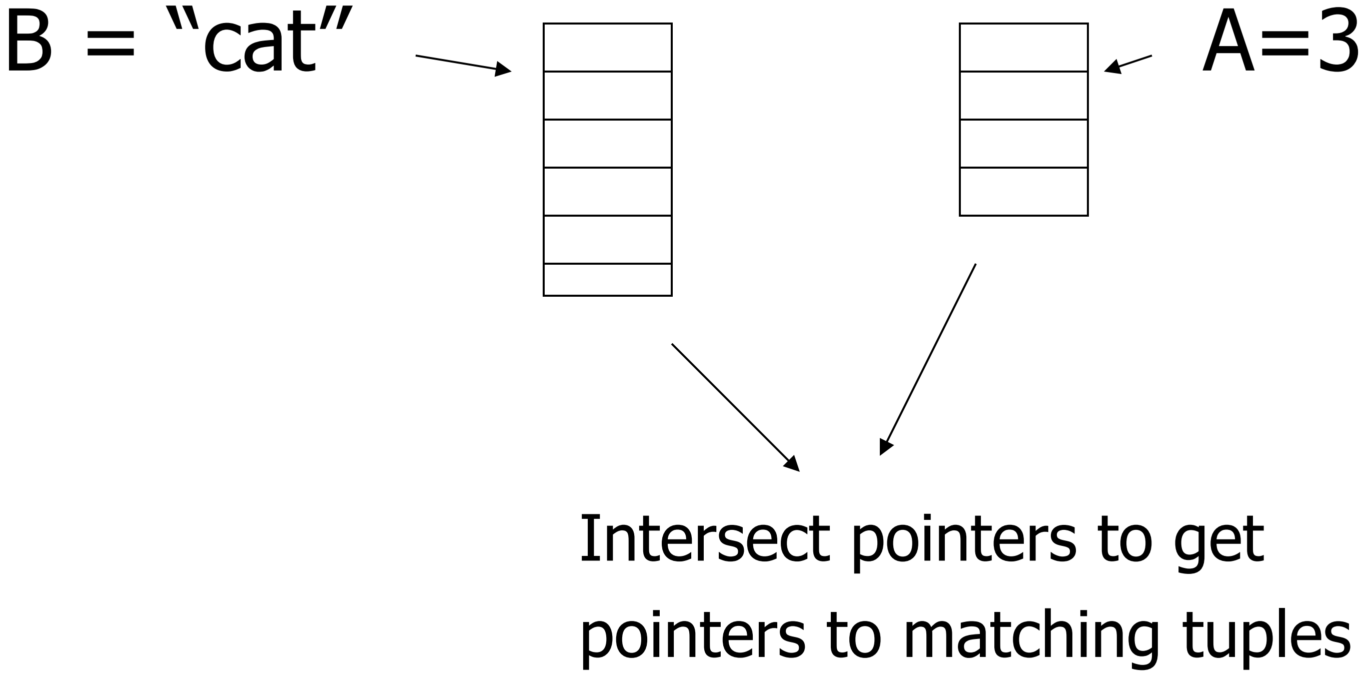
- $\sigma_{p1 \wedge p2}(R) \rightarrow \sigma_{p1}[\sigma_{p2}(R)]$
- $\sigma_{p}(R \bowtie S) \rightarrow [\sigma_{p}(R)] \bowtie S$
- $R \bowtie S \rightarrow S \bowtie R$
- $\pi_{x}[\sigma_{p}(R)] \rightarrow \pi_{x}\{\sigma_{p}[\pi_{xz}(R)]\}$

Conventional wisdom:
do projects early

Example: R(A,B,C,D,E) x={E}
P: (A=3) ∧ (B="cat")

$\pi_x \{\sigma_p (R)\}$ vs. $\pi_E \left\{\sigma_p\{\pi_{ABE}(R)\}\right\}$

But What if we have A, B indexes?



B = “cat”

A=3

Intersect pointers to get pointers to matching tuples

Bottom line:

- No transformation is always good
- Usually good: early selections

In textbook: more transformations

- Eliminate common sub-expressions
- Other operations: duplicate elimination

Outline - Query Processing

- Relational algebra level
  - transformations
  - good transformations
- Detailed query plan level
  - estimate costs
  - generate and compare plans

- Estimating cost of query plan

(1) Estimating size of results
(2) Estimating # of IOs

Estimating result size

- Keep statistics for relation R
  - T(R) : # tuples in R
  - S(R) : # of bytes in each R tuple
  - B(R): # of blocks to hold all R tuples
  - V(R, A) : # distinct values in R
    for attribute A

Example

R

A	B	C	D
cat	1	10	a
cat	1	20	b
dog	1	30	a
dog	1	40	c
bat	1	50	d

A: 20 byte string

B: 4 byte integer

C: 8 byte date

D: 5 byte string

Example

R

A	B	C	D
cat	1	10	a
cat	1	20	b
dog	1	30	a
dog	1	40	c
bat	1	50	d

A: 20 byte string

B: 4 byte integer

C: 8 byte date

D: 5 byte string

T(R) = 5 S(R) = 37

V(R,A) = 3 V(R,C) = 5

V(R,B) = 1 V(R,D) = 4

Size estimates for W = R1 x R2

T(W) =

S(W) =

Size estimates for W = R1 x R2

T(W) = T(R1) × T(R2)

S(W) = S(R1) + S(R2)

Size estimate for $W = \sigma_{A=a}(R)$

$S(W) = S(R)$

$T(W) = ?$

Example

R

A	B	C	D
cat	1	10	a
cat	1	20	b
dog	1	30	a
dog	1	40	c
bat	1	50	d

$V(R,A)=3$

$V(R,B)=1$

$V(R,C)=5$

$V(R,D)=4$

$W = \sigma_{z=val}(R)$ $T(W) =$

Example

R

A	B	C	D
cat	1	10	a
cat	1	20	b
dog	1	30	a
dog	1	40	c
bat	1	50	d

V(R,A)=3

V(R,B)=1

V(R,C)=5

V(R,D)=4

what is probability this tuple will be in answer?

$W = \sigma_{z=val}(R)$ $T(W) =$

Example

R

A	B	C	D
cat	1	10	a
cat	1	20	b
dog	1	30	a
dog	1	40	c
bat	1	50	d

$V(R,A)=3$

$V(R,B)=1$

$V(R,C)=5$

$V(R,D)=4$

$$W = \sigma_{z=val}(R) \quad T(W) = \frac{T(R)}{V(R,Z)}$$

Assumption:

Values in select expression Z = val
are uniformly distributed
over possible V(R,Z) values.

Alternate Assumption:

Values in select expression Z = val
are uniformly distributed
over domain with DOM(R,Z) values.

Example

R

A	B	C	D
cat	1	10	a
cat	1	20	b
dog	1	30	a
dog	1	40	c
bat	1	50	d

Alternate assumption

V(R,A)=3 DOM(R,A)=10

V(R,B)=1 DOM(R,B)=10

V(R,C)=5 DOM(R,C)=10

V(R,D)=4 DOM(R,D)=10

$W = \sigma_{z=val}(R) \quad T(W) = ?$

Example

R

A	B	C	D
cat	1	10	a
cat	1	20	b
dog	1	30	a
dog	1	40	c
bat	1	50	d

Alternate assumption

V(R,A)=3 DOM(R,A)=10

V(R,B)=1 DOM(R,B)=10

V(R,C)=5 DOM(R,C)=10

V(R,D)=4 DOM(R,D)=10

what is probability this tuple will be in answer?

$W = \sigma_{z=val}(R) \quad T(W) = ?$

Example

R

A	B	C	D
cat	1	10	a
cat	1	20	b
dog	1	30	a
dog	1	40	c
bat	1	50	d

Alternate assumption

$V(R,A)=3$ $DOM(R,A)=10$

$V(R,B)=1$ $DOM(R,B)=10$

$V(R,C)=5$ $DOM(R,C)=10$

$V(R,D)=4$ $DOM(R,D)=10$

$$W = \sigma_{z=val}(R) \quad T(W) = \frac{T(R)}{DOM(R,Z)}$$

Selection cardinality

SC(R,A) = average # records that satisfy equality condition on R.A

$$SC(R,A) = \begin{cases} \dfrac{T(R)}{V(R,A)} \\ \\ \dfrac{T(R)}{DOM(R,A)} \end{cases}$$

What about $W = \sigma_{z \geq val}(R)$?

$T(W) = ?$

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- Solution # 1:

  $T(W) = T(R)/2$

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$T(W) = ?$

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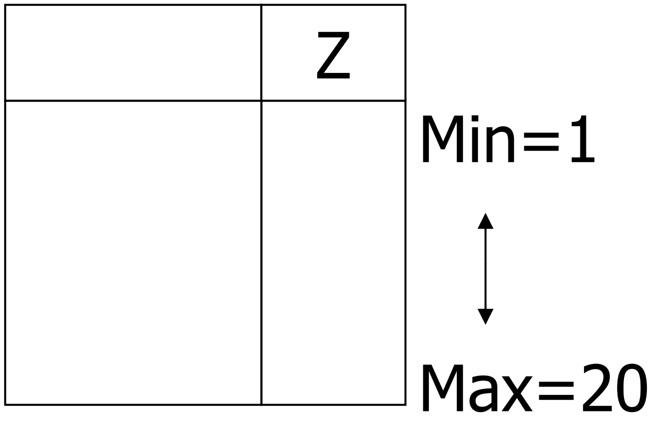
  $T(W) = T(R)/2$

- Solution # 2:

  $T(W) = T(R)/3$

- Solution # 3: Estimate values in range

<u>Example</u> R

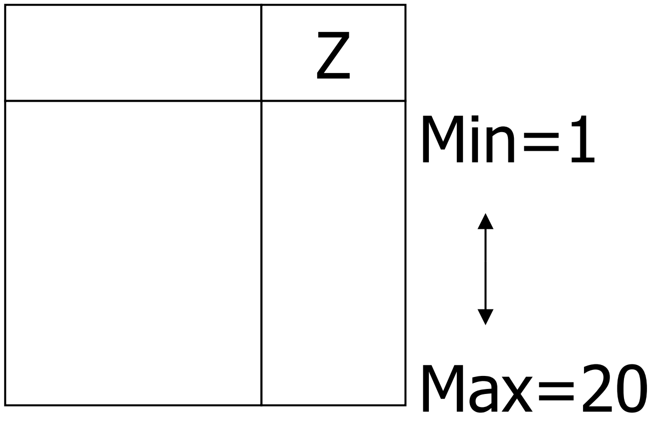


$V(R,Z)=10$

$W= \sigma_{Z \geq 15} (R)$

- Solution # 3: Estimate values in range

<u>Example</u> R



$V(R,Z)=10$

$W= \sigma_{Z \geq 15} (R)$

$$f = \frac{20-15+1}{20-1+1} = \frac{6}{20} \qquad \text{(fraction of range)}$$

$$T(W) = f \times T(R)$$

Equivalently:

$f \times V(R,Z)$ = fraction of distinct values

$$T(W) = [f \times V(Z,R)] \times \frac{T(R)}{V(Z,R)} = f \times T(R)$$

Size estimate for $W = R1 \bowtie R2$

Let x = attributes of R1

y = attributes of R2

Size estimate for W = R1 $\bowtie$ R2

Let x = attributes of R1

y = attributes of R2

Case 1

$$X \cap Y = \emptyset$$

Same as R1 x R2

Case 2 $W = R1 \bowtie R2 \qquad X \cap Y = A$

R1

A	B	C

R2

A	D

Case 2 $W = R1 \bowtie R2$ $X \cap Y = A$

R1

A	B	C

R2

A	D

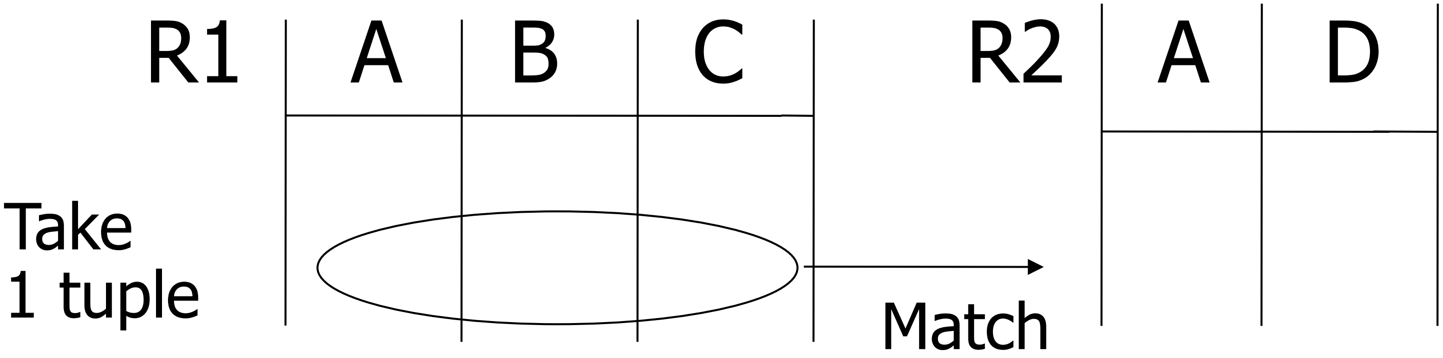
Assumption:

$V(R1,A) \leq V(R2,A) \Rightarrow$ Every A value in R1 is in R2

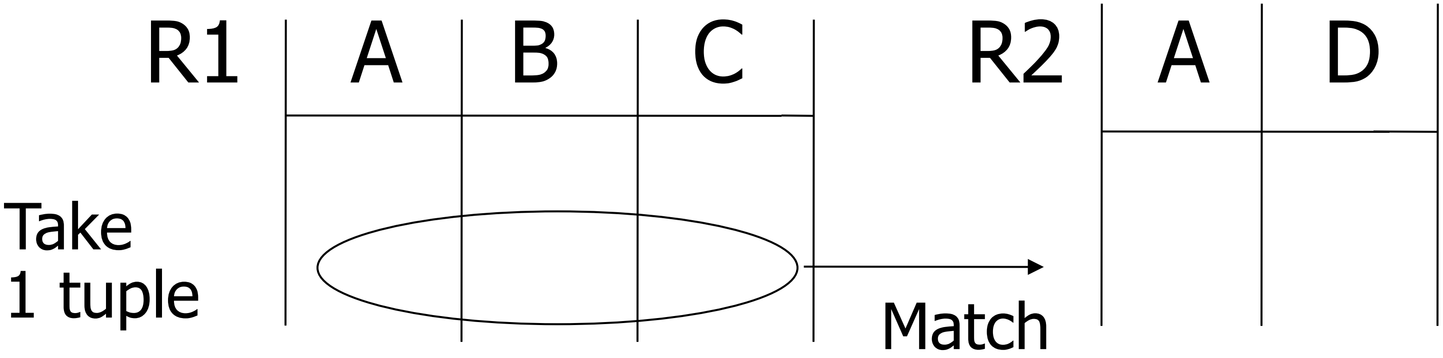
$V(R2,A) \leq V(R1,A) \Rightarrow$ Every A value in R2 is in R1

"containment of value sets" Sec. 7.4.4

Computing T(W) when V(R1,A) ≤ V(R2,A)



Computing T(W) when $V(R1,A) \le V(R2,A)$



1 tuple matches with $\frac{T(R2)}{V(R2,A)}$ tuples...

$$\text{so} \quad T(W) = \frac{T(R2)}{V(R2,\ A)} \times T(R1)$$

- $V(R1,A) \leq V(R2,A) \quad T(W) = \frac{T(R2)\ T(R1)}{V(R2,A)}$

- $V(R2,A) \leq V(R1,A) \quad T(W) = \frac{T(R2)\ T(R1)}{V(R1,A)}$

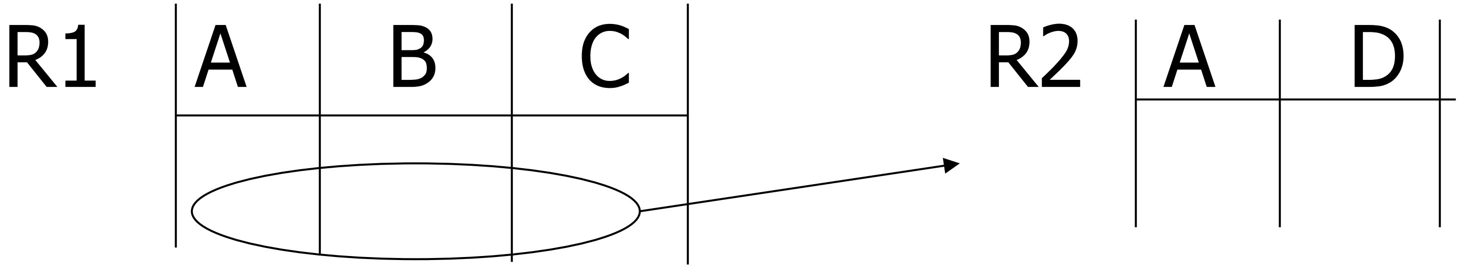
[A is common attribute]

In general $W = R1 \bowtie R2$

$$T(W) = \frac{T(R2)\ T(R1)}{\max\{\ V(R1,A),\ V(R2,A)\ \}}$$

$\boxed{\text{Case 2}}$ with alternate assumption

Values uniformly distributed over domain



This tuple matches T(R2)/DOM(R2,A) so

$$T(W) = \frac{T(R2)\ T(R1)}{DOM(R2,\ A)} = \frac{T(R2)\ T(R1)}{DOM(R1,\ A)}$$

Assume the same

In all cases:

$$S(W) = S(R1) + S(R2) - S(A)$$

size of attribute A

Using similar ideas, we can estimate sizes of:

$\Pi_{AB}(R)$ ….. Sec. 16.4.2 (same for either edition)

$\sigma_{A=a \wedge B=b}(R)$ …. Sec. 16.4.3

$R \bowtie S$ with common attribs. A,B,C

Sec. 16.4.5

Union, intersection, diff, ….

Sec. 16.4.7

Note: for complex expressions, need intermediate T,S,V results.

E.g. $W = [\underbrace{\sigma_{A=a}(R1)}_{\text{Treat as relation } U}] \bowtie R2$

$T(U) = T(R1)/V(R1,A) \qquad S(U) = S(R1)$

Also need V (U, *) !!

To estimate Vs

E.g., $U = \sigma_{A=a}(R1)$

Say R1 has attribs A,B,C,D

$V(U, A) =$

$V(U, B) =$

$V(U, C) =$

$V(U, D) =$

Example

R1

A	B	C	D
cat	1	10	10
cat	1	20	20
dog	1	30	10
dog	1	40	30
bat	1	50	10

$V(R1,A)=3$

$V(R1,B)=1$

$V(R1,C)=5$

$V(R1,D)=3$

$U = \sigma_{A=a}(R1)$

Example

R1

A	B	C	D
cat	1	10	10
cat	1	20	20
dog	1	30	10
dog	1	40	30
bat	1	50	10

$V(R1,A)=3$

$V(R1,B)=1$

$V(R1,C)=5$

$V(R1,D)=3$

$U = \sigma_{A=a}(R1)$

$V(U,A) = 1 \quad V(U,B) = 1 \quad V(U,C) = \frac{T(R1)}{V(R1,A)}$

$V(D,U)$ ... somewhere in between

Possible Guess $U = \sigma_{A=a}(R)$

$V(U,A) = 1$

$V(U,B) = V(R,B)$

For Joins $U = R1(A,B) \bowtie R2(A,C)$

$V(U,A) = \min \{ V(R1, A), V(R2, A) \}$

$V(U,B) = V(R1, B)$

$V(U,C) = V(R2, C)$

[called "preservation of value sets" in section 7.4.4]

Example:

$Z = R1(A,B) \bowtie R2(B,C) \bowtie R3(C,D)$

$\boxed{R1}$ T(R1) = 1000 V(R1,A)=50 V(R1,B)=100

$\boxed{R2}$ T(R2) = 2000 V(R2,B)=200 V(R2,C)=300

$\boxed{R3}$ T(R3) = 3000 V(R3,C)=90 V(R3,D)=500

Partial Result: $U = R1 \bowtie R2$

$$T(U) = \frac{1000 \times 2000}{200}$$

$V(U,A) = 50$

$V(U,B) = 100$

$V(U,C) = 300$

$Z = U \bowtie R3$

$$T(Z) = \frac{1000 \times 2000 \times 3000}{200 \times 300}$$

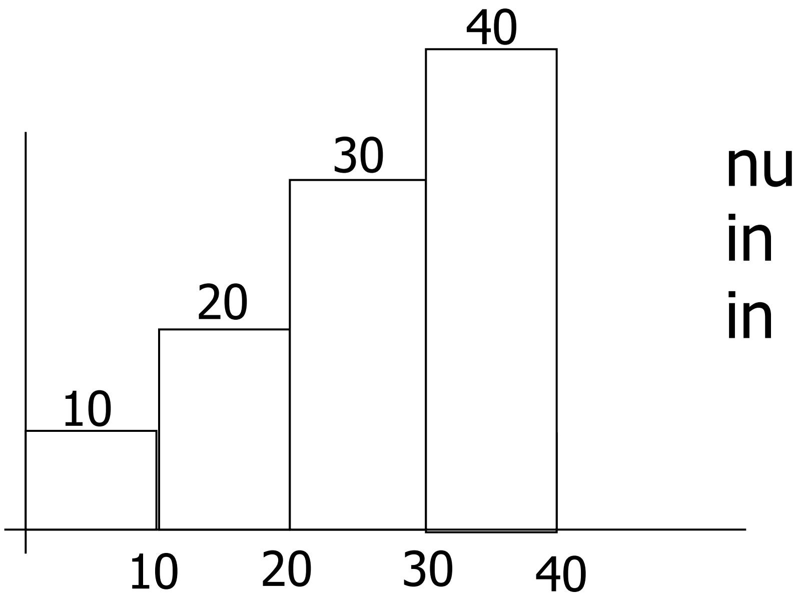
$V(Z,A) = 50$

$V(Z,B) = 100$

$V(Z,C) = 90$

$V(Z,D) = 500$

A Note on Histograms



number of tuples in R with A value in given range

$\sigma_{A=val}(R) = ?$

Summary

- Estimating size of results is an “art”

- Don’t forget:

  Statistics must be kept up to date…

  (cost?)

Outline

- Estimating cost of query plan
  - Estimating size of results ← done!
  - Estimating # of IOs ← next…
- Generate and compare plans