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**Carnegie Mellon Univ.
Dept. of Computer Science
15-415 - Database Applications**

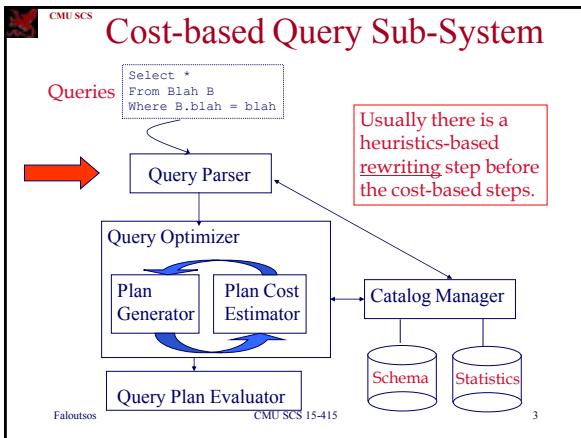
Query Optimization
(R&G ch. 15; Sys. R q-opt paper)

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Overview - detailed

- Why q-opt?
- Equivalence of expressions
- Cost estimation
- Plan generation
- Plan evaluation

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Why Q-opt?

- SQL: ~declarative
- good q-opt \rightarrow big difference
 - eg., seq. Scan vs
 - B-tree index, on P=1,000 pages

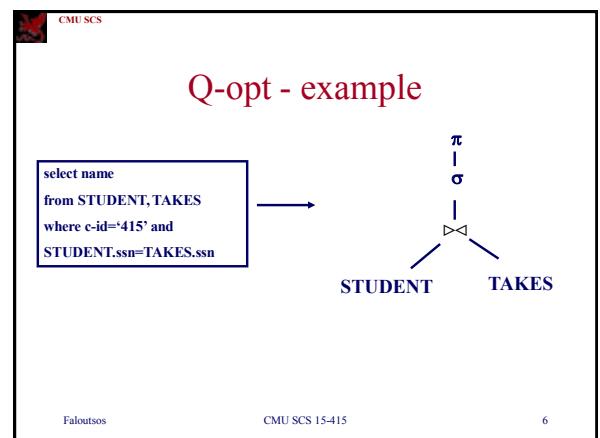
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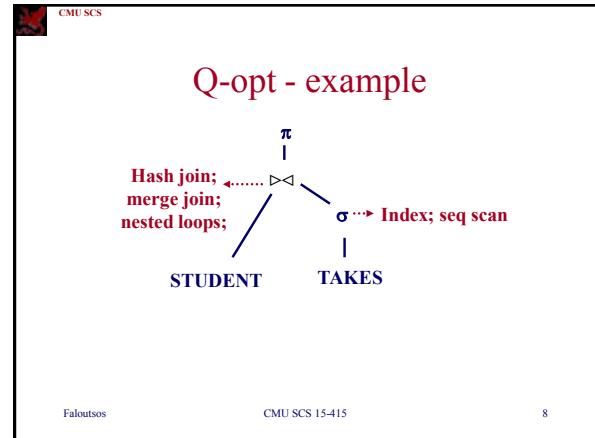
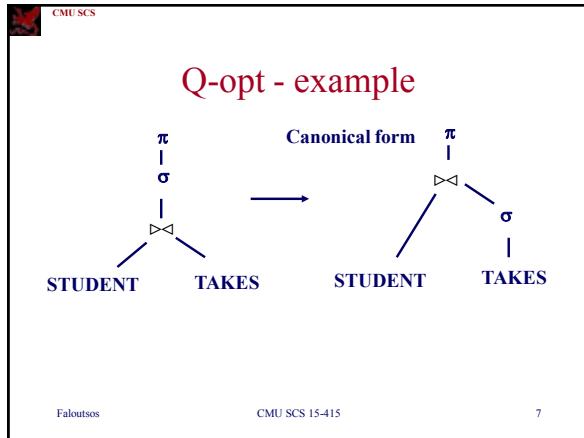
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Q-opt steps

- bring query in internal form (eg., parse tree)
- ... into ‘canonical form’ (syntactic q-opt)
- generate alt. plans
- estimate cost; pick best

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- ### Overview - detailed
- Why q-opt?
 - **Equivalence of expressions**
 - Cost estimation
 - ...
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- ### Equivalence of expressions
- A.k.a.: syntactic q-opt
 - in short: perform selections and projections early
 - More details: see transf. rules in text
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- ### Equivalence of expressions
- Q: How to prove a transf. rule?
 $\sigma_p(R1 \bowtie R2) = \sigma_p(R1) \bowtie \sigma_p(R2)$
 - A: use RTC, to show that LHS = RHS, eg:
 $\sigma_p(R1 \cup R2) = \sigma_p(R1) \cup \sigma_p(R2)$
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Equivalence of expressions

$$\sigma_p(R1 \cup R2) = \sigma_p(R1) \cup \sigma_p(R2)$$

$$t \in LHS \Leftrightarrow$$

$$t \in (R1 \cup R2) \wedge P(t) \Leftrightarrow$$

$$(t \in R1 \vee t \in R2) \wedge P(t) \Leftrightarrow$$

$$(t \in R1 \wedge P(t)) \vee (t \in R2 \wedge P(t)) \Leftrightarrow$$

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Equivalence of expressions

$$\sigma_p(R1 \cup R2) = ? \sigma_p(R1) \cup \sigma_p(R2)$$

...

$$(t \in R1 \wedge P(t)) \vee (t \in R2 \wedge P(t)) \Leftrightarrow$$

$$(t \in \sigma_p(R1)) \vee (t \in \sigma_p(R2)) \Leftrightarrow$$

$$t \in \sigma_p(R1) \cup \sigma_p(R2) \Leftrightarrow$$

$$t \in RHS$$

QED

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Equivalence of expressions

- Q: how to disprove a rule??

$$\pi_A(R1 - R2) = ? \pi_A(R1) - \pi_A(R2)$$

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Equivalence of expressions

- Selections
 - perform them early
 - break a complex predicate, and push

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

- simplify a complex predicate
 - ('X=Y and Y=3') -> 'X=3 and Y=3'

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Equivalence of expressions

- Projections
 - perform them early (but carefully...)
 - Smaller tuples
 - Fewer tuples (if duplicates are eliminated)
 - project out all attributes except the ones requested or required (e.g., joining attr.)

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Equivalence of expressions

- Joins
 - Commutative , associative

$$R \bowtie S = S \bowtie R$$

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

- Q: n-way join - how many diff. orderings?

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Equivalence of expressions

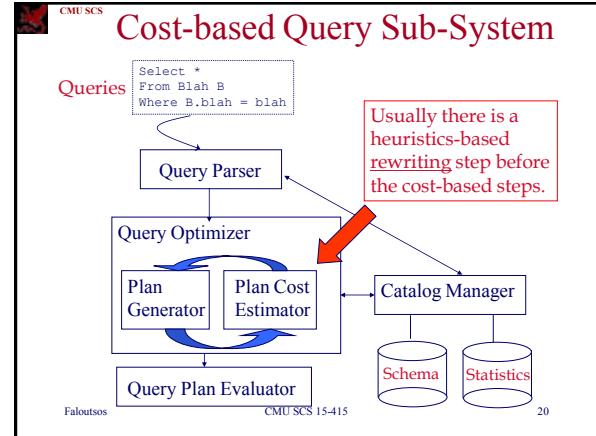
- Joins - Q: n-way join - how many diff. orderings?
- A: Catalan number $\sim 4^n$
 - Exhaustive enumeration: too slow.

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Q-opt steps

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- ... into ‘canonical form’ (syntactic q-opt)
- generate alt. plans
- **estimate cost**; pick best

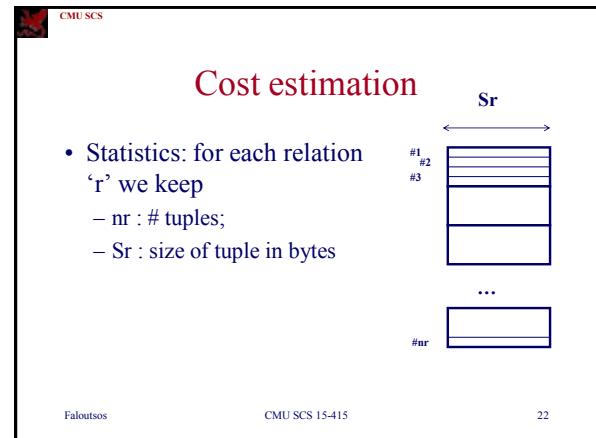
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Cost estimation

- Eg., find ssn’s of students with an ‘A’ in 415 (using seq. scanning)
- How long will a query take?
 - CPU (but: small cost; decreasing; tough to estimate)
 - Disk (mainly, # block transfers)
- How many tuples will qualify?
- (what statistics do we need to keep?)

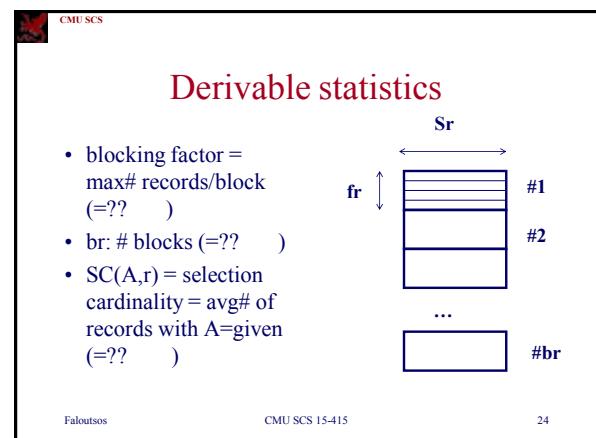
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Cost estimation

- Statistics: for each relation ‘r’ we keep
 - ...
 - $V(A,r)$: number of distinct values of attr. ‘A’
 - (recently, histograms, too)

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Derivable statistics

- blocking factor = max# records/block (= B/S_r ; B : block size in bytes)
- br: # blocks (= $nr / (\text{blocking-factor})$)

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Derivable statistics

- $SC(A,r) = \text{selection cardinality} = \text{avg\# of records with } A=\text{given}$ ($= nr / V(A,r)$)
(assumes uniformity...) – eg: 10,000 students, 10 colleges – how many students in SCS?

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Additional quantities we need:

- For index ‘i’:
 - f_i : average fanout (~50-100)
 - HT_i : # levels of index ‘i’ (~2-3)
 - $\sim \log(\#\text{entries})/\log(f_i)$
 - LBi : # blocks at leaf level

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Statistics

- Where do we store them?
- How often do we update them?

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Q-opt steps

- bring query in internal form (eg., parse tree)
- ... into ‘canonical form’ (syntactic q-opt)
- generate alt. plans
- **estimate cost**; pick best

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Selections

- we saw simple predicates ($A=\text{constant}$; eg., ‘name=Smith’)
- how about more complex predicates, like
 - ‘salary > 10K’
 - ‘age = 30 **and** job-code=“analyst”’
- what is their selectivity?

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Selections – complex predicates

- selectivity $\text{sel}(P)$ of predicate P :
 - fraction of tuples that qualify
$$\text{sel}(P) = \text{SC}(P) / \text{nr}$$

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Selections – complex predicates

- eg., assume that $V(\text{grade}, \text{TAKES})=5$ distinct values
- simple predicate P: $A=\text{constant}$
 - $\text{sel}(A=\text{constant}) = 1/V(A,r)$
 - eg., $\text{sel}(\text{grade}='B') = 1/5$
- (what if $V(A,r)$ is unknown??)

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Selections – complex predicates

- range query: $\text{sel}(\text{grade} \geq 'C')$
 - $\text{sel}(A > a) = (\text{Amax} - a) / (\text{Amax} - \text{Amin})$

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Selections - complex predicates

- negation: $\text{sel}(\text{grade} \neq 'C')$
 - $\text{sel}(\text{not } P) = 1 - \text{sel}(P)$
 - (Observation: selectivity \approx probability)

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Selections – complex predicates

conjunction:

- $\text{sel}(\text{grade} = 'C' \text{ and course} = '415')$
- $\text{sel}(P1 \text{ and } P2) = \text{sel}(P1) * \text{sel}(P2)$
- INDEPENDENCE ASSUMPTION

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Selections – complex predicates

disjunction:

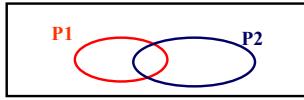
- $\text{sel}(\text{grade} = 'C' \text{ or course} = '415')$
- $\text{sel}(P1 \text{ or } P2) = \text{sel}(P1) + \text{sel}(P2) - \text{sel}(P1 \text{ and } P2)$
- $= \text{sel}(P1) + \text{sel}(P2) - \text{sel}(P1)*\text{sel}(P2)$
- INDEPENDENCE ASSUMPTION, again

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Selections – complex predicates

disjunction: in general
 $\text{sel}(P_1 \text{ or } P_2 \text{ or } \dots \text{ or } P_n) =$
 $1 - (1 - \text{sel}(P_1)) * (1 - \text{sel}(P_2)) * \dots * (1 - \text{sel}(P_n))$



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Selections – summary

- $\text{sel}(A=\text{constant}) = 1/V(A,r)$
- $\text{sel}(A>a) = (A_{\max} - a) / (A_{\max} - A_{\min})$
- $\text{sel}(\text{not } P) = 1 - \text{sel}(P)$
- $\text{sel}(P_1 \text{ and } P_2) = \text{sel}(P_1) * \text{sel}(P_2)$
- $\text{sel}(P_1 \text{ or } P_2) = \text{sel}(P_1) + \text{sel}(P_2) - \text{sel}(P_1)*\text{sel}(P_2)$
- $\text{sel}(P_1 \text{ or } \dots \text{ or } P_n) = 1 - (1-\text{sel}(P_1))*\dots*(1-\text{sel}(P_n))$

- UNIFORMITY and INDEPENDENCE ASSUMPTIONS

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Result Size Estimation for Joins

- Q: Given a join of R and S, what is the range of possible result sizes (in #of tuples)?
 - Hint: what if $R_{\text{cols}} \cap S_{\text{cols}} = \emptyset$?
 - $R_{\text{cols}} \cap S_{\text{cols}}$ is a key for R (and a Foreign Key in S)?

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Result Size Estimation for Joins

- General case: $R_{\text{cols}} \cap S_{\text{cols}} = \{A\}$ (and A is key for neither)
 - match each R-tuple with S-tuples

$$\text{est_size} \leq NTuples(R) * NTuples(S) / NKeys(A, S)$$

$$\leq nr * ns / V(A, S)$$
 - symmetrically, for S:

$$\text{est_size} \leq NTuples(R) * NTuples(S) / NKeys(A, R)$$

$$\leq nr * ns / V(A, R)$$
- Overall:

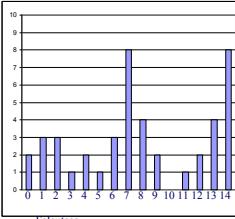
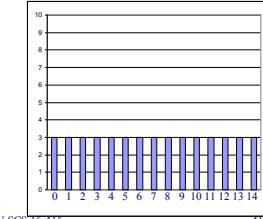
$$\text{est_size} = NTuples(R) * NTuples(S) / \max\{NKeys(A, S), NKeys(A, R)\}$$

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On the Uniform Distribution Assumption

- Assuming uniform distribution is rather crude

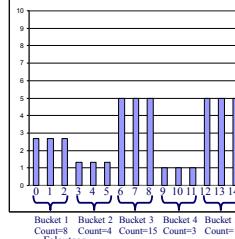
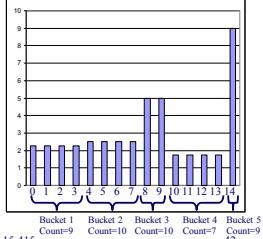
Distribution D	Uniform distribution approximating D
	

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Histograms

- For better estimation, use a *histogram*

Equiwidth histogram	Equidepth histogram ~ quantiles
	

Bucket 1 Count=8 Bucket 2 Count=4 Bucket 3 Count=15 Bucket 4 Count=3 Bucket 5 Count=15
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Bucket 1 Count=9 Bucket 2 Count=10 Bucket 3 Count=10 Bucket 4 Count=7 Bucket 5 Count=9
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Q-opt steps

- bring query in internal form (eg., parse tree)
- ... into ‘canonical form’ (syntactic q-opt)
- **generate alt. plans**
 - single relation
 - multiple relations
- estimate cost; pick best

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plan generation

- Selections – eg.,


```
select *
from TAKES
where grade = 'A'
```
- Plans?

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plan generation

- Plans?
 - seq. scan
 - binary search
 - (if sorted & consecutive)
 - index search
 - if an index exists

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plan generation

seq. scan – cost?

- br (worst case)
- br/2 (average, if we search for primary key)

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plan generation

binary search – cost?

if sorted and consecutive:

- $\sim \log(br) +$
- $SC(A,r)/fr$ (=blocks spanned by qual. tuples)

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plan generation

estimation of selection cardinalities $SC(A,r)$:

non-trivial – we saw it earlier

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plan generation

method#3: index – cost?
 – levels of index +
 – blocks w/ qual. tuples

case#1: primary key
 case#2: sec. key – clustering index
 case#3: sec. key – non-clust. index

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plan generation

method#3: index – cost?
 – levels of index +
 – blocks w/ qual. tuples

case#1: primary key – cost:
 $HTi + 1$

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plan generation

method#3: index – cost?
 – levels of index +
 – blocks w/ qual. tuples

case#2: sec. key – clustering index
 $HTi + SC(A,r)/fr$

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plan generation

method#3: index – cost?
 – levels of index +
 – blocks w/ qual. tuples

case#3: sec. key – non-clust. index
 $HTi + SC(A,r)$
 (actually, pessimistic...)

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plan generation

method#3: index – cost?
 – levels of index +
 – blocks w/ qual. tuples

(actually, pessimistic...)
 better estimates:
 Cardenas' formula

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Cardena's formula

- q: # qual records
- Q: # qual. blocks
- N: # records total
- B: # blocks total
- $Q = ??$

Alfonso Cardenas
UCLA 54

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Cardena's formula

- Pessimistic:
 - $- Q = q$

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Cardena's formula

- Pessimistic:
 - $- Q = q$
- More realistic
 - $- Q = q$ if $q \leq B$
 - $- Q = B$ otherwise

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Cardena's formula

- Cardenas' formula

$$Q = B [1 - (1 - 1/B)^q]$$

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Plans for single relation - summary

- no index: scan (dup-elim; sort)
- with index:
 - single index access path
 - multiple index access path
 - sorted index access path
 - index-only access path

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Citation

- P. G. Selinger, M. M. Astrahan, D. D. Chamberlin, R. A. Lorie, and T. G. Price. *Access path selection in a relational database management system*. In SIGMOD Conference, pages 23–34, 1979.

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Frequently cited database publications

<http://www.informatik.uni-trier.de/~ley/db/about/top.html>

#	Publication
608	Peter P. Chen: The Entity-Relationship Model - Toward a Unified View of Data. ACM Trans. Database Syst. 1(1): 9-36(1976)
580	E. F. Codd: A Relational Model of Data for Large Shared Data Banks. Commun. ACM 13(6): 377-387(1970)
371	Patricia G. Selinger, Morton M. Astrahan, Donald D. Chamberlin, Raymond A. Lorie, Thomas G. Price: Access Path Selection in a Relational Database Management System. SIGMOD Conference 1979: 23-34
366	Jeffrey D. Ullman: Principles of Database and Knowledge Base Systems, Volume I. Computer Science Press 1988, ISBN 0-7167-8158-1
...	...

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Statistics for Optimization

- NCARD (T) - cardinality of relation T in tuples
- TCARD (T) - number of pages containing tuples from T
- $P(T) = \text{TCARD}(T)/(\# \text{ of non-empty pages in the segment})$
 - If segments only held tuples from one relation there would be no need for $P(T)$
- ICARD(I) - number of distinct keys in index I
- NIDX(I) - number of pages in index I

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Predicate Selectivity Estimation

attr = value	$F = 1/\text{ICARD}(\text{attr index})$ – if index exists $F = 1/10$ otherwise
attr1 = attr2	$F = 1/\max(\text{ICARD}(I1), \text{ICARD}(I2))$ or $F = 1/\text{ICARD}(Ii)$ – if only index i exists, or $F = 1/10$
val1 < attr < val2	$F = (\text{value2}-\text{value1})/(\text{high key}-\text{low key})$ $F = 1/4$ otherwise
expr1 or expr2	$F = F(\text{expr1}) + F(\text{expr2}) - F(\text{expr1}) * F(\text{expr2})$
expr1 and expr2	$F = F(\text{expr1}) * F(\text{expr2})$
NOT expr	$F = 1 - F(\text{expr})$

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Costs per Access Path Case

Unique index matching equal predicate	$1+1+W$
Clustered index I matching ≥ 1 preds	$F(\text{preds}) * (\text{NIDX}(I) + \text{TCARD}) + W * \text{RSICARD}$
Non-clustered index I matching ≥ 1 preds	$F(\text{preds}) * (\text{NIDX}(I) + \text{NCARD}) + W * \text{RSICARD}$
Segment scan	$\text{TCARD}/P + W * \text{RSICARD}$

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Q-opt steps

- bring query in internal form (eg., parse tree)
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- generate alt. plans
 - single relation
 - **multiple relations**
 - Main idea
 - Dynamic programming – reminder
 - Example
- estimate cost; pick best

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n-way joins

- $r1 \text{ JOIN } r2 \text{ JOIN } \dots \text{ JOIN } rn$
- typically, break problem into 2-way joins
 - choose between NL, sort merge, hash join, ...

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Queries Over Multiple Relations

- As number of joins increases, number of alternative plans grows rapidly → *need to restrict search space*
- Fundamental decision in System R: only left-deep join trees are considered. Advantages?

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Queries Over Multiple Relations

- As number of joins increases, number of alternative plans grows rapidly → *need to restrict search space*
- Fundamental decision in System R: *only left-deep join trees* are considered. Advantages?
 - fully pipelined* plans.
 - Intermediate results not written to temporary files.
 - Not all left-deep trees are fully pipelined (e.g., SM join).

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Queries over Multiple Relations

- Enumerate the orderings (= left deep tree)
- enumerate the plans for each operator
- enumerate the access paths for each table

Dynamic programming, to save cost estimations

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Q-opt steps

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(Reminder: Dynamic Programming)

Flight network diagram showing cities PIT, BOS, ATL, JFK, CDG, FRA, and SG. Flight costs are labeled on the edges:

- PIT to BOS: \$200
- PIT to ATL: \$150
- BOS to CDG: \$500
- ATL to FRA: \$650
- JFK to FRA: \$850
- CDG to SG: \$800
- FRA to SG: \$450
- ATL to SG: \$950

Cheapest flight PIT → SG ?

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(Reminder: Dynamic Programming)

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- ATL to SG: \$950

Assumption: NO package deals: cost CDG->SG is always \$800, no matter how reached CDG

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(Reminder: Dynamic Programming)

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- CDG to SG: \$800
- FRA to SG: \$450
- ATL to SG: \$950

Solution: compute partial optimal, left-to-right:

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(Reminder: Dynamic Programming)

```

graph LR
    PIT((PIT)) -- $200 --> BOS((BOS))
    PIT -- $150 --> HKF((HKF))
    PIT -- $50 --> ATL((ATL))
    BOS -- $200 --> HKF
    BOS -- $500 --> CGC((CGC))
    BOS -- $1050 --> ATL
    HKF -- $150 --> ATL
    HKF -- $850 --> FRA((FRA))
    ATL -- $50 --> FRA
    ATL -- $650 --> FRA
    FRA -- $450 --> CGC
    FRA -- $650 --> SG((SG))
    CGC -- $800 --> SG
    
```

Solution: compute partial optimal, left-to-right:

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(Reminder: Dynamic Programming)

```

graph LR
    PIT((PIT)) -- $200 --> BOS((BOS))
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    BOS -- $200 --> HKF
    BOS -- $500 --> CGC((CGC))
    BOS -- $1050 --> ATL
    HKF -- $150 --> ATL
    HKF -- $850 --> FRA((FRA))
    ATL -- $50 --> FRA
    ATL -- $650 --> FRA
    FRA -- $450 --> CGC
    FRA -- $650 --> SG((SG))
    FRA -- $950 --> SG
    CGC -- $800 --> SG
    
```

Solution: compute partial optimal, left-to-right:

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(Reminder: Dynamic Programming)

```

graph LR
    PIT((PIT)) -- $200 --> BOS((BOS))
    PIT -- $150 --> HKF((HKF))
    PIT -- $50 --> ATL((ATL))
    BOS -- $200 --> HKF
    BOS -- $500 --> CGC((CGC))
    BOS -- $1050 --> ATL
    HKF -- $150 --> ATL
    HKF -- $850 --> FRA((FRA))
    ATL -- $50 --> FRA
    ATL -- $650 --> FRA
    FRA -- $450 --> CGC
    FRA -- $650 --> SG((SG))
    FRA -- $1500 --> SG
    CGC -- $800 --> SG
    
```

Solution: compute partial optimal, left-to-right:

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(Reminder: Dynamic Programming)

```

graph LR
    PIT((PIT)) -- $200 --> BOS((BOS))
    PIT -- $150 --> HKF((HKF))
    PIT -- $50 --> ATL((ATL))
    BOS -- $200 --> HKF
    BOS -- $500 --> CGC((CGC))
    BOS -- $1050 --> ATL
    HKF -- $150 --> ATL
    HKF -- $850 --> FRA((FRA))
    ATL -- $50 --> FRA
    ATL -- $650 --> FRA
    FRA -- $450 --> CGC
    FRA -- $650 --> SG((SG))
    FRA -- $1500 --> SG
    CGC -- $800 --> SG
    
```

So, best price is \$1,500 – which legs?

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(Reminder: Dynamic Programming)

```

graph LR
    PIT((PIT)) -- $200 --> BOS((BOS))
    PIT -- $150 --> HKF((HKF))
    PIT -- $50 --> ATL((ATL))
    BOS -- $200 --> HKF
    BOS -- $500 --> CGC((CGC))
    BOS -- $1050 --> ATL
    HKF -- $150 --> ATL
    HKF -- $850 --> FRA((FRA))
    ATL -- $50 --> FRA
    ATL -- $650 --> FRA
    FRA -- $450 --> CGC
    FRA -- $650 --> SG((SG))
    FRA -- $1500 --> SG
    CGC -- $800 --> SG
    
```

So, best price is \$1,500 – which legs?

A: follow the winning edges, backwards

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(Reminder: Dynamic Programming)

```

graph LR
    PIT((PIT)) -- $200 --> BOS((BOS))
    PIT -- $150 --> HKF((HKF))
    PIT -- $50 --> ATL((ATL))
    BOS -- $200 --> HKF
    BOS -- $500 --> CGC((CGC))
    BOS -- $1050 --> ATL
    HKF -- $150 --> ATL
    HKF -- $850 --> FRA((FRA))
    ATL -- $50 --> FRA
    ATL -- $650 --> FRA
    FRA -- $450 --> CGC
    FRA -- $650 --> SG((SG))
    FRA -- $1500 --> SG
    CGC -- $800 --> SG
    
```

So, best price is \$1,500 – which legs?

A: follow the winning edges, backwards

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(Reminder: Dynamic Programming)

So, best price is \$1,500 – which legs?

A: follow the winning edges, backwards

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(Reminder: Dynamic Programming)

Q: what are the states, costs and arrows, in q-opt?

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(Reminder: Dynamic Programming)

Q: what are the states (and costs and arrows), in q-opt?

A: set of intermediate result tables

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Q-opt and Dyn. Programming

- E.g., compute $R \text{ join } S \text{ join } T$

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Q-opt and Dyn. Programming

- Details: how to record the fact that, say R is sorted on $R.a$? or that the user requires sorted output?
- A:
- E.g., consider the query

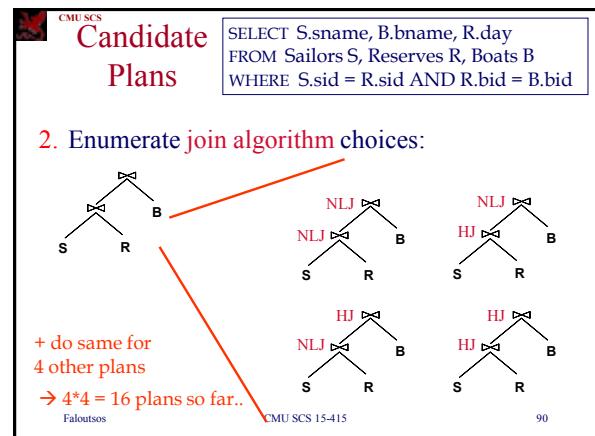
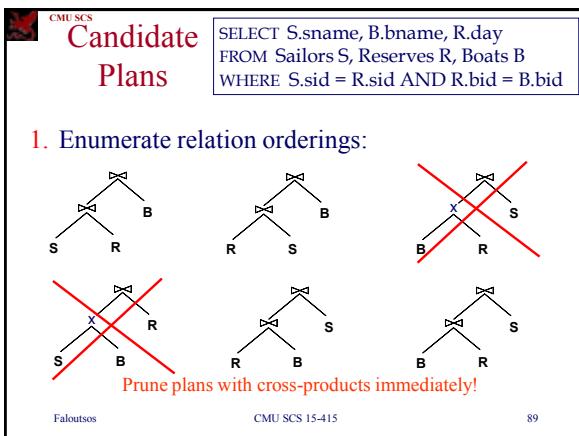
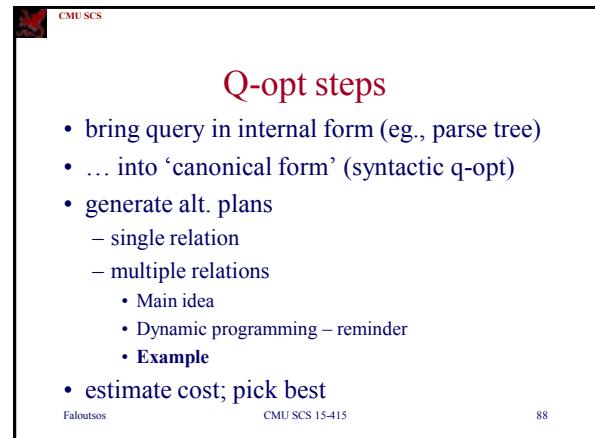
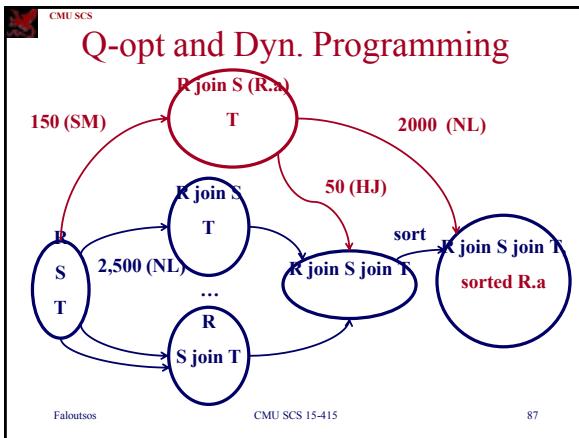
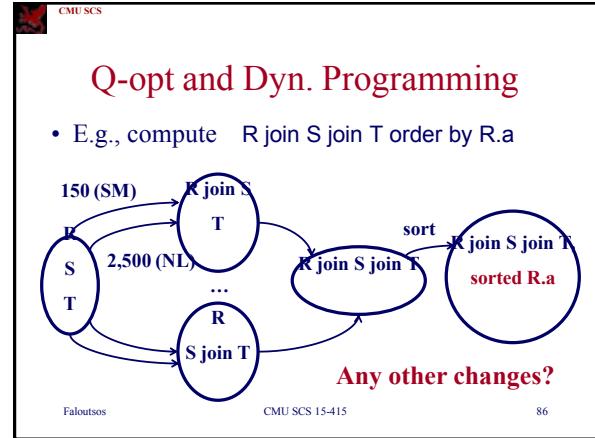
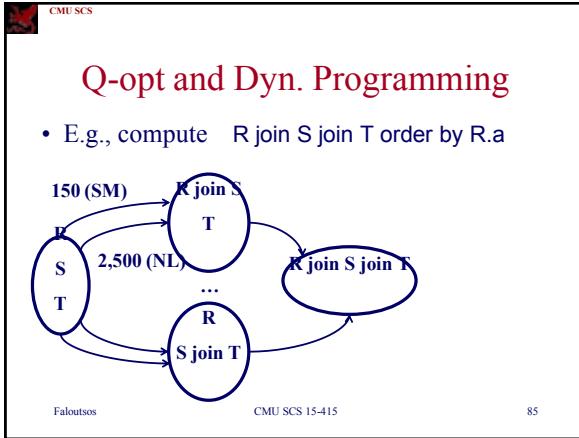

```
select *
from R, S, T
where R.a = S.a and S.b = T.b
order by R.a
```

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Q-opt and Dyn. Programming

- Details: how to record the fact that, say R is sorted on $R.a$? or that the user requires sorted output?
- A: record orderings, in the state
- E.g., consider the query


```
select *
from R, S, T
where R.a = S.a and S.b = T.b
order by R.a
```



Candidate Plans

```
SELECT S.sname, B.bname, R.day
FROM Sailors S, Reserves R, Boats B
WHERE S.sid = R.sid AND R.bid = B.bid
```

3. Enumerate access method choices:

+ do same for other plans

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Now estimate the cost of each plan

Example:

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Q-opt steps

- bring query in internal form (eg., parse tree)
- ... into ‘canonical form’ (syntactic q-opt)
- generate alt. plans
 - single relation
 - multiple relations
 - **nested subqueries**
- estimate cost; pick best

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Q-opt steps

- Everything so far: about a single query block

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Query Rewriting

- Re-write nested queries
- to: **de-correlate** and/or **flatten** them

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Example: Decorrelating a Query

```
SELECT S.sid
FROM Sailors S
WHERE EXISTS
(SELECT *
FROM Reserves R
WHERE R.bid=103
AND R.sid=S.sid)
```

Equivalent uncorrelated query:
SELECT S.sid
FROM Sailors S
WHERE S.sid IN
(SELECT R.sid
FROM Reserves R
WHERE R.bid=103)

- **Advantage:** nested block only needs to be executed **once** (rather than once per S tuple)

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Example: “Flattening” a Query

```
SELECT S.sid
FROM Sailors S
WHERE S.sid IN
  (SELECT R.sid
   FROM Reserves R
   WHERE R.bid=103)
```

Equivalent non-nested query:

```
SELECT S.sid
FROM Sailors S, Reserves R
WHERE S.sid=R.sid
AND R.bid=103
```

- **Advantage:** can use a join algorithm + optimizer can select among join algorithms & reorder freely

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Structure of query optimizers:

System R:

- break query in query blocks
- simple queries (ie., no joins): look at stats
- n-way joins: left-deep join trees; ie., only one intermediate result at a time
 - pros: smaller search space; pipelining
 - cons: may miss optimal
- 2-way joins: NL and sort-merge

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Structure of query optimizers:

More heuristics by Oracle, Sybase and Starburst (-> DB2)

In general: q-opt is very important for large databases.

(‘explain select <sql-statement>’ gives plan)

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Q-opt steps

- bring query in internal form (eg., parse tree)
- ... into ‘canonical form’ (syntactic q-opt)
- generate alt. plans
- estimate cost; pick best

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Conclusions

- Ideas to remember:
 - syntactic q-opt – do selections early
 - selectivity estimations (uniformity, indep.; histograms; join selectivity)
 - hash join (nested loops; sort-merge)
 - left-deep joins
 - dynamic programming

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