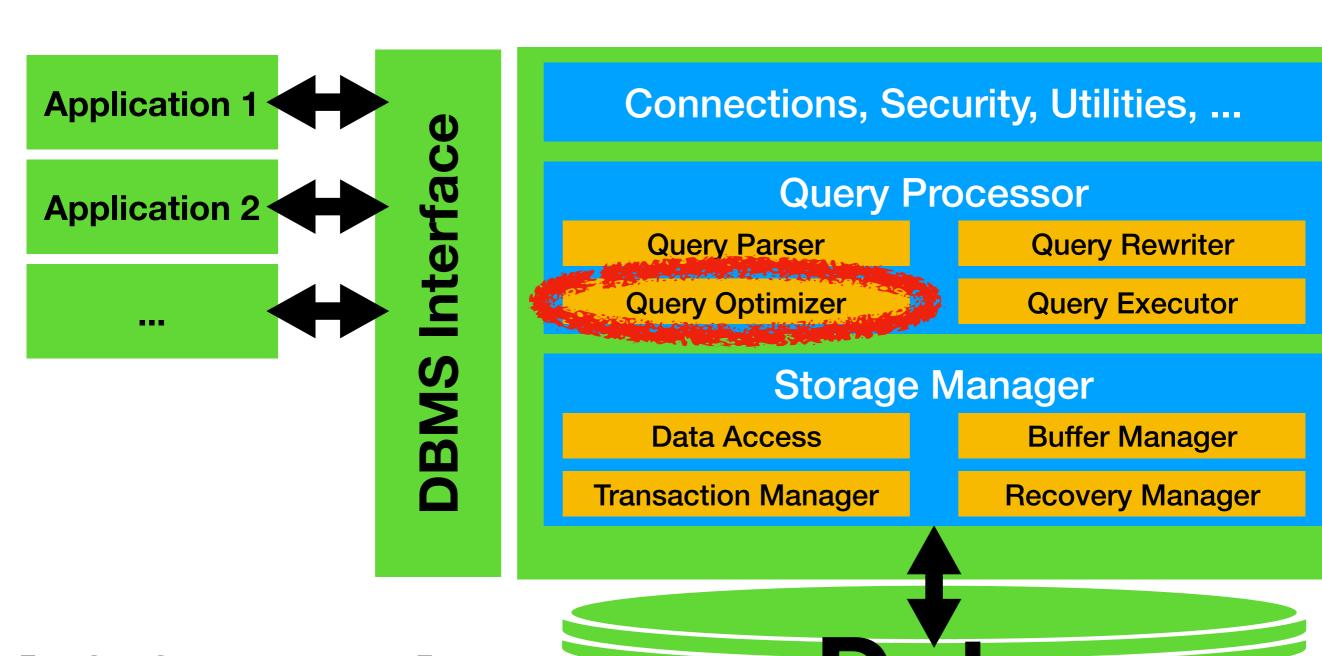
#### Query Optimization

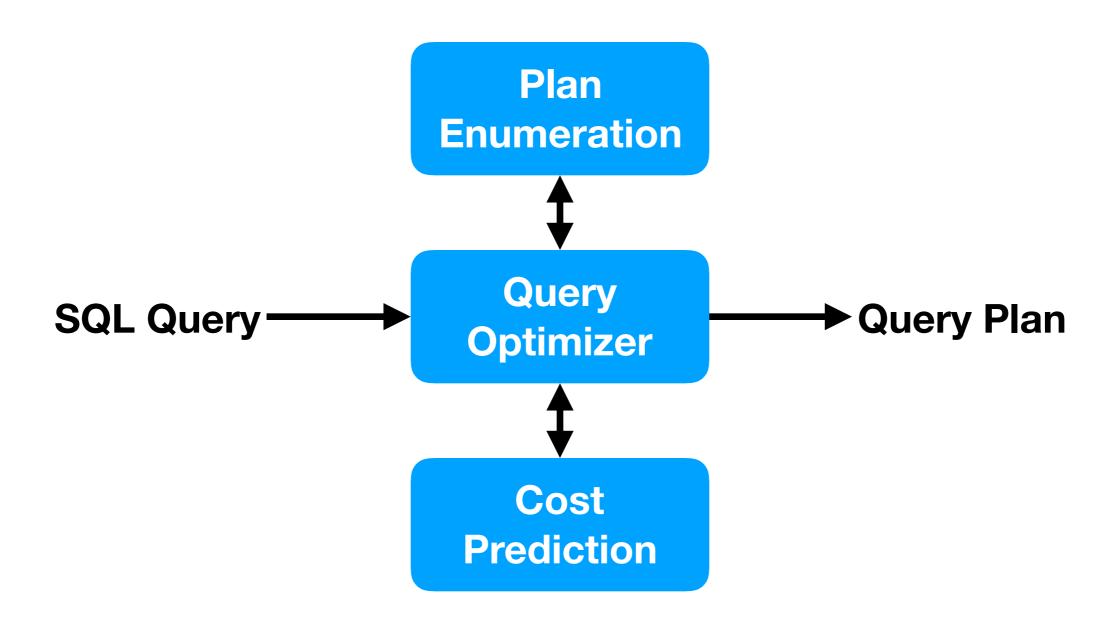
Immanuel Trummer itrummer@cornell.edu www.itrummer.org

# Database Management Systems (DBMS)

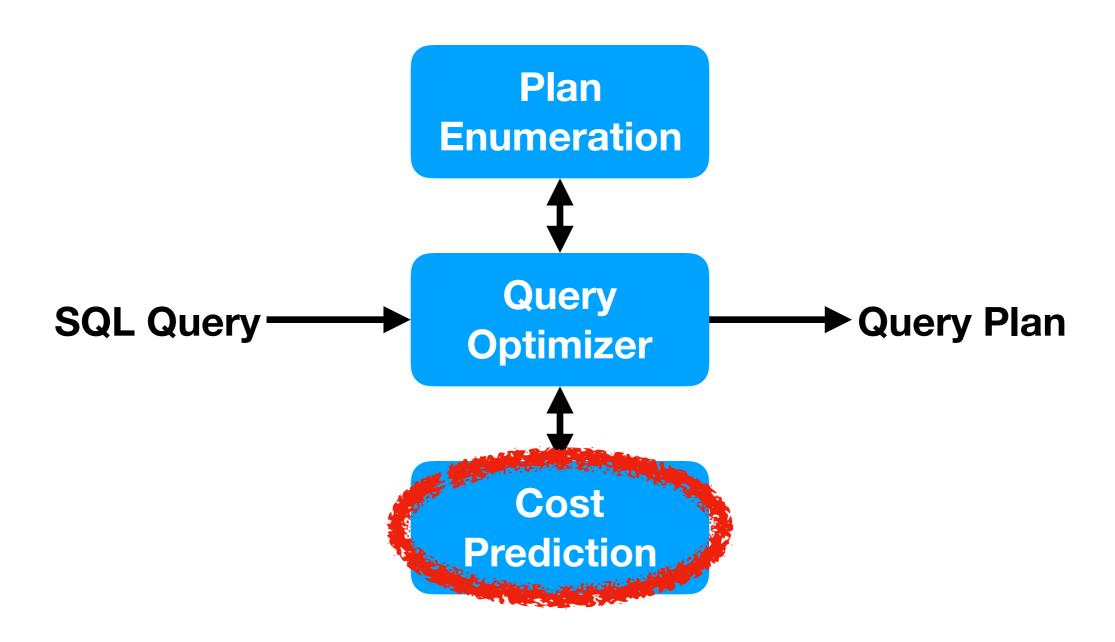


[RG, Sec. 13, 14]

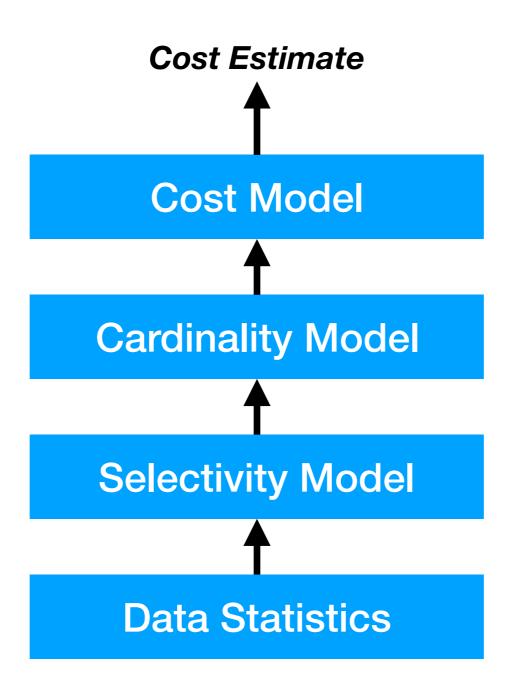
#### Optimizer Overview



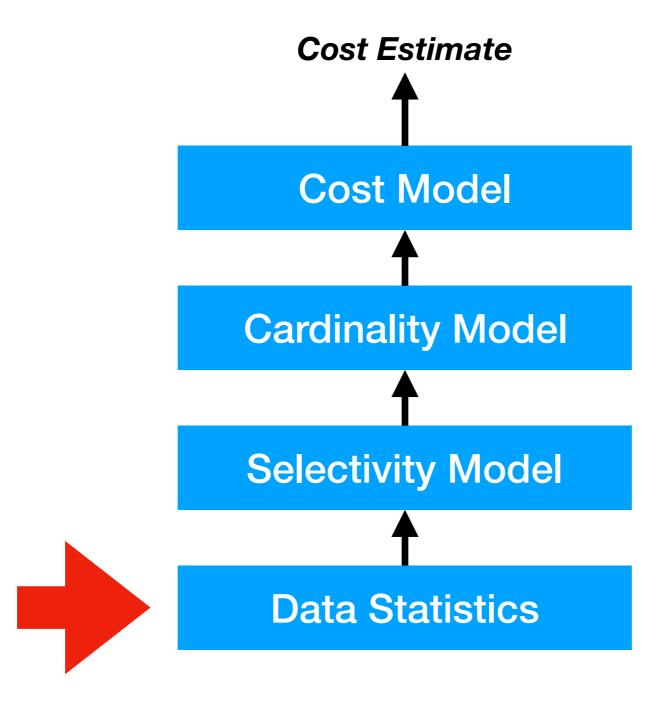
#### Optimizer Overview



#### Cost Model Structure



#### Cost Model Structure



#### Data Statistics

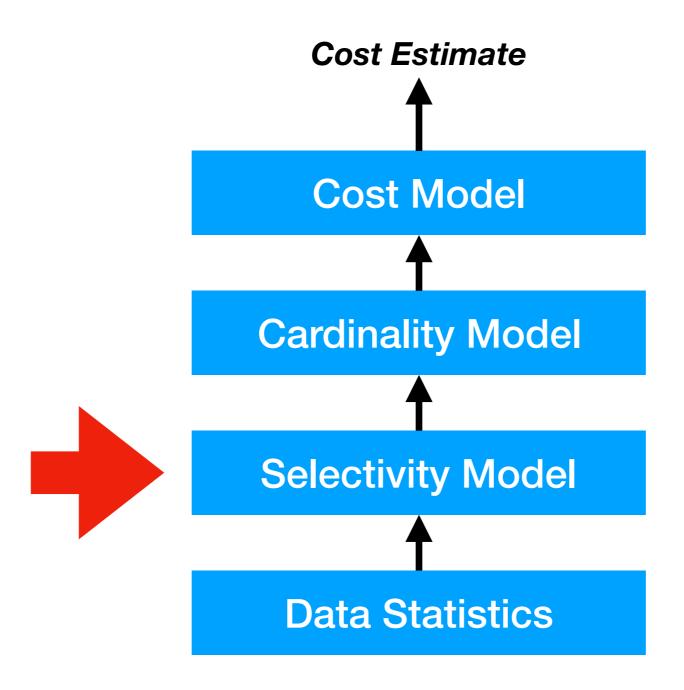
- Different DBMS collect different statistics
- Example statistics (collected in Postgres):
  - Number of distinct values in column
  - Ratio of SQL null values in column
  - Most frequent values with associated frequency
  - Histograms approximating column data distribution

• ...

#### Data Statistics in PG

- Force DBMS to create statistics via VACUUM ANALYZE
- Generated statistics are exploited by query optimizer
  - Inexplicably bad performance? Might miss statistics
- Can access column statistics via pg\_stats view
  - E.g., tablename, attname, n\_distinct, null\_frac, ...

#### Cost Model Structure



# Estimating Selectivity

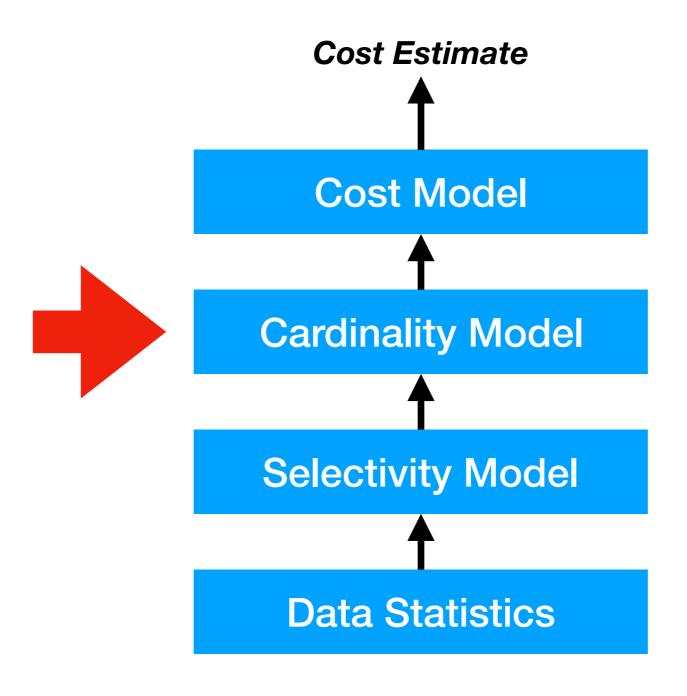
- Selectivity: probability that row satisfies predicate
- Estimate via constraints and data statistics
- E.g., Selectivity(Column=Value) = 1/NrValues
- E.g., NrRows(Column=Value) ≤ 1 if Key Column
- Selectivity(A \wedge B) = Selectivity(A) \* Selectivity(B)

# Which Simplifying Assumptions Do We Make?

# Estimating Selectivity

- Selectivity: probability that row satisfies predicate
- Estimate via constraints and data statistics
- E.g., Selectivity(Column=Value) = 1/NrValues
   Assumes Uniform Data
- E.g., NrRows(Column=Value) ≤ 1 if Key Column
- Selectivity(A \ B) = Selectivity(A) \* Selectivity(B)
   Assumes Independent Predicates

#### Cost Model Structure



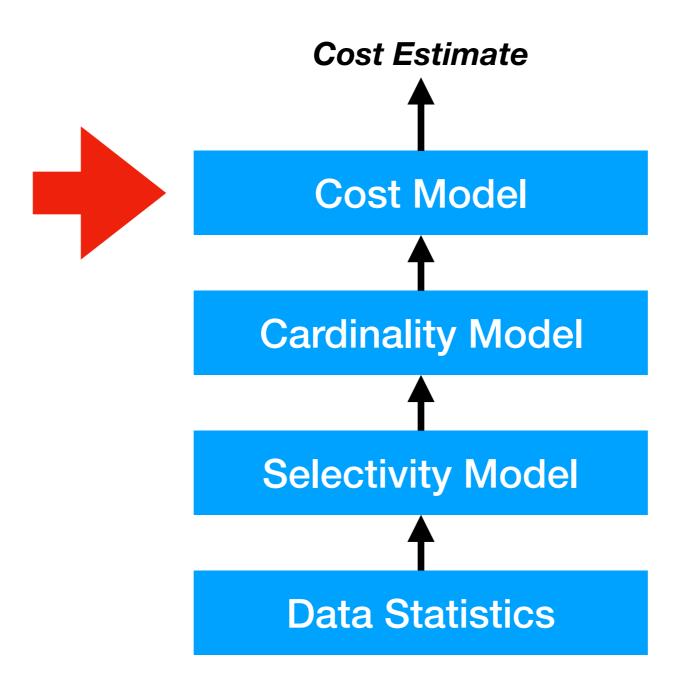
#### **Estimating Cardinality**

- Assume that cardinality of single tables is given
  - DBMS store that information for each base table
- Calculate cardinality product for tables in from clause
  - Number of result rows if predicates are always true
- Now multiply with selectivity estimates for all predicates
  - Estimate how many rows pass predicate filter

# Estimating Page Size

- Cost functions are based on number of data pages
- Calculate average byte size per record for each result
- Calculate how many records fit on one data page
  - Pages cannot store fractional records → round down
- Divide number of rows by number of records per page
  - Result is size, measured in pages

#### Cost Model Structure



#### Estimating Cost

- Generally only count cost of page reads and writes
- Sum up cost over all operators in the plan
- For each operator consider the following:
  - For each input, how often is it read from disk?
  - For each output, is it written to disk? Is it final result?
  - Does the operator read/write intermediate results?

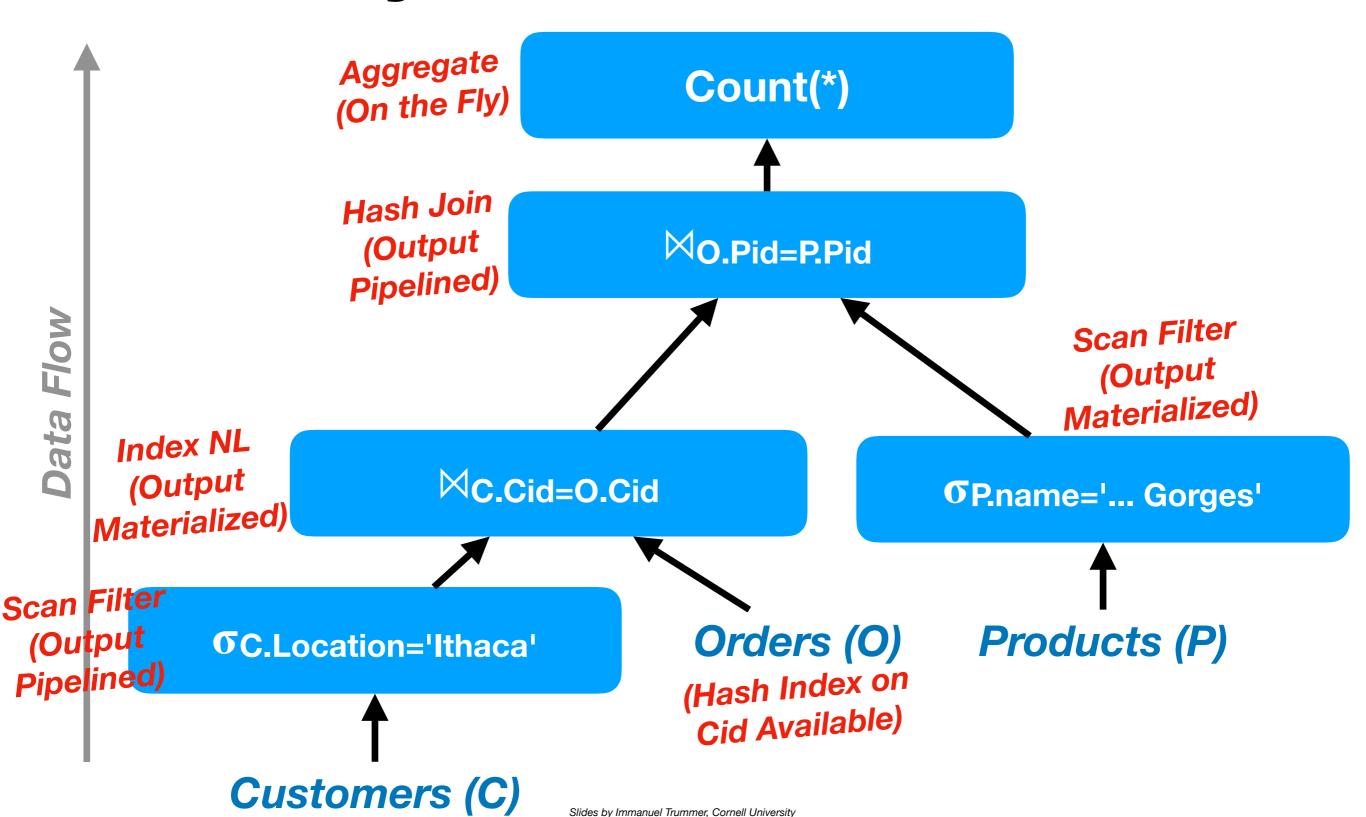
# Cost Estimation Example

#### Input Query and Database

```
SELECT Count(*)
FROM Customers C JOIN Orders O USING (cid)
JOIN Products P USING (pid)
WHERE C.location = 'Ithaca' AND
P.name = 'Book Ithaca is Gorges'
```

Customers(Cid, name, location)
Orders(Cid, Pid, date)
Products(Pid, name, price)

#### Query Plan Candidate



# Selectivity Estimation

Element	Property	Value
Customers	Cardinality	10,000
Orders	Cardinality	100,000
Products	Cardinality	5,000
<b>Customers.Location</b>	# Values	100
Products.Name	# Values	2,500

Foreign Key: Orders.Cid to Customers.Cid Foreign Key: Orders.Pid to Products.Pid

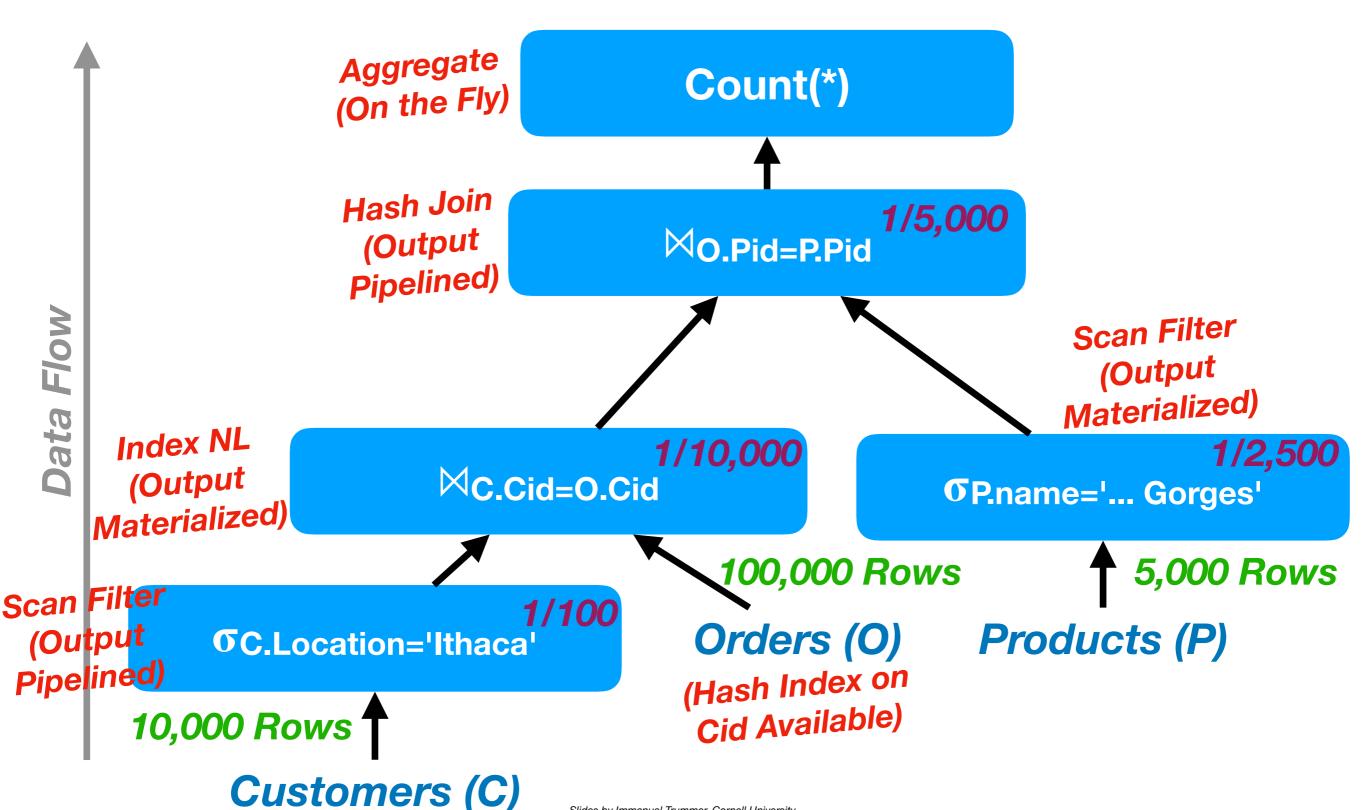
Predicate	Selectivity
σc.Location='Ithaca'	?
C.Cid=O.Cid	?
σP.name=' Gorges'	?
O.Pid=P.Pid	ummer, Corneir Oniversity

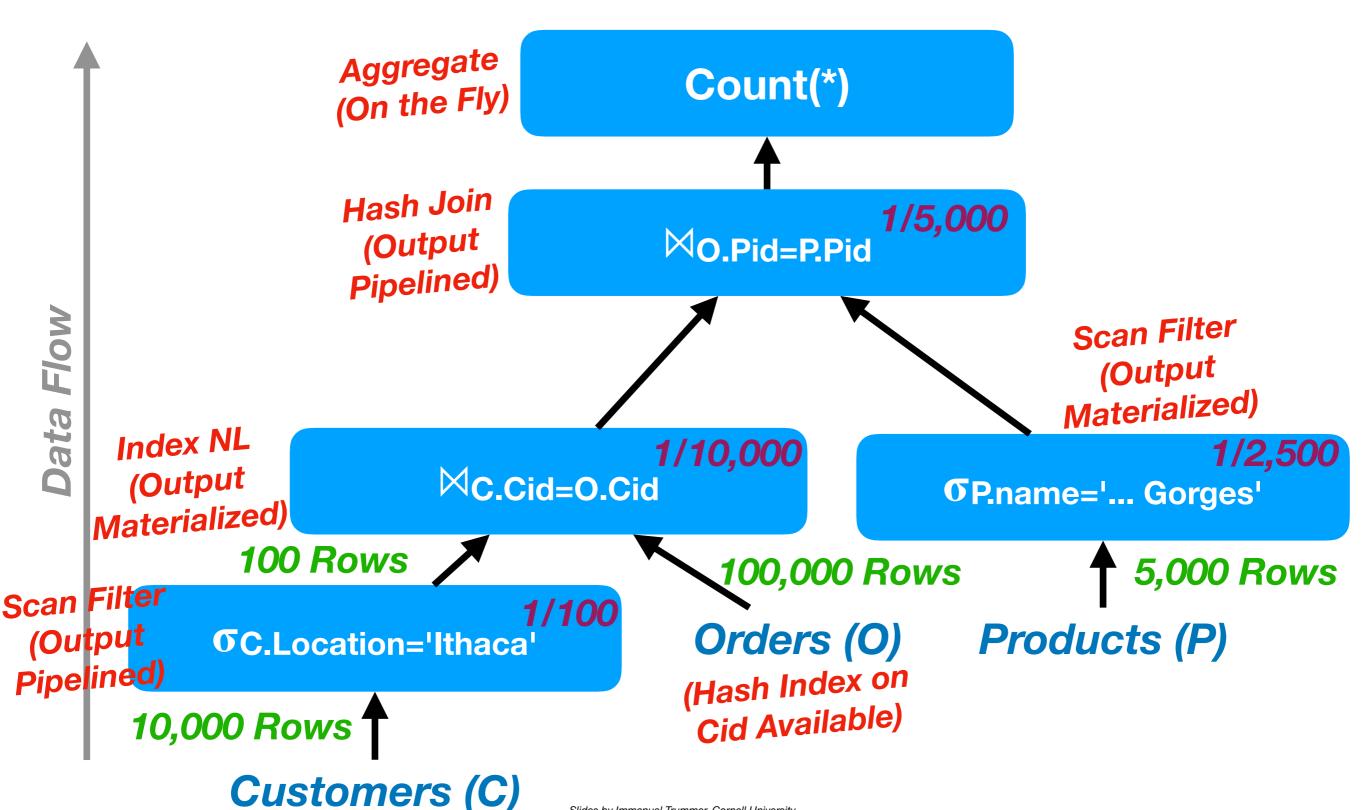
# Selectivity Estimation

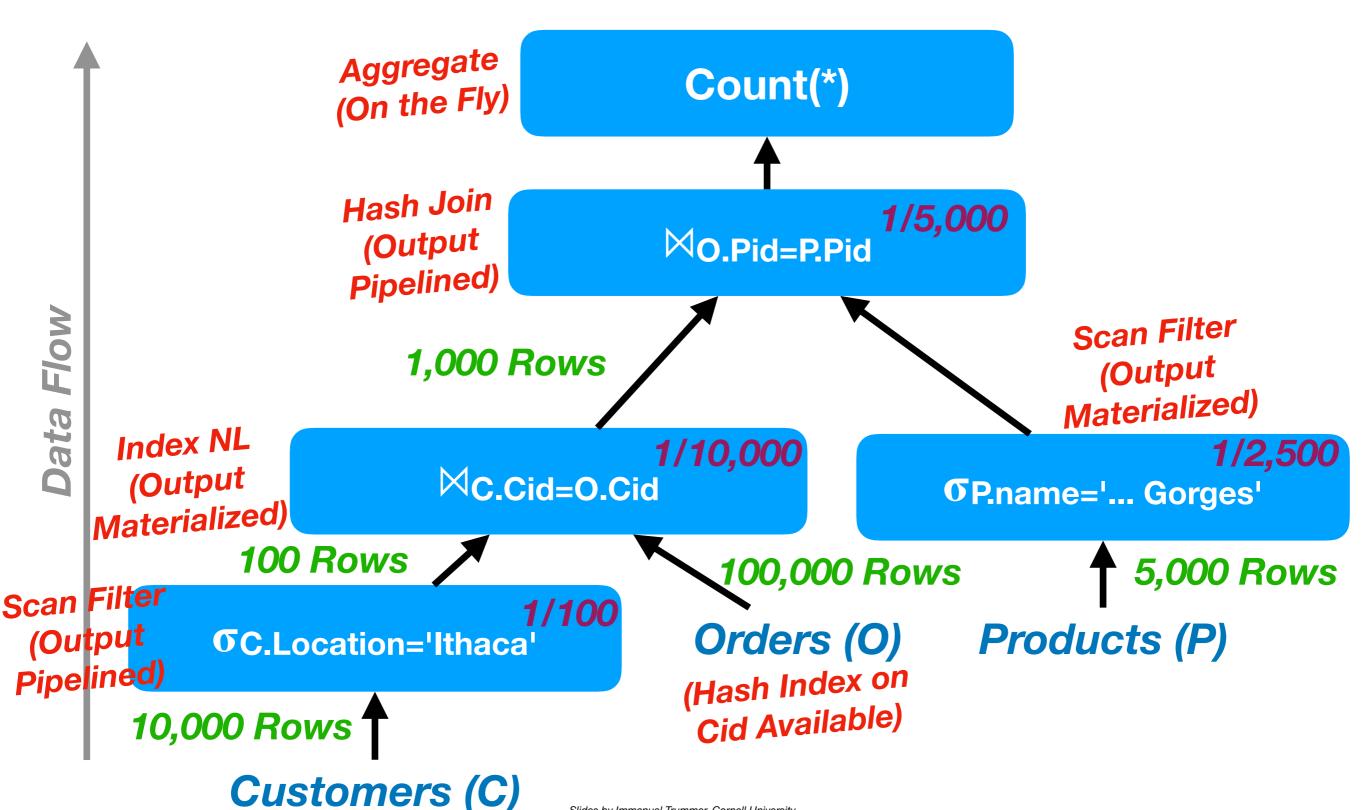
Element	Property	Value
Customers	Cardinality	10,000
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<b>Customers.Location</b>	# Values	100
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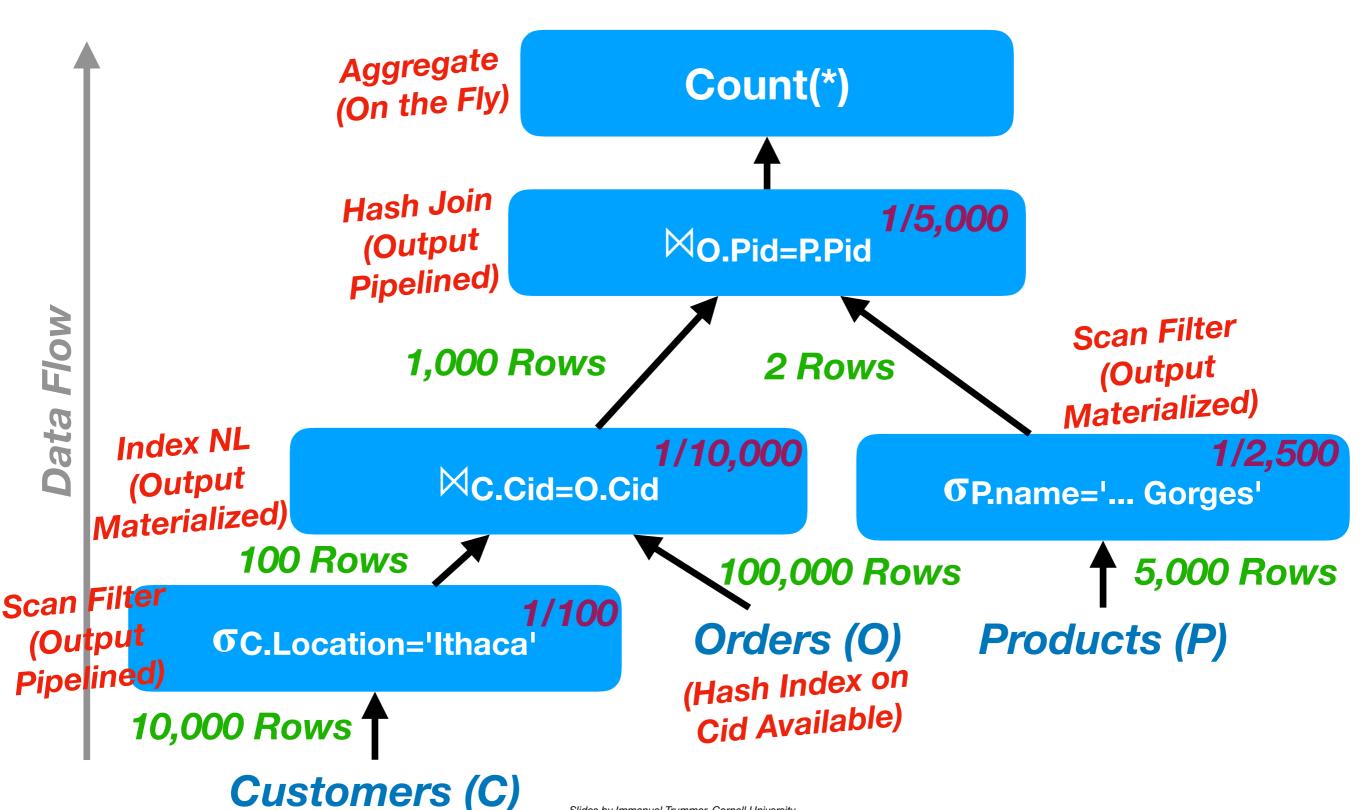
Foreign Key: Orders.Cid to Customers.Cid Foreign Key: Orders.Pid to Products.Pid

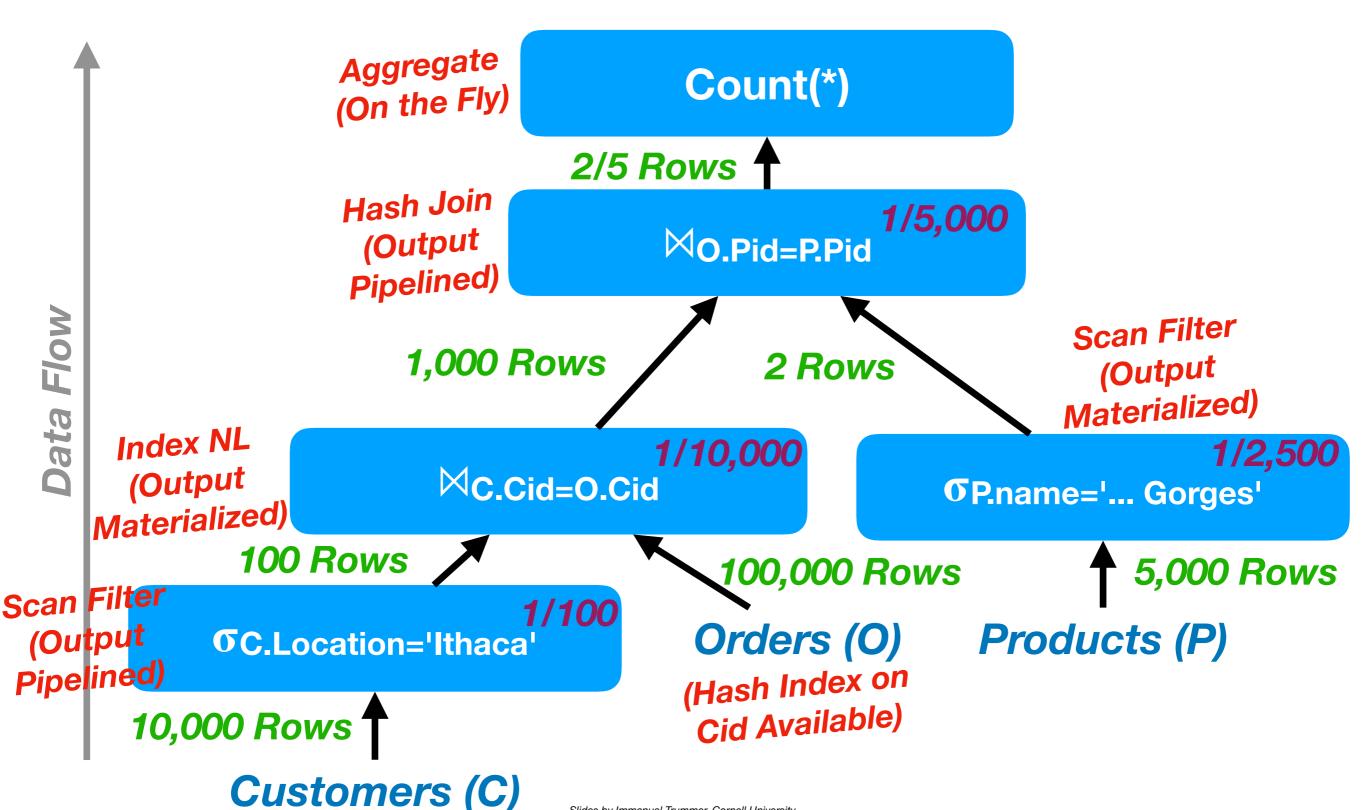
Predicate	Selectivity
σc.Location='Ithaca'	1/100
C.Cid=O.Cid	1/10,000
σP.name=' Gorges'	1/2,500
O.Pid=P.Pid	1/5,000











#### Is That Realistic ...?

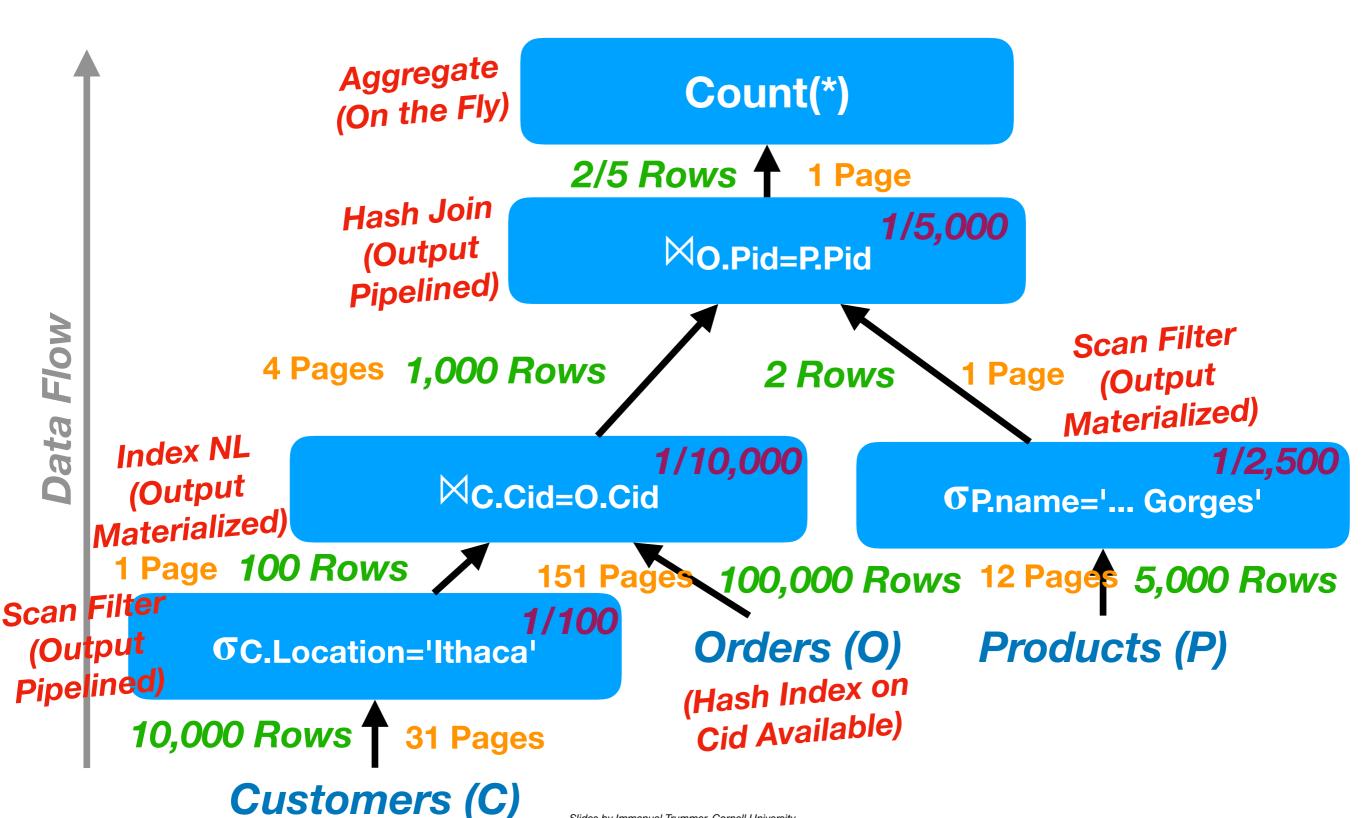
Table	Columns	Size/Entry	Rows/Page
Customers	Cid, name, location	?	?
Orders	Cid, Pid, date	?	?
Products	Pid, name, price	?	?
CMO	Cid, name, location, pid, date	?	?
CMOMP	Cid, C.name, location, pid, date, P.name, price	?	?

Table	Columns	Size/Entry	Rows/Page
Customers	Cid, name, location	24	?
Orders	Cid, Pid, date	?	?
Products	Pid, name, price	?	?
CMO	Cid, name, location, pid, date	?	?
CMOMP	Cid, C.name, location, pid, date, P.name, price	?	?

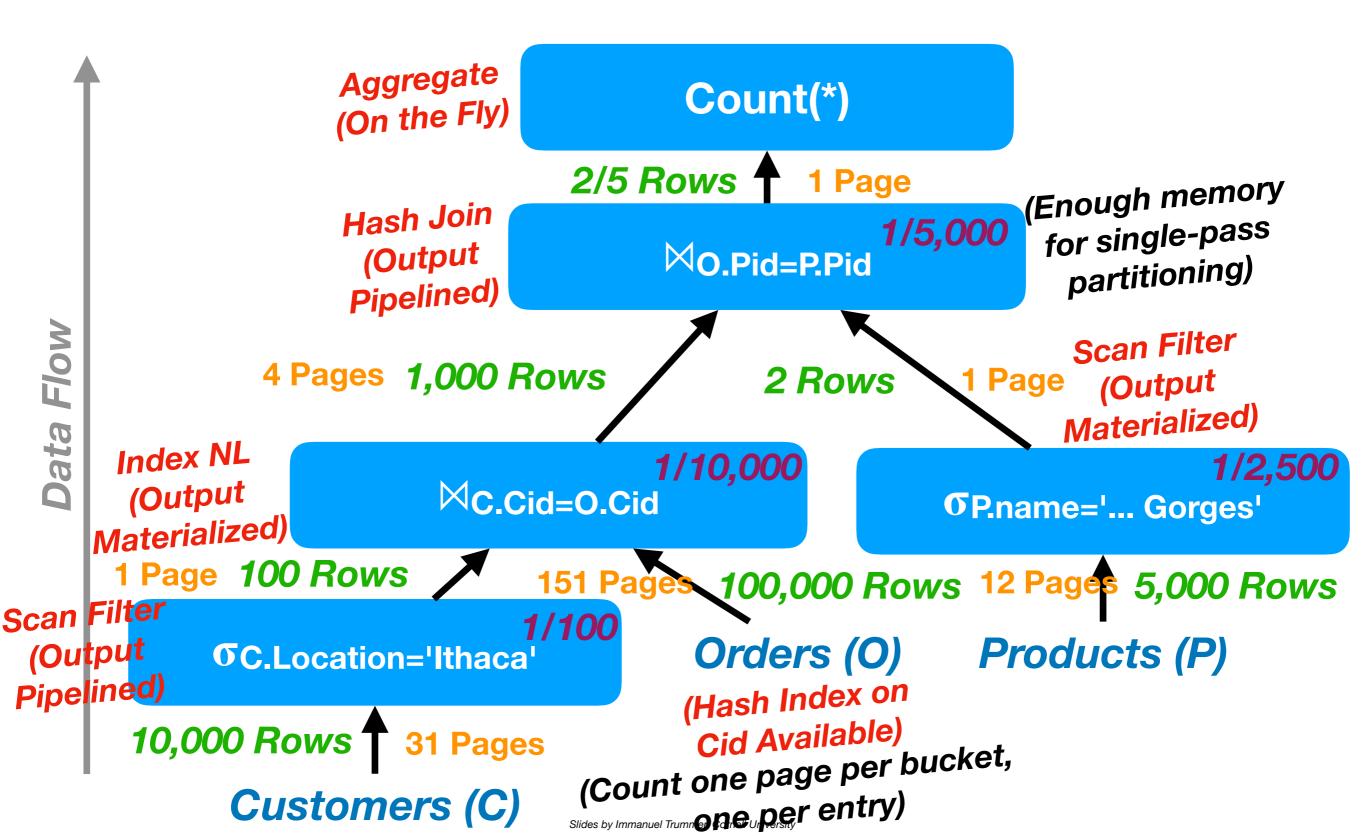
Table	Columns	Size/Entry	Rows/Page
Customers	Cid, name, location	24	333
Orders	Cid, Pid, date	?	?
Products	Pid, name, price	?	?
CMO	Cid, name, location, pid, date	?	?
CMOMP	Cid, C.name, location, pid, date, P.name, price	?	?

Table	Columns	Size/Entry	Rows/Page
Customers	Cid, name, location	24	333
Orders	Cid, Pid, date	12	666
Products	Pid, name, price	18	444
CMO	Cid, name, location, pid, date	32	250
CMOMP	Cid, C.name, location, pid, date, P.name, price	46	173

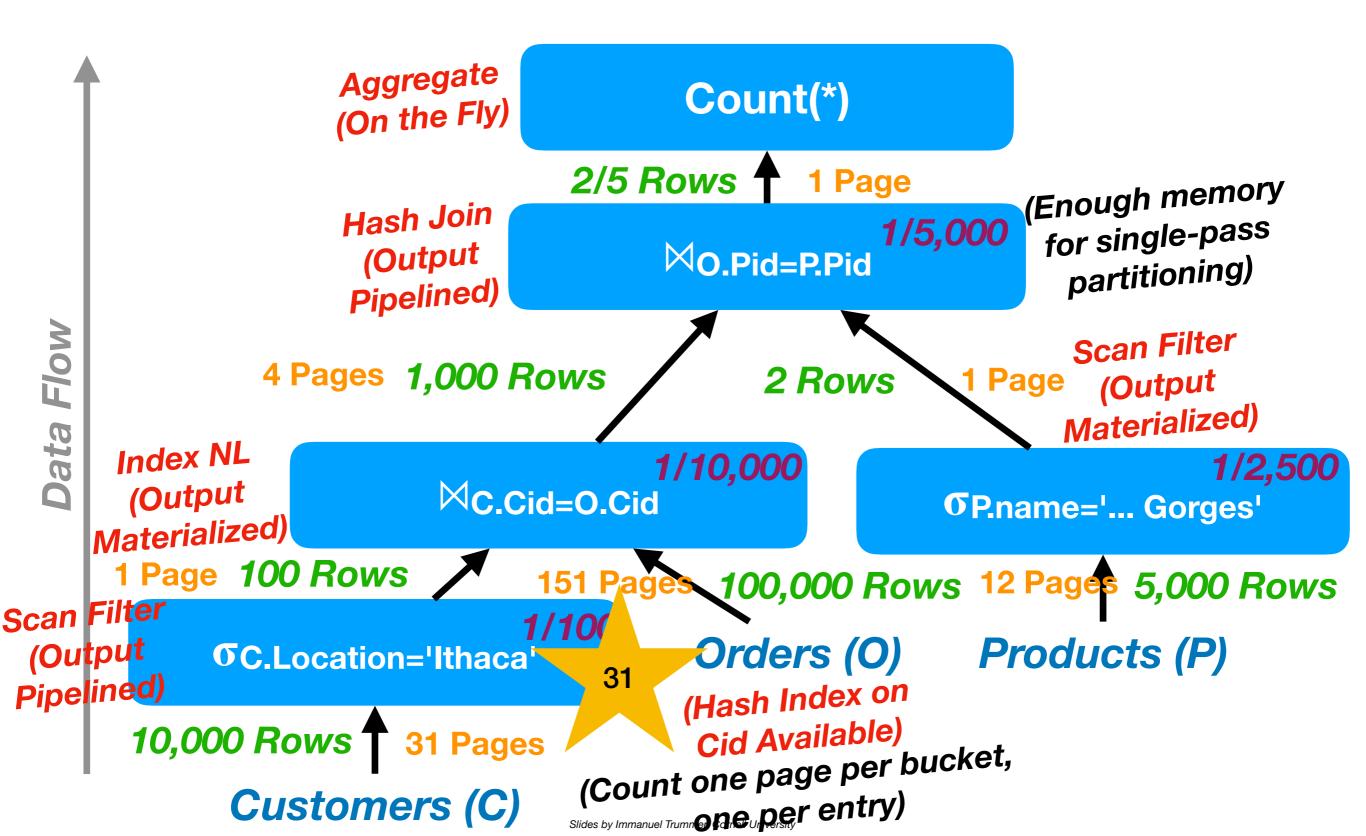
#### Size Estimation



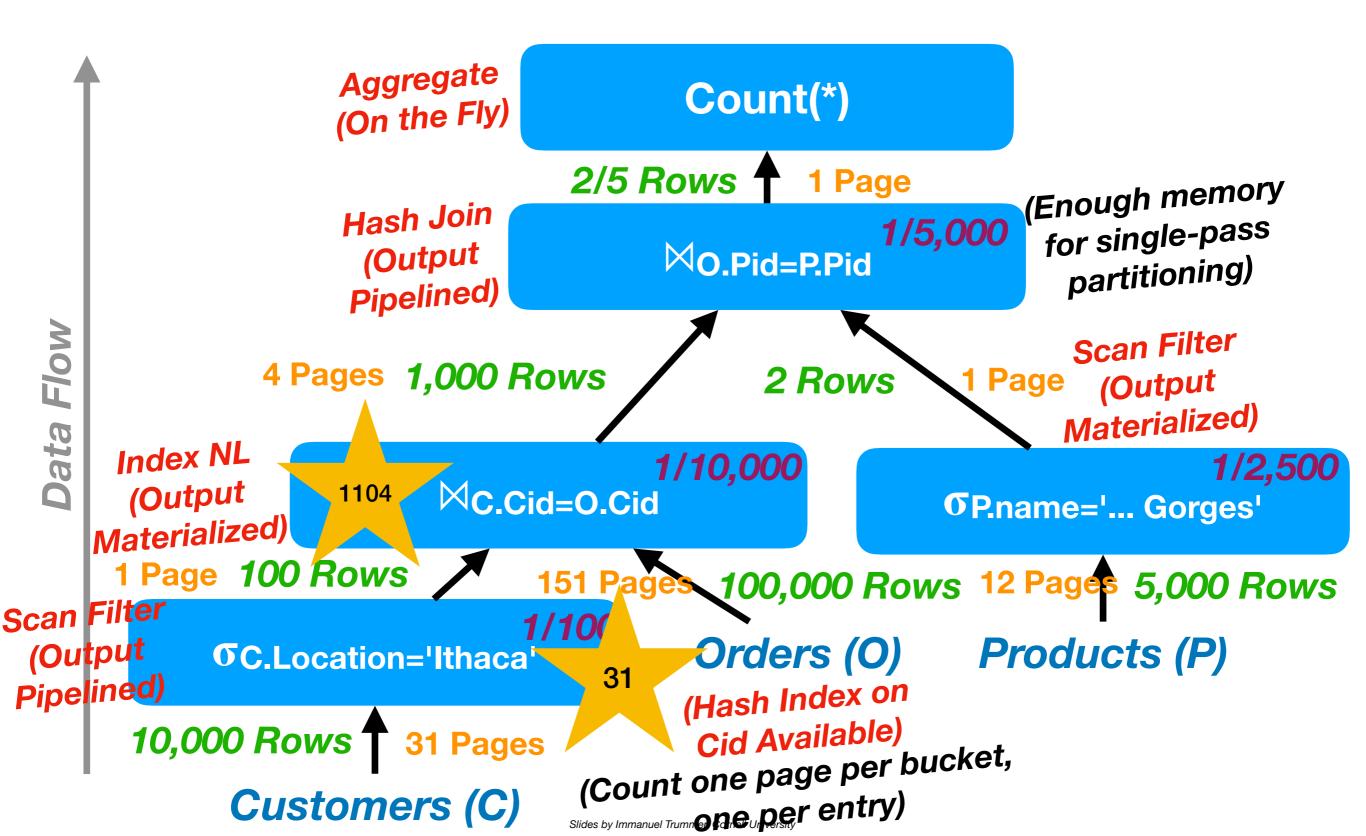
#### Cost Estimation

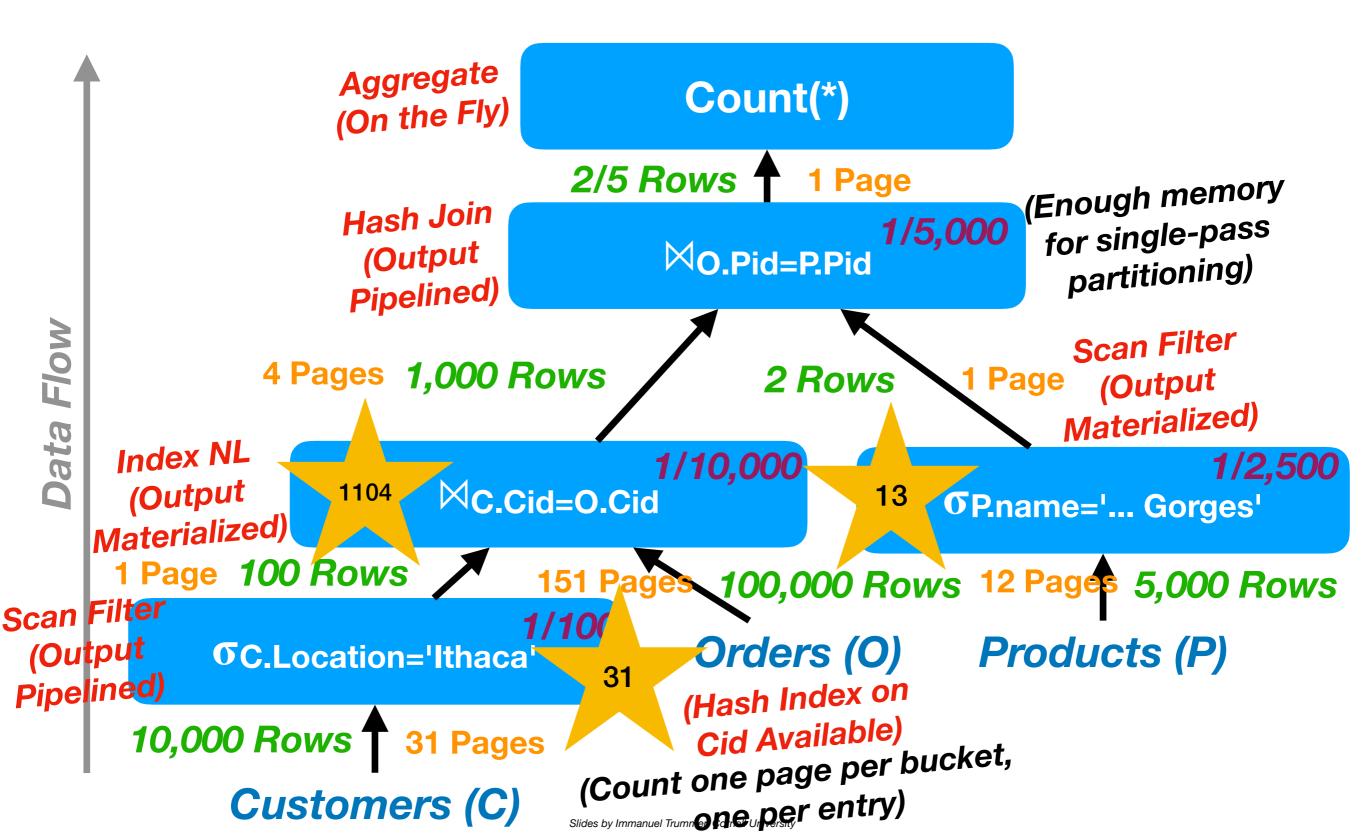


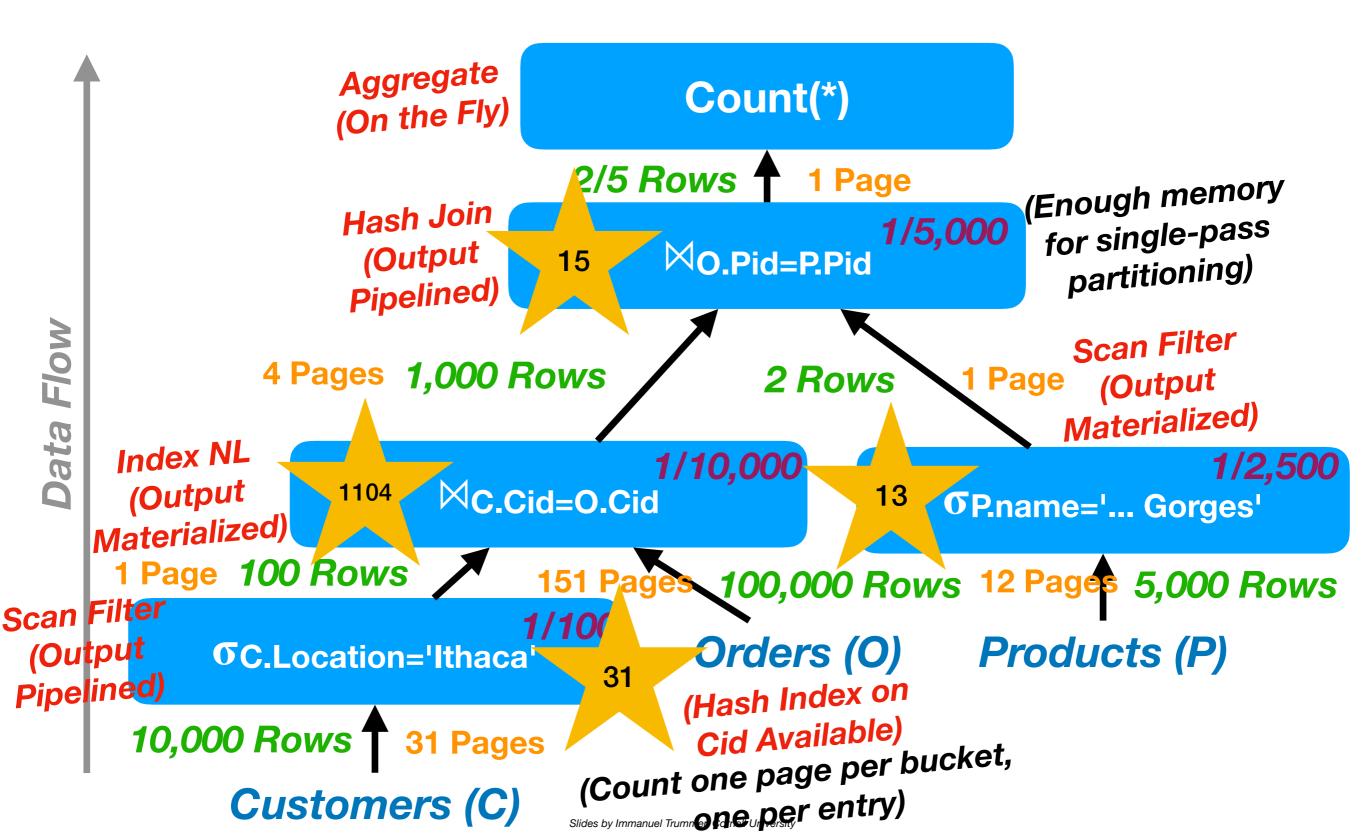
#### Cost Estimation

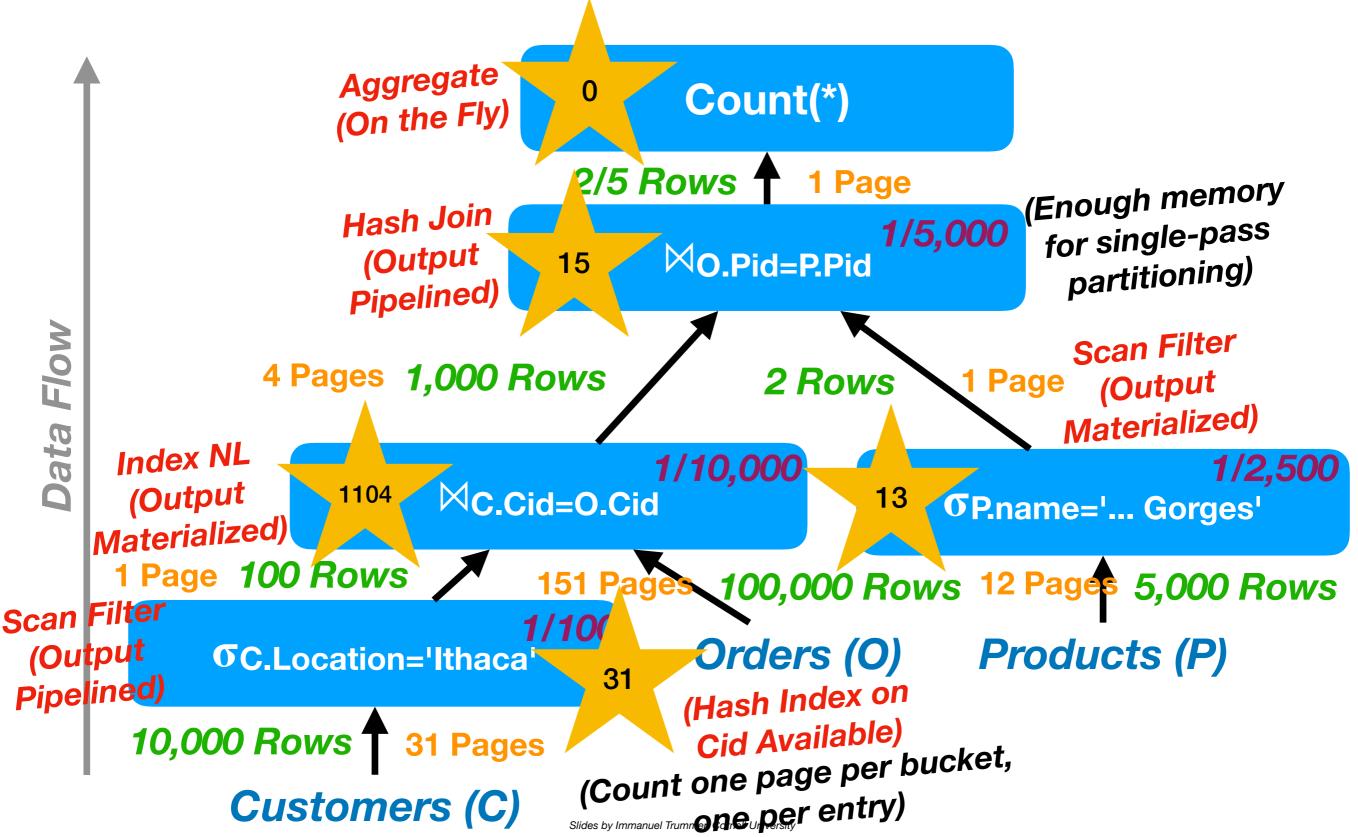


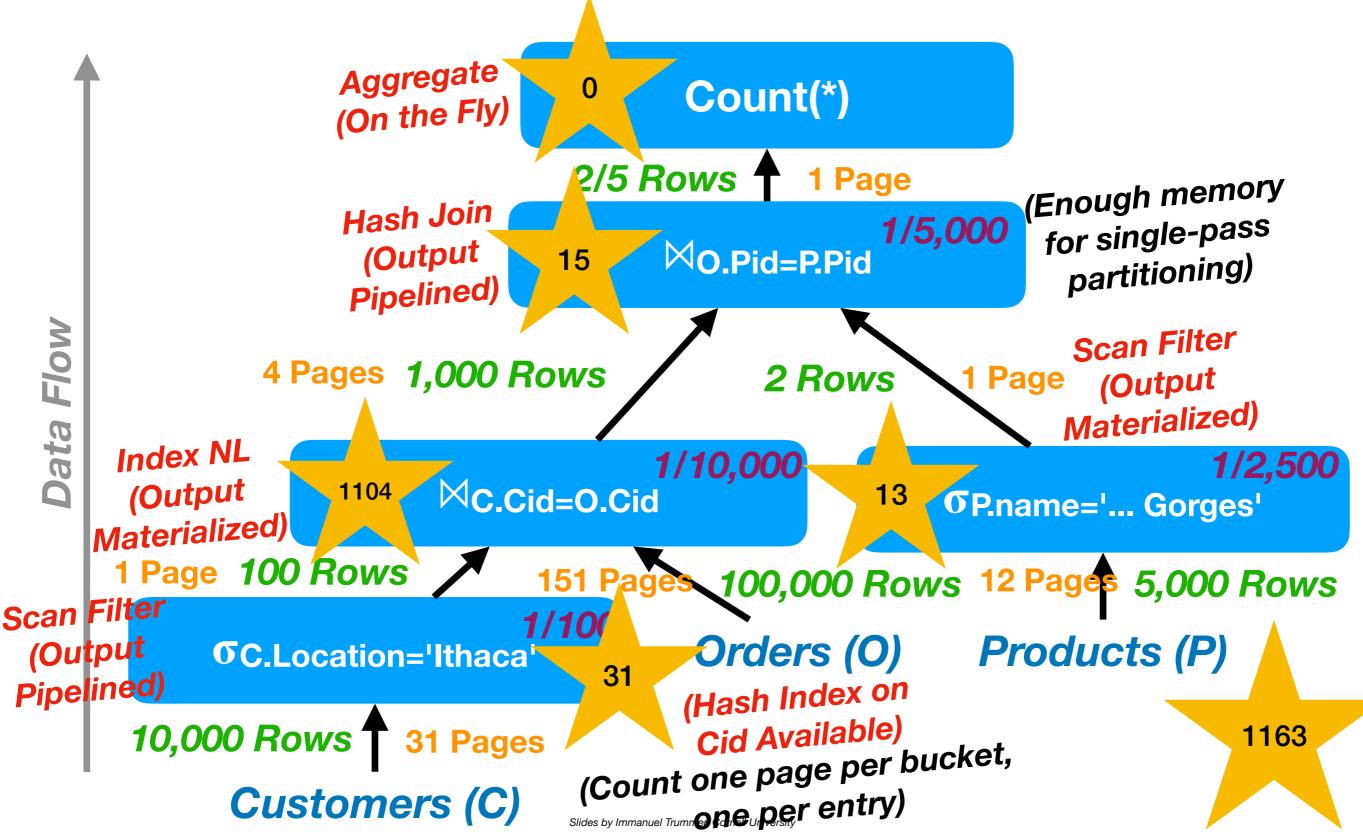
#### Cost Estimation



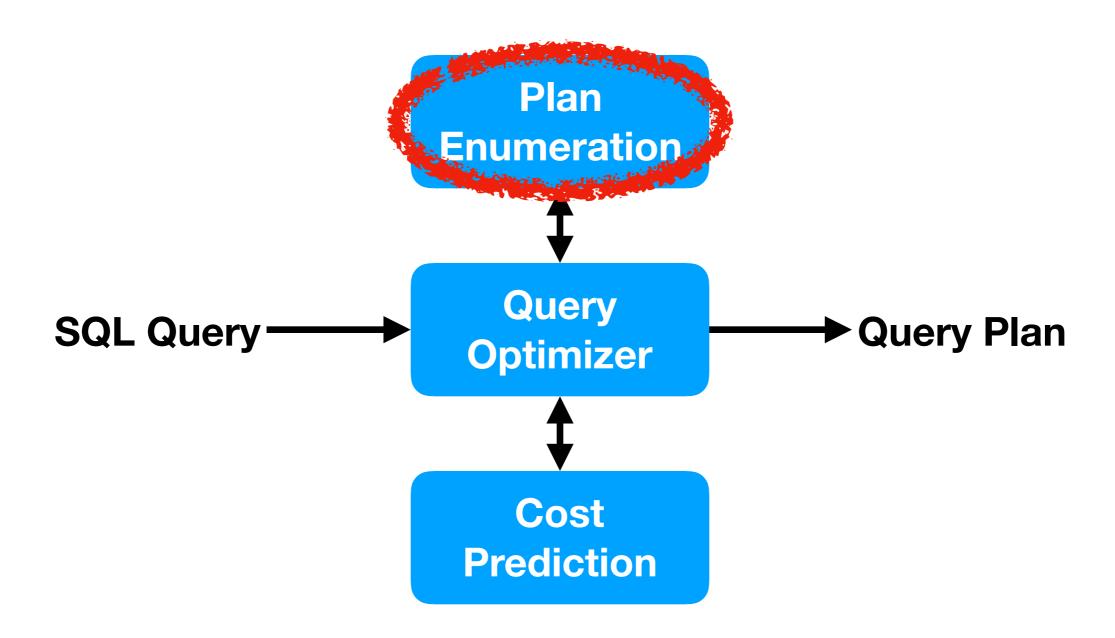








## Optimizer Overview



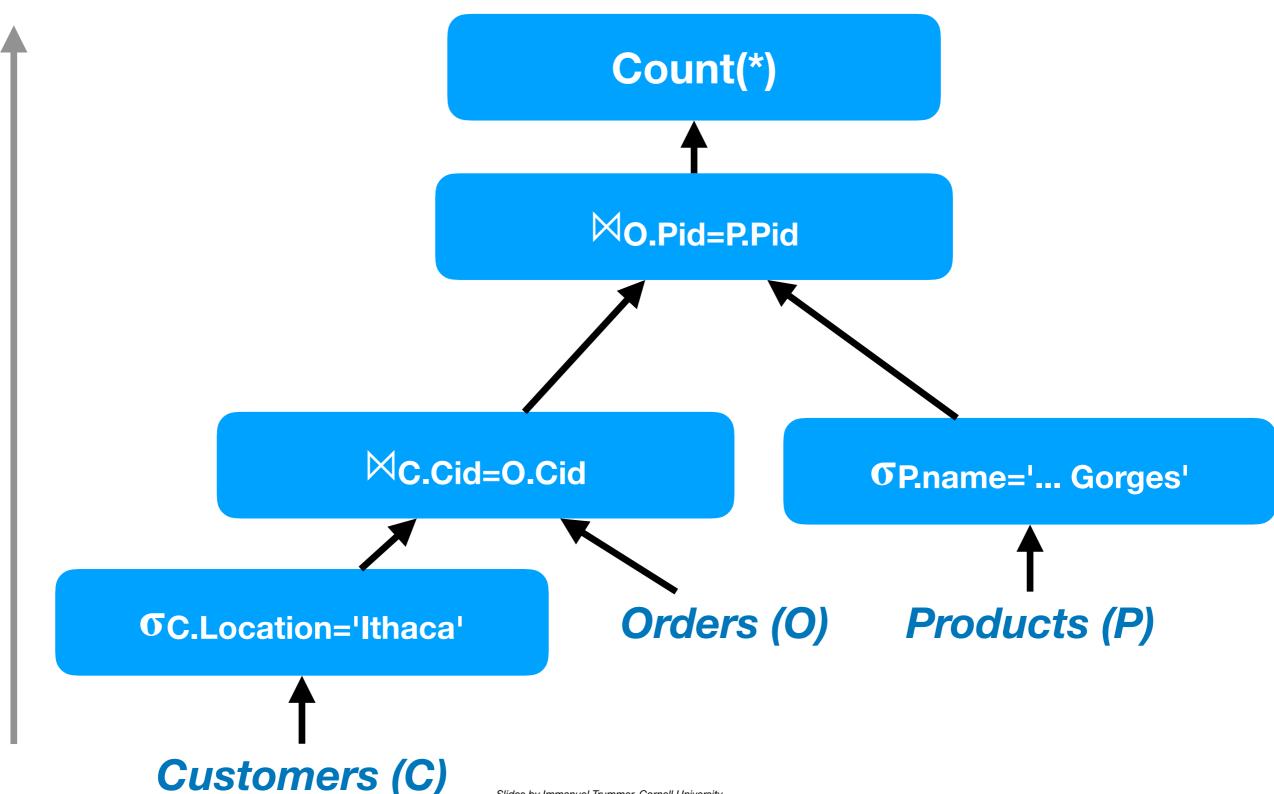
## Query Plan Space

- Decide order of operations and implementation
- Apply heuristic restrictions
  - H1: Apply predicates/projections early
  - H2: Avoid predicate-less joins
  - H3: Focus on left-deep plans

# H1: Early Predicates and Projections

- Processing more data is more expensive
- Want to reduce data size as quickly as possible
- Can do that by adding predicates (discarding rows)
- Can also do that by adding projections (discard columns)

## Can We Improve This?

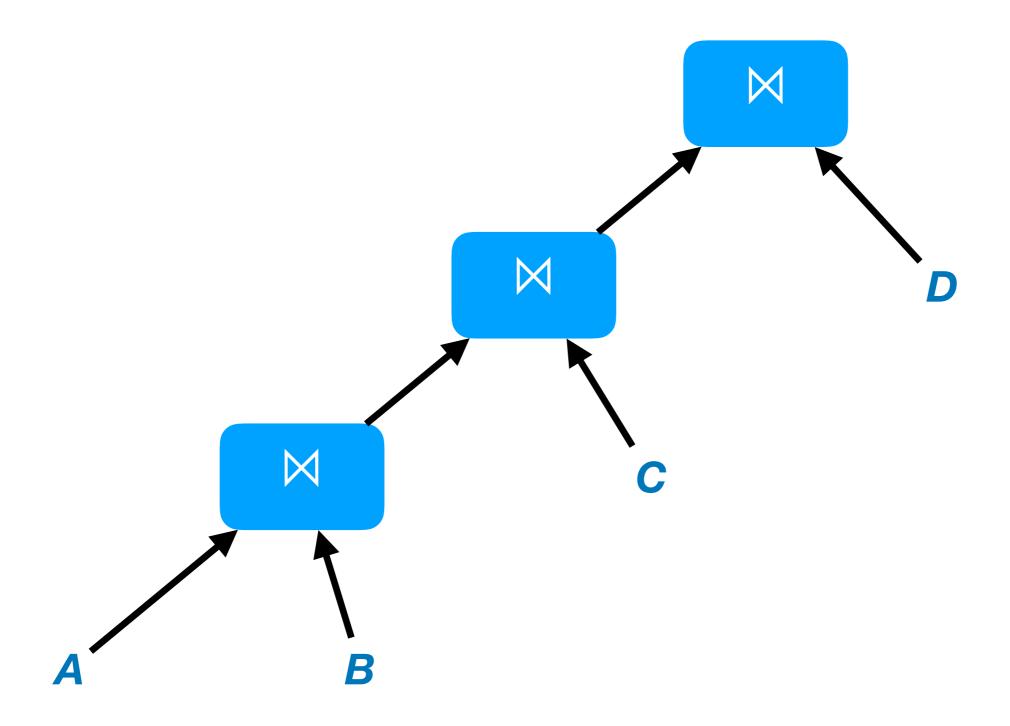


Slides by Immanuel Trummer, Cornell University

## H2: Avoid Joins without Predicates

- Join result size: product of input cardinality \* selectivity
- Selectivity is one when joining tables without predicates
- Often means very large join results, probably sub-optimal
- (Heuristic may discard optimal order in special cases)

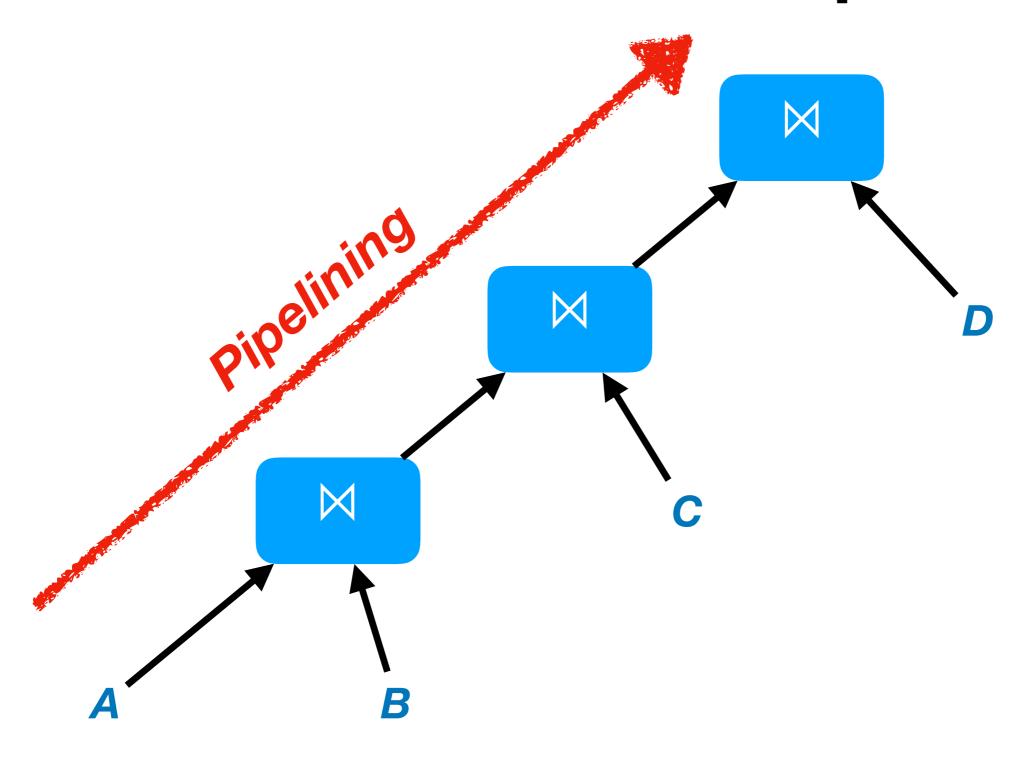
#### H3: Use "Left-Deep Plans"



## Why Left-Deep Plans?

- Allows pipelining: joins pass on result parts in-memory
- Allows to use indices on join columns (of base tables)

#### Focus on "Left-Deep Plans"



#### Naive Plan Enumeration

- Generate every possible plan
- Estimate cost for each plan
- Select plan with minimal cost

## Asymptotic Analysis

- Number of join orders can grow quickly O(nrTables!).
- Number of join orders lower-bounds number of plans
- Enumerating plans is considered impractical

## Principle of Optimality

- Want cheapest plan to join set of tables
- This plan joins table subsets "on the way"
- Assume we use sub-optimal plan for joining table subset
- Replacing by a better plan can only improve overall cost
- This is generally called the principle of optimality

## Efficient Optimization

- Find optimal plans for (smaller) sub-queries first
  - Sub-query: joins subsets of tables
- Compose optimal plans from optimal sub-query plans



