

# CSE544

# Data Management

## Lectures 11-12

## Advanced Query Processing

# Announcements

- HW3 due on Friday
- No lecture on Monday (President's day)
- Project milestone due next Friday

# Advanced Query Processing

Current optimization techniques:

optimal plan given current statistics

- Ignores asymptotic runtime
- Sometimes asymptotic is provably bad

Advanced techniques: find optimal *asymptotic* runtime

# Examples

```
SELECT count(*)  
FROM Author;
```

Answer: 2419705  
Time: < 1s  
Asymptotic:  $O(N)$

```
SELECT count(*)  
FROM Publication;
```

Answer: 4659997  
Time: < 1s  
Asymptotic:  $O(N)$

```
SELECT count(*)  
FROM Author, Publication;
```

Answer: 2419705 \* 4659997  
**Timeout**  
Asymptotic:  $O(N^2)$   
Should be:  $O(N)$

# Examples

```
SELECT count(*)  
FROM Author x,  
      Authored y,  
      Publication z  
WHERE x.authorid=y.authorid  
      and y.pubid=z.pubid  
      and z.year < 2015
```

Optimize this! (At home...)

# Outline

- Acyclic queries, Yannakakis algorithm
- Tree decomposition of cyclic queries
- Worst-case optimal algorithm; next week

# Conjunctive Queries

- A CQ is:

$$Q(X) :- R_1(X_1), R_2(X_2), \dots, R_m(X_m)$$

- Same as a single datalog rule

# Types of CQ

- **Full CQ:** all variables are head variables

$Q(x,y,z,u) :- R(x,y),S(y,z),T(z,u)$

$Q(*) :- \dots$

- **Boolean CQ:** no variables are head variables

$Q() :- R(x,y),S(y,z),T(z,u)$

- CQ with aggregates:

$Q(x,\text{sum}(u)) :- R(x,y),S(y,z),T(z,u)$



# Generalized Distributivity Law

- Basic idea: group-by commutes with join if we write it the right way

# Generalized Distributivity Law

$$Q(\text{count}(*)) = R(x,y), S(y,z)$$

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R:

x	y
a	b
c	b
d	f
g	h

S:

y	z
b	g
b	k
h	m

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Answer = 5

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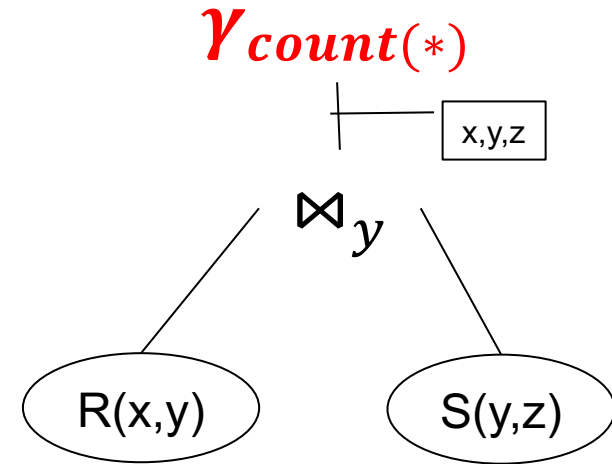
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Runtime =  $O(N^2)$



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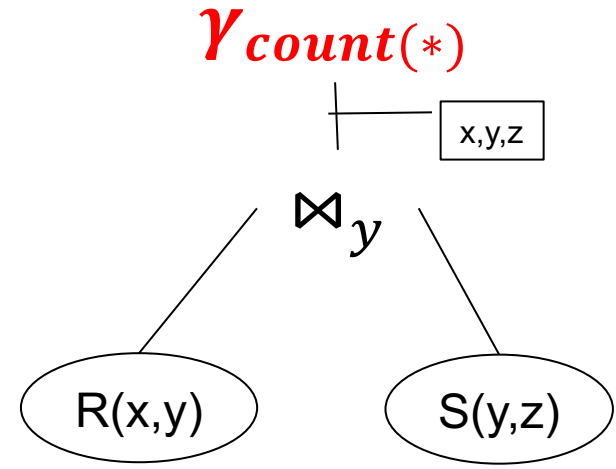
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$$A(y, \text{count}(x) \text{ as } c) = R(x,y)$$

$$B(y, \text{count}(z) \text{ as } d) = S(y,z)$$

$$Q(\text{sum}(c*d)) = A(y,c), B(y,d)$$



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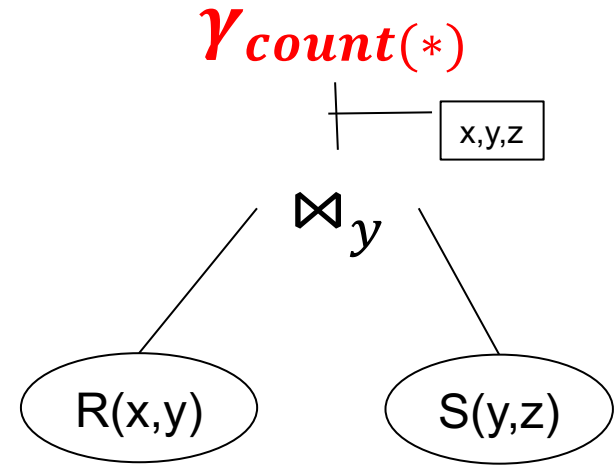
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A:

y	c
b	2
f	1
h	1

B:

y	c
b	2
h	1

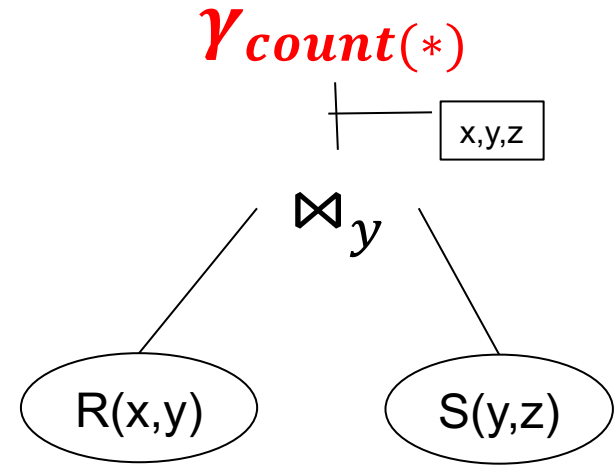
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R:	x	y	S:	y	z
	a	b		b	g
	c	b		b	k
	d	f		h	m
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A:	y	c	B:	y	c	A⋈B	y	c	d
	b	2		b	2		b	2	2
	f	1		h	1		h	1	1
	h	1							



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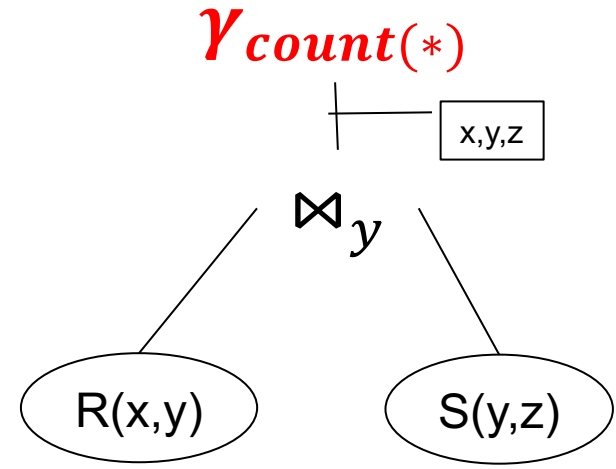
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A:

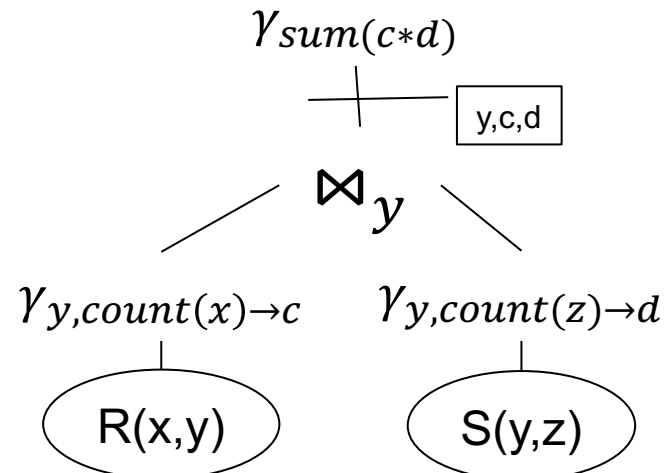
y	c
b	2
f	1
h	1

B:

y	c
b	2
h	1

$A \bowtie B$

y	c	d
b	2	2
h	1	1



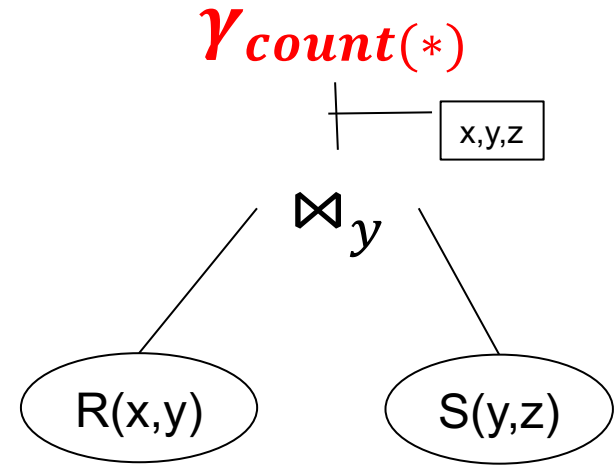
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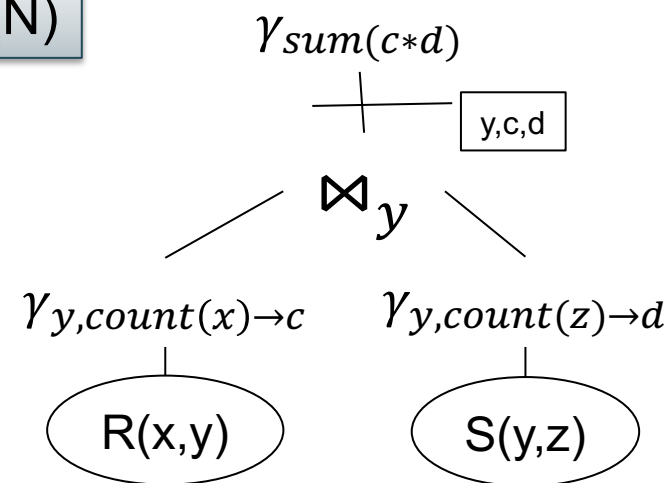
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# Acyclic Queries

Q is acyclic if its atoms can be placed in a tree T such that for every variable the set of nodes that contain that variable form a connected component

T is called  
join tree

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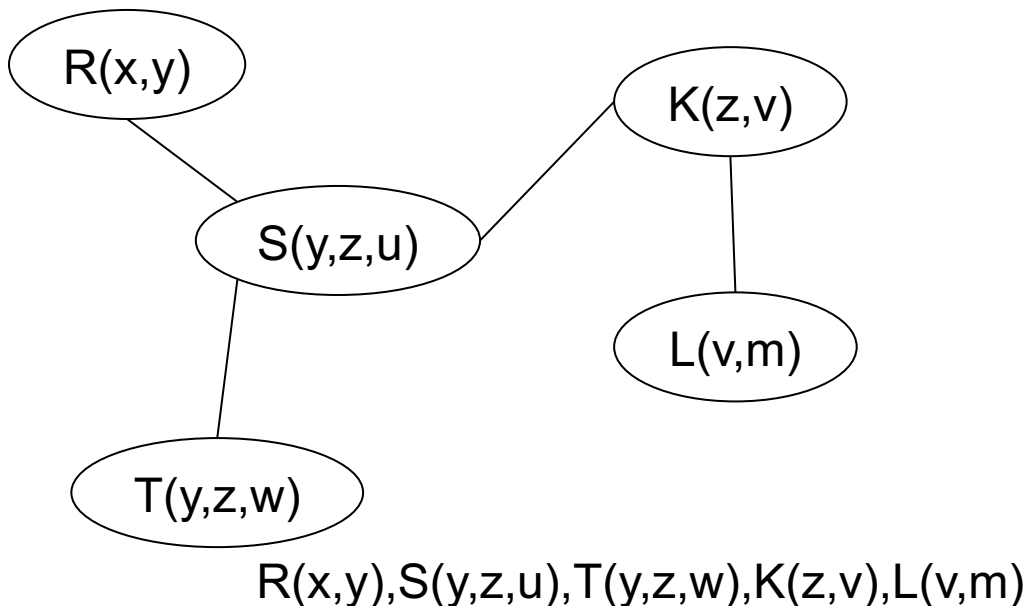
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$R(x,y), S(y,z,u), T(y,z,w), K(z,v), L(v,m)$

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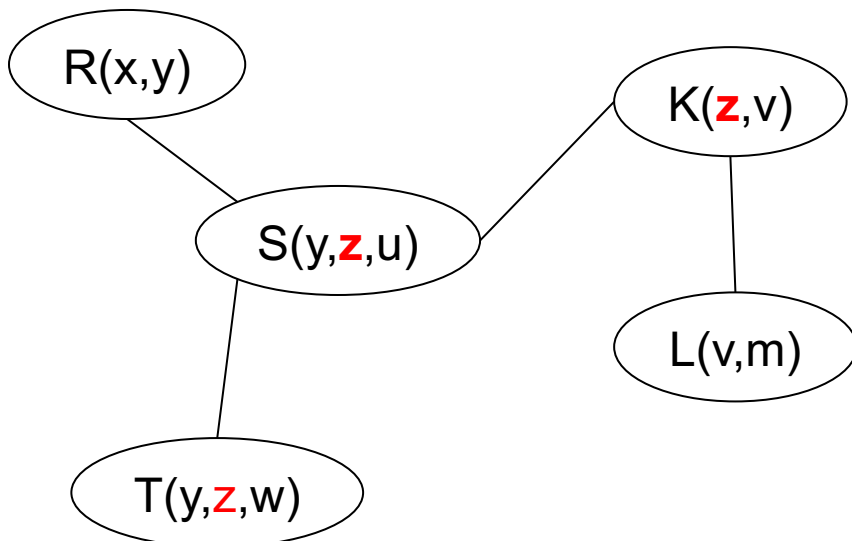
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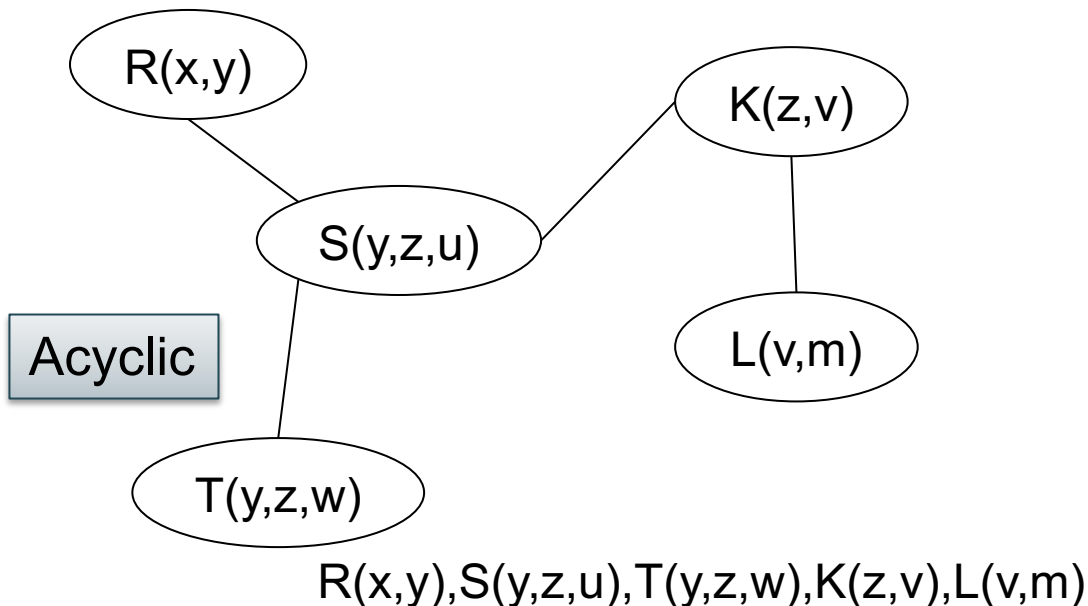
E.g. **z** forms a connected component

$R(x,y), S(y,z,u), T(y,z,w), K(z,v), L(v,m)$

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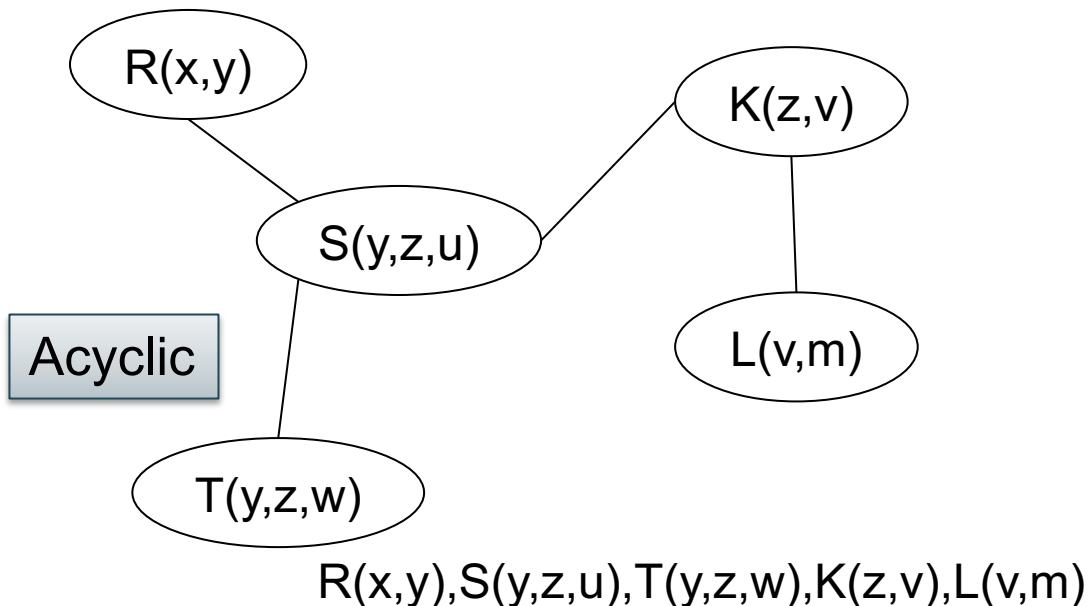




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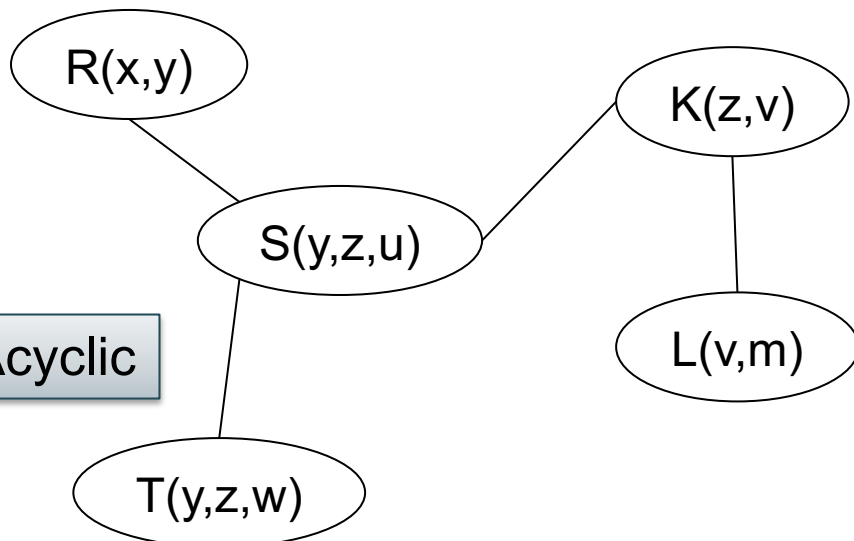


$R(x,y), S(y,z), T(z,x)$

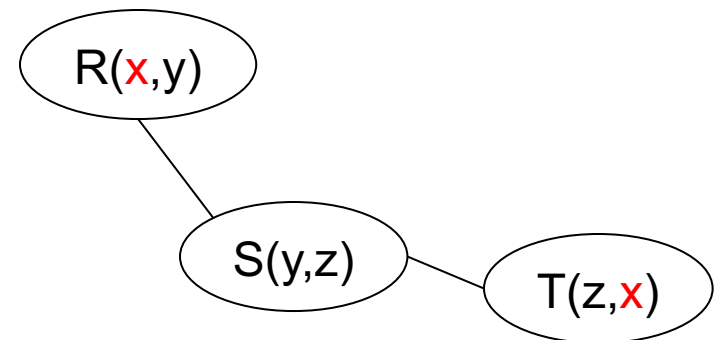
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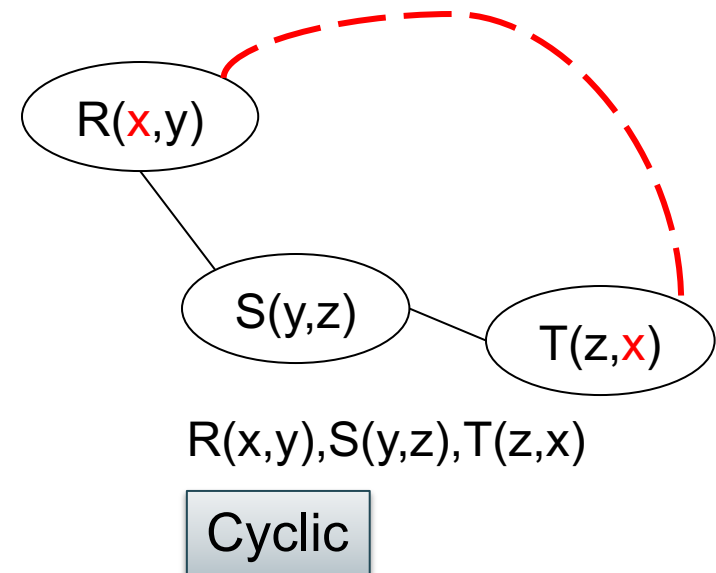
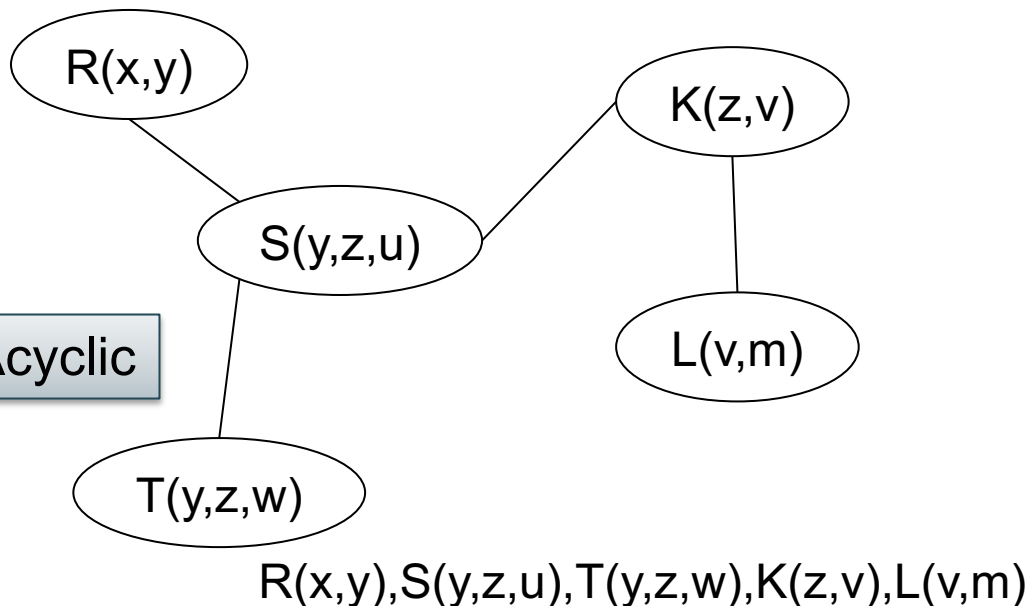


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# A Theorem

Q = an acyclic query Q that is:

- Boolean, or
- Full, or
- Aggregate with  $\leq 1$  group-by variable

**Theorem** Q can be computed in time\*:  
 $\tilde{O}(|Input| + |Output|)$

\*  $\tilde{O}$  means *plus a logarithmic factor (for sorting)*

# Yannakakis Algorithm

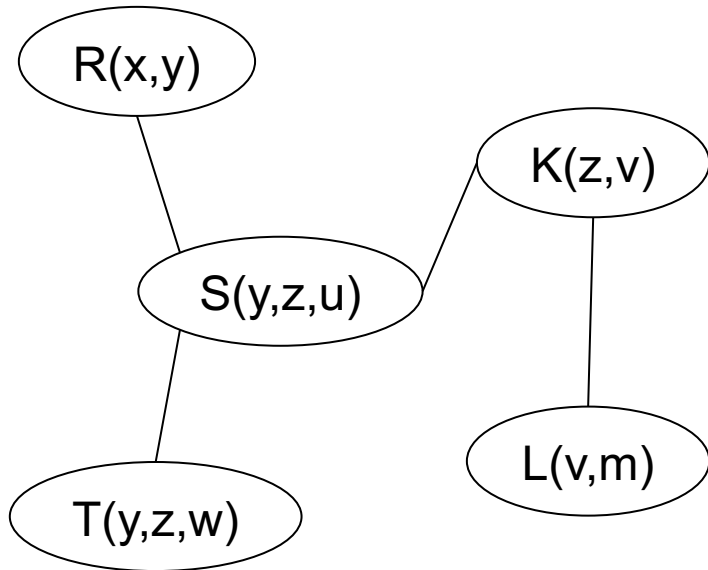
- Step 1: semi-join reduction
  - Pick any root node in the join tree of  $Q$
  - Semi-join reduction from leaves to root
  - Semi-join reduction from root to leaves
- Step 2:
  - Compute the joins bottom up,
  - Push group-by down

# Example: Full CQ

$$Q(*) = R(x,y), S(y,z,u), T(y,z,w), K(z,v), L(v,m)$$

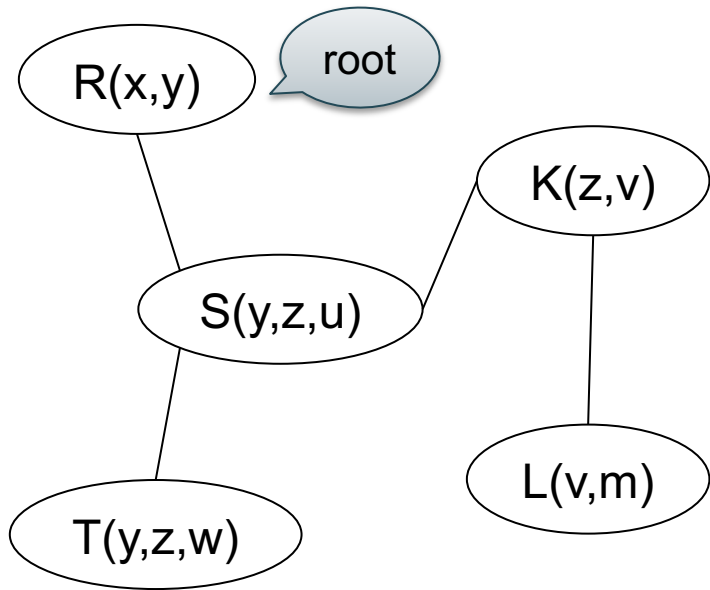
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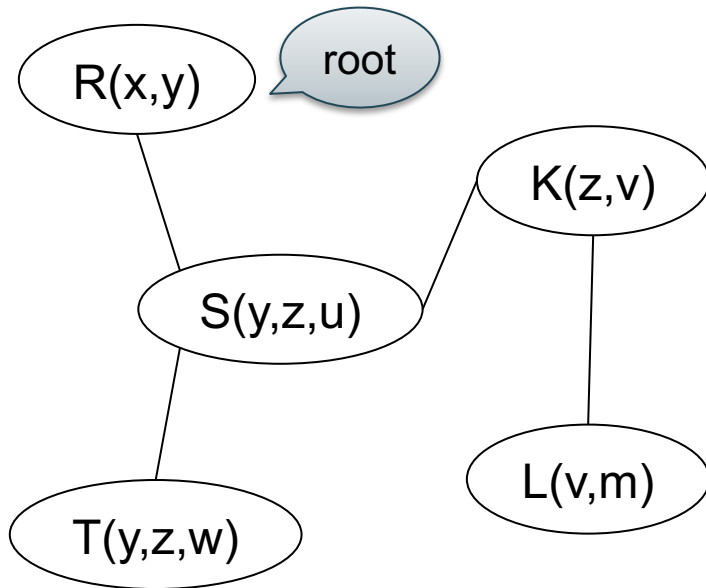
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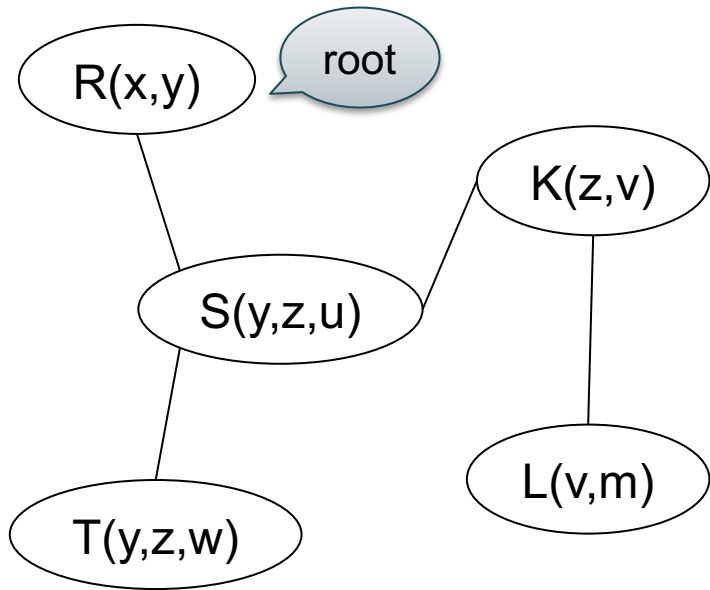
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-- Leaves to root:  
 $K :- K \bowtie L$

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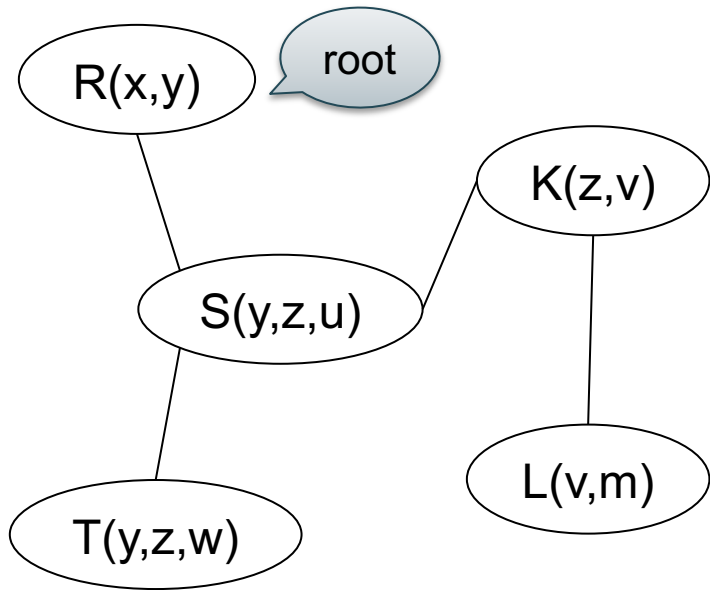
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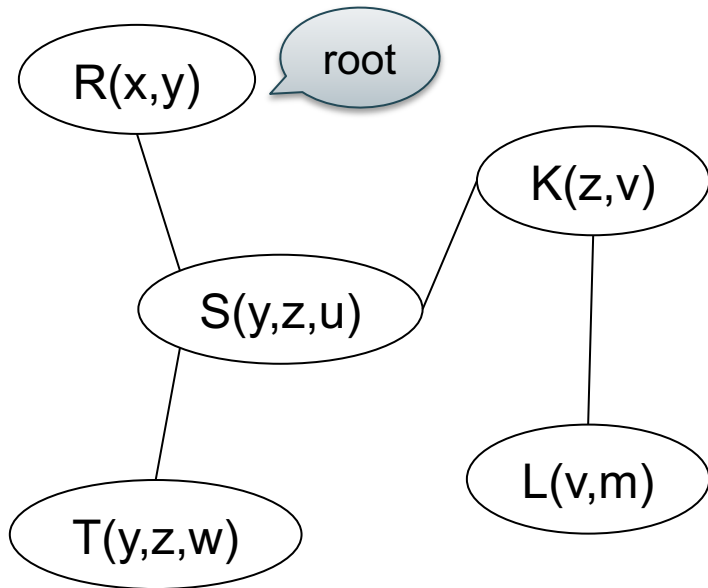
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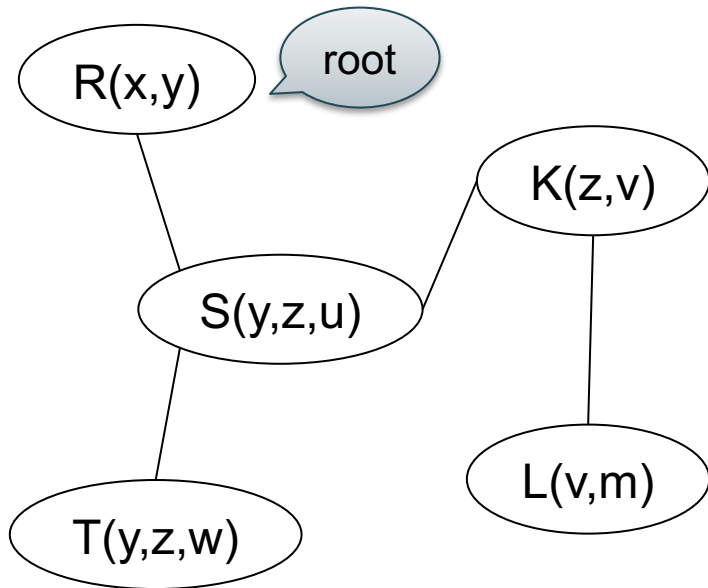
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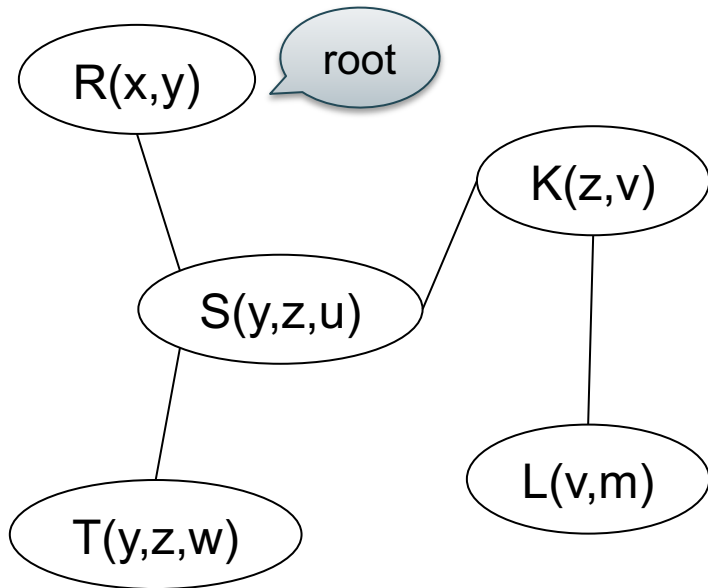
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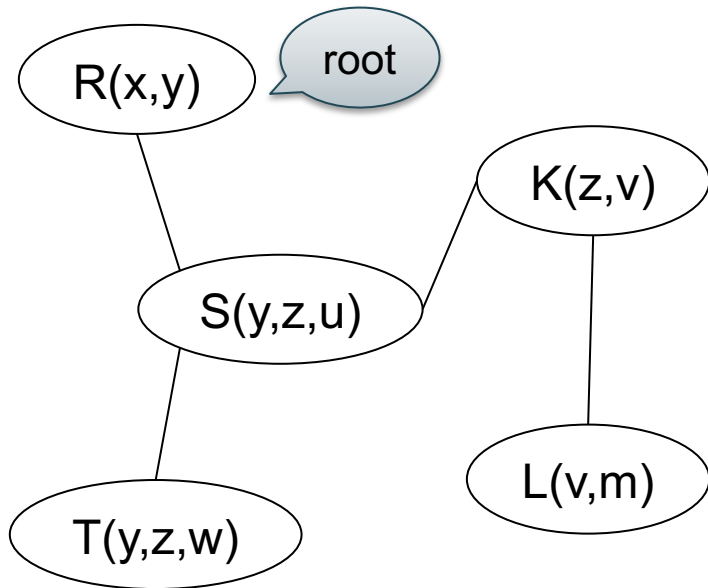
$R :- R \bowtie S$

-- Root to leaves:

$S :- S \bowtie R$

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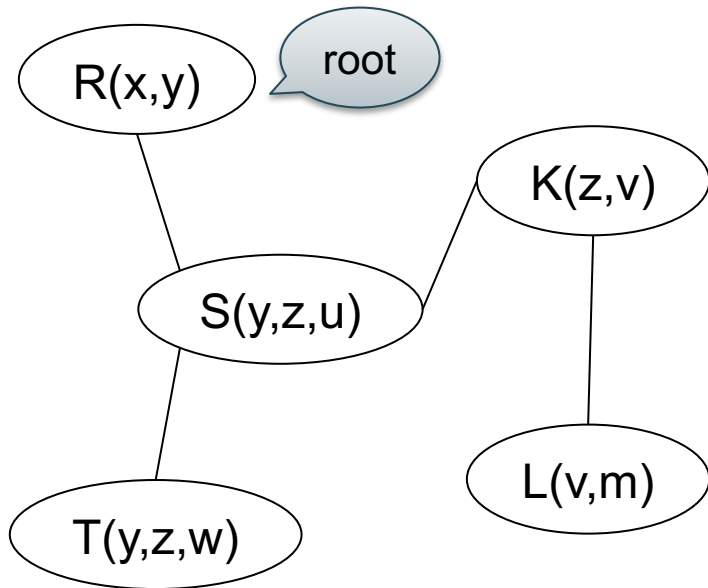
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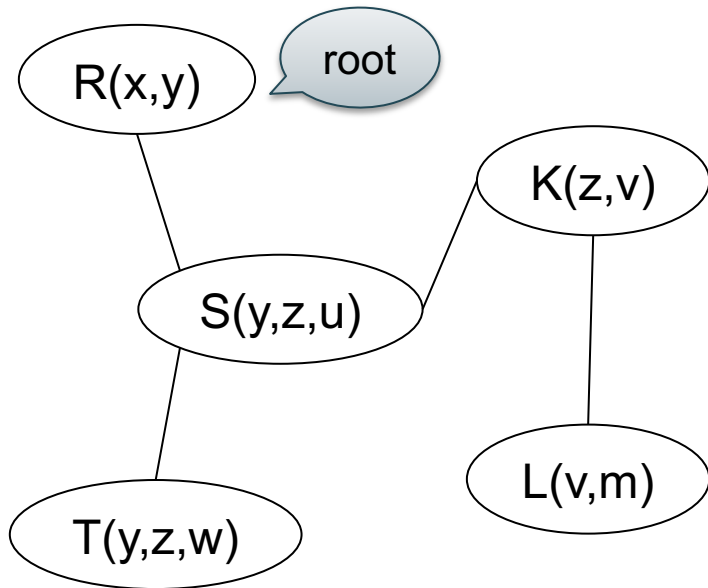
$T :- T \bowtie S$

$K :- K \bowtie S$



# Example: Full CQ

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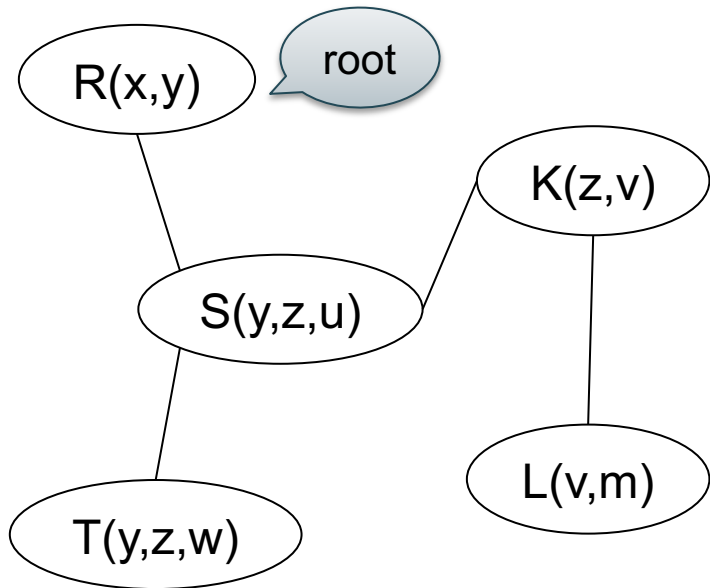
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# Example: Full CQ

$Q(*) = R(x,y), S(y,z,u), T(y,z,w), K(z,v), L(v,m)$

Join (any order in the tree)



-- Leaves to root:

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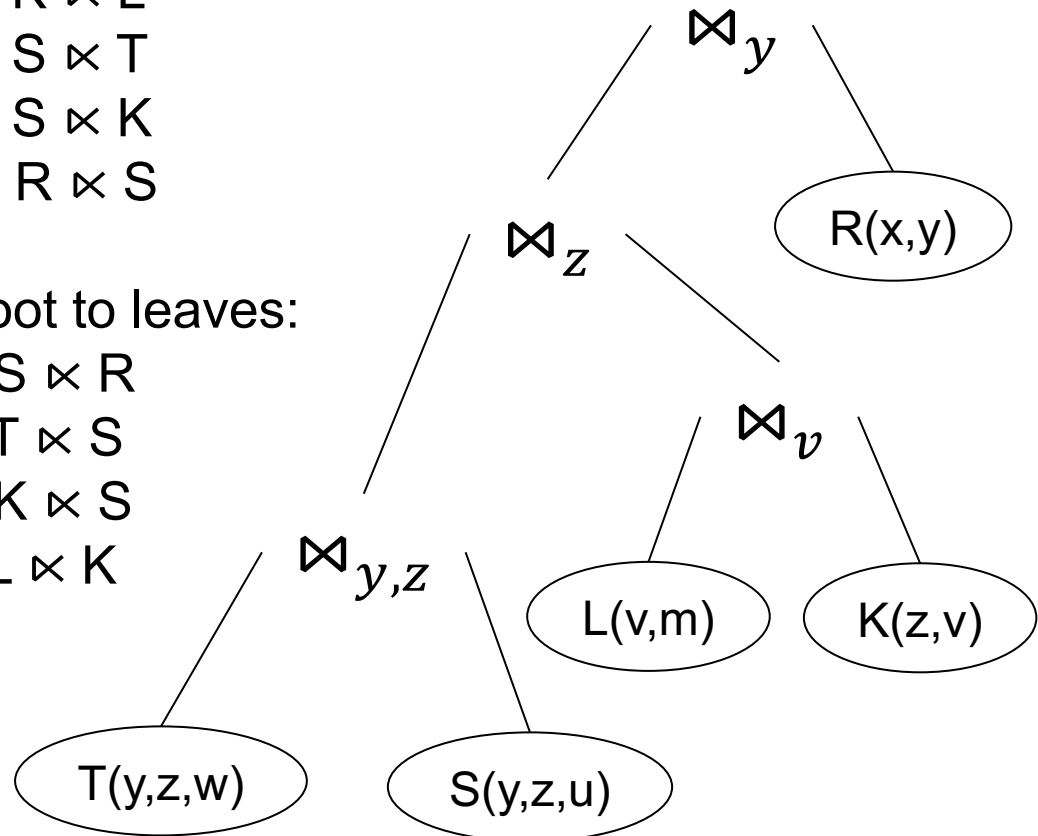
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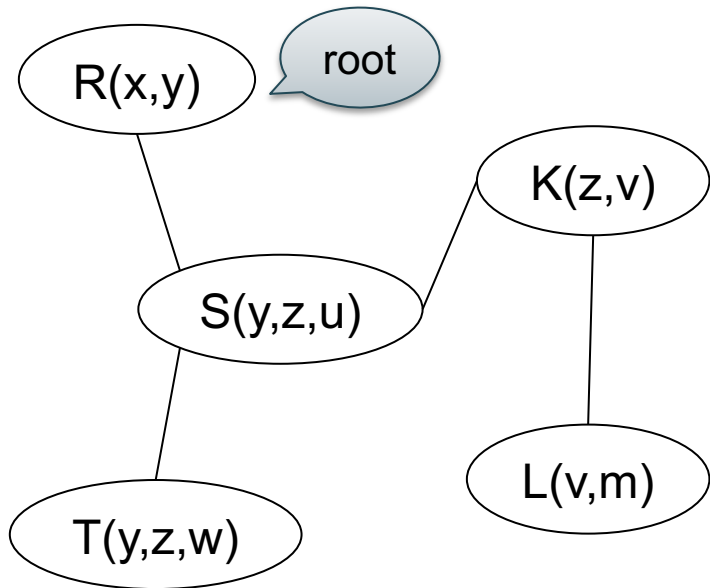
$L :- L \bowtie K$



# Example: Full CQ

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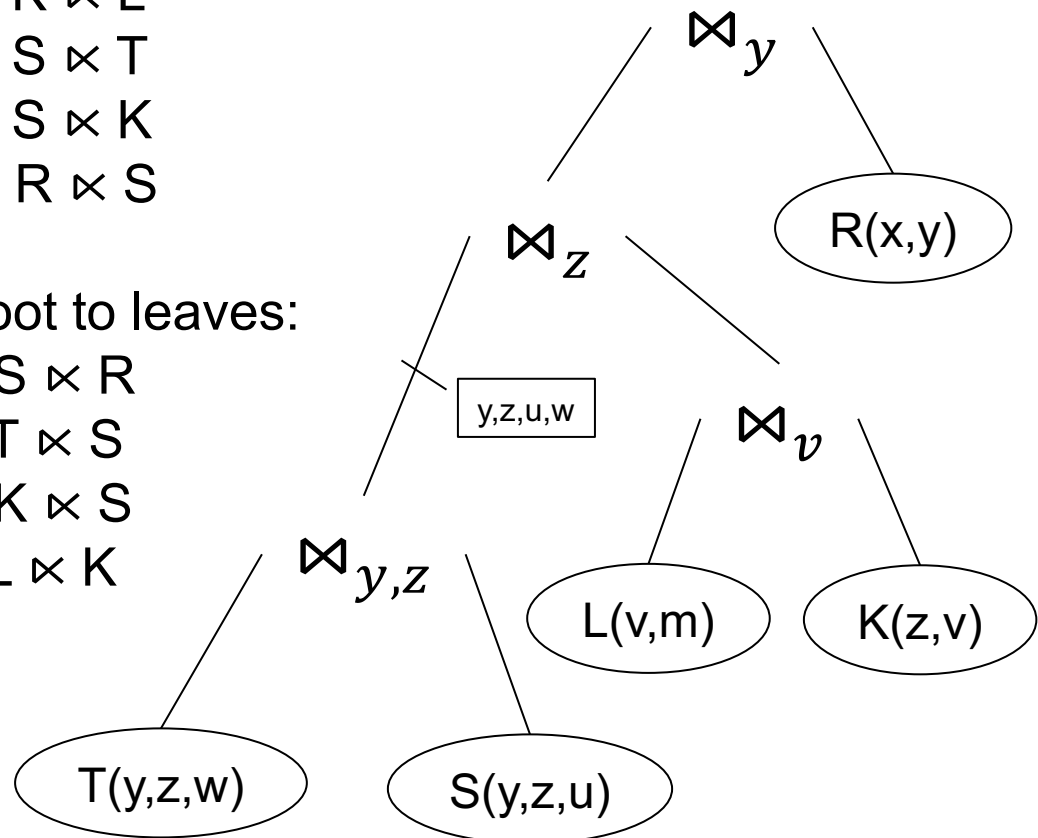
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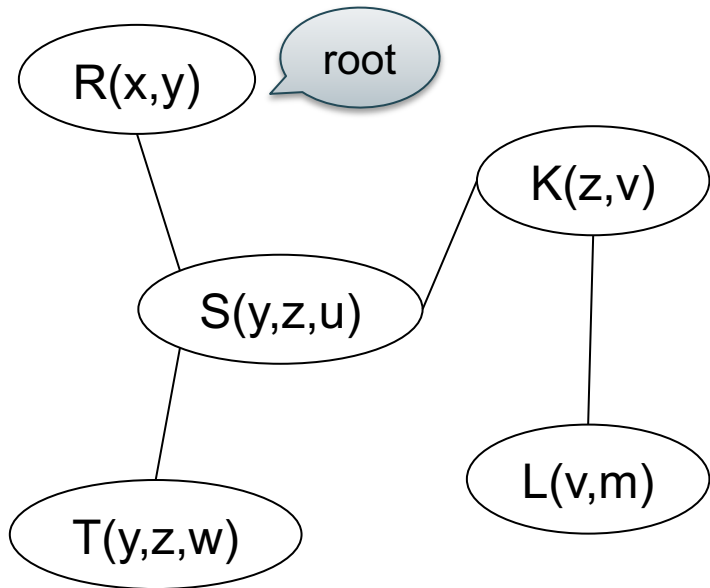
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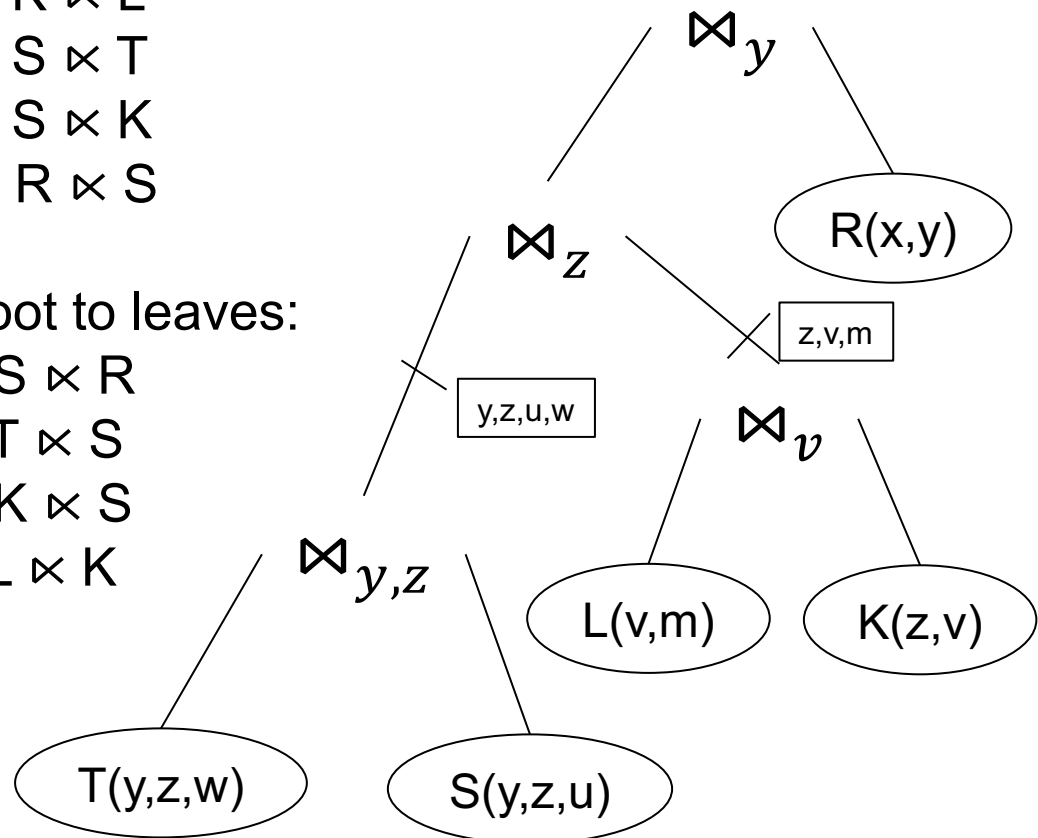
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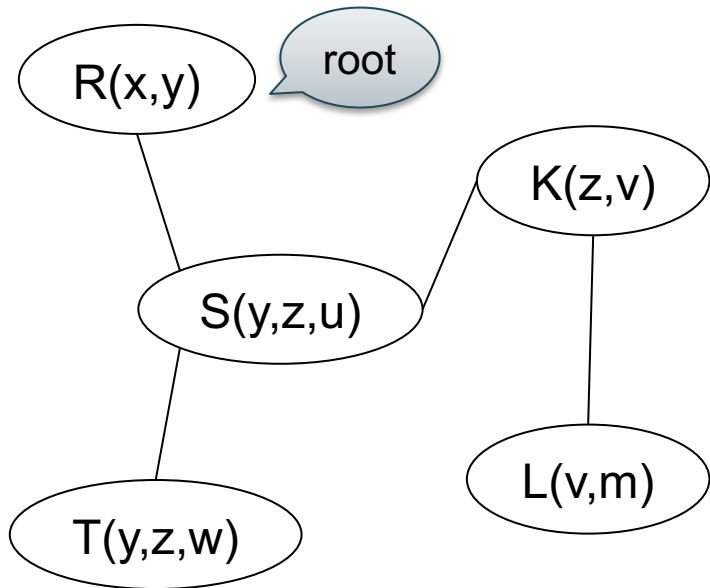
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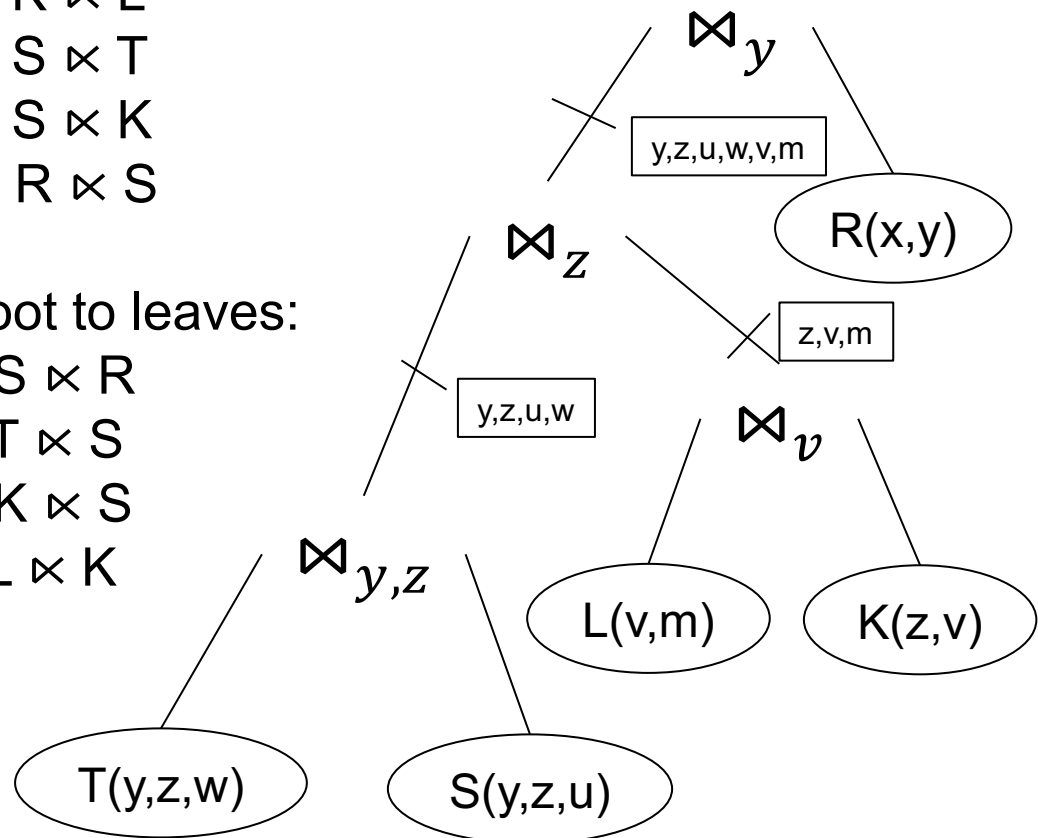
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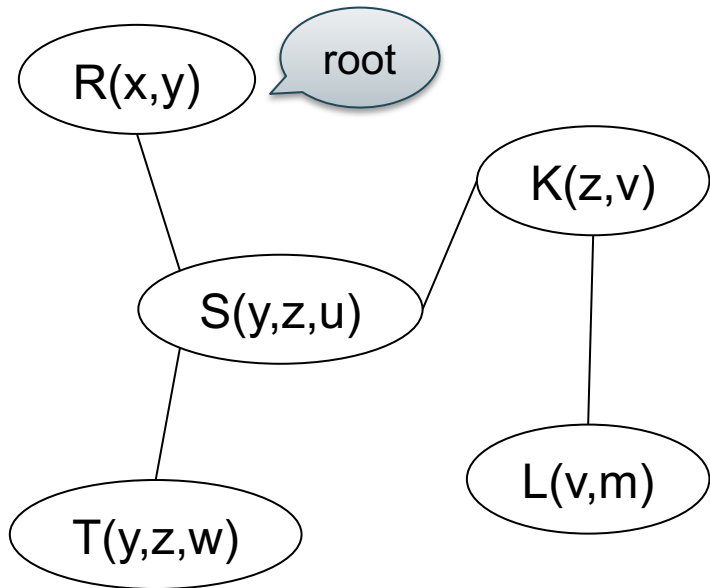
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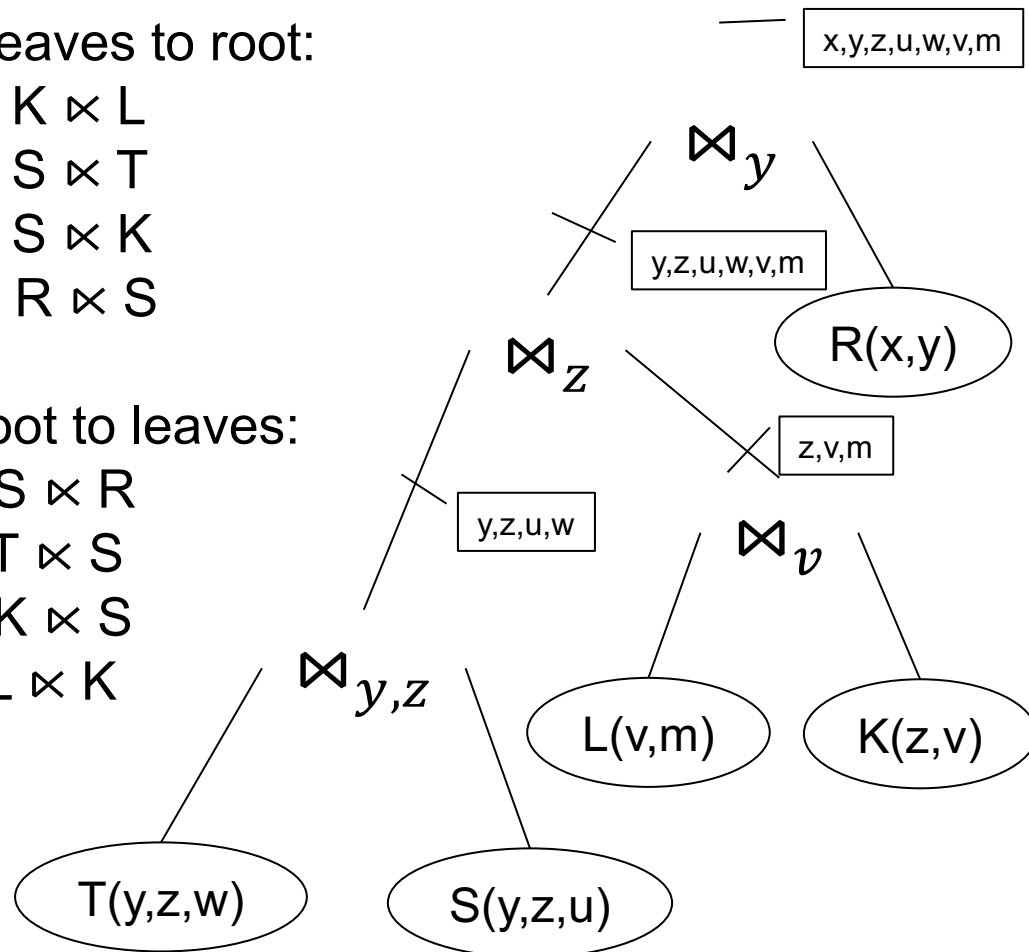
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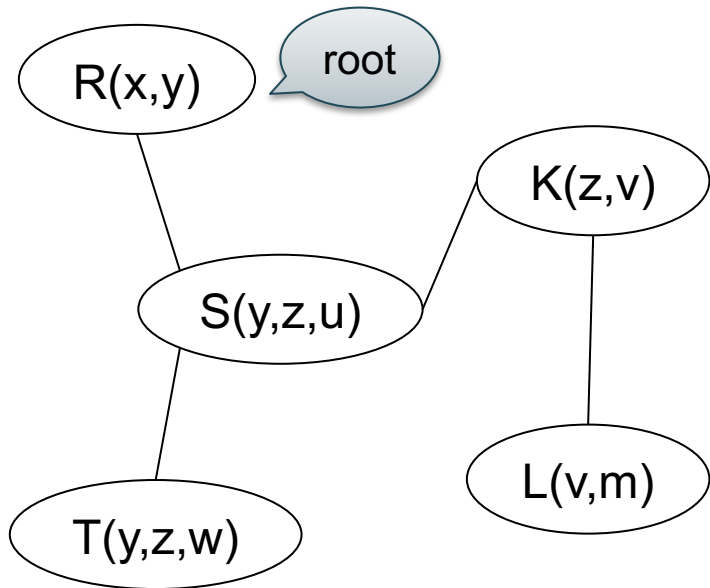
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$R :- R \bowtie S$

-- Root to leaves:

$S :- S \bowtie R$

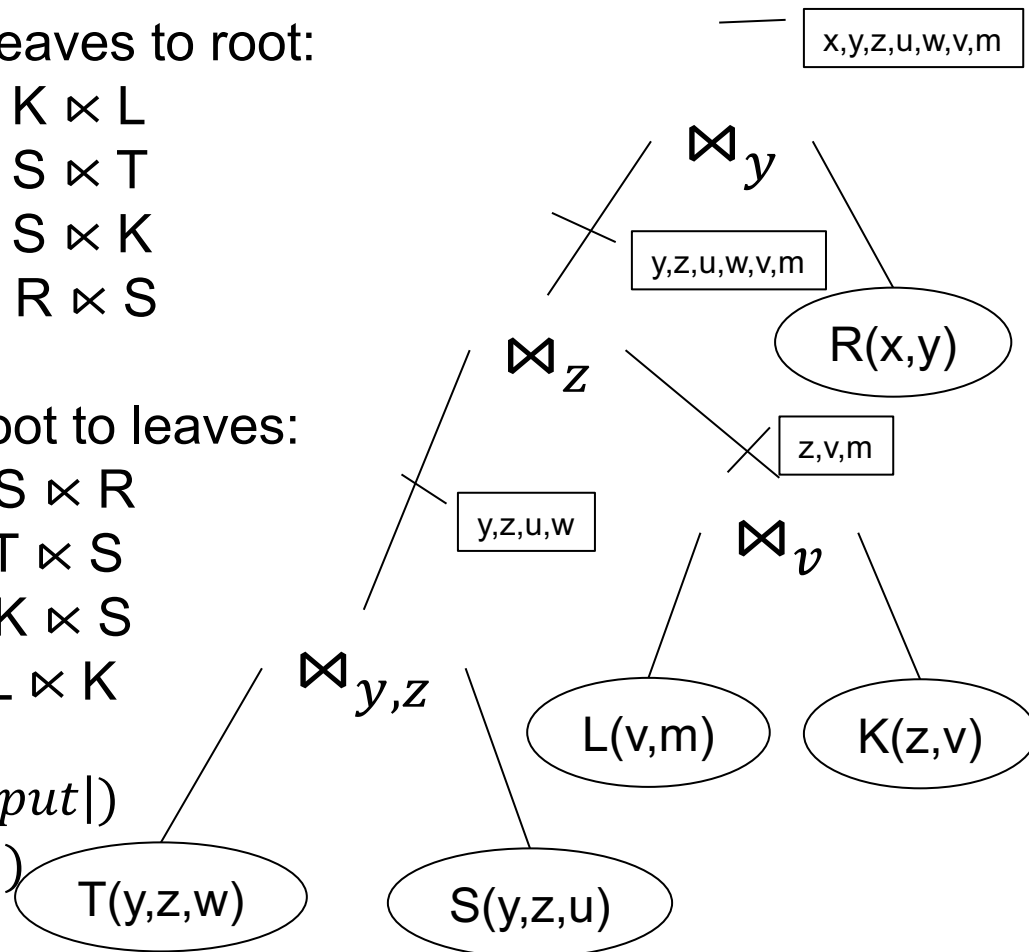
$T :- T \bowtie S$

$K :- K \bowtie S$

$L :- L \bowtie K$

Runtime:

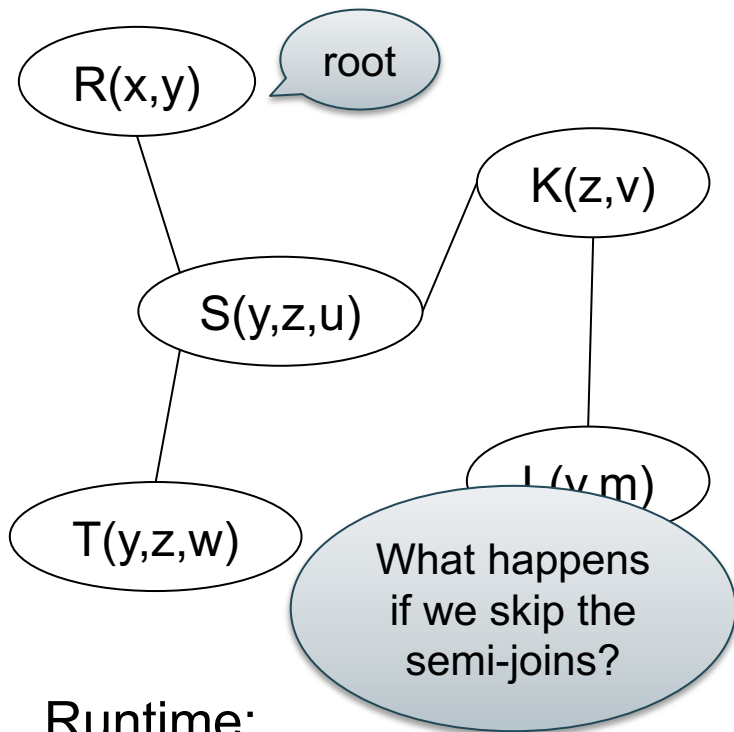
- Every semi-join takes time  $\tilde{O}(|Input|)$
- Every join takes time  $\tilde{O}(|Output|)$



# Example: Full CQ

$Q(*) = R(x,y), S(y,z,u), T(y,z,w), K(z,v), L(v,m)$

Join (any order in the tree)

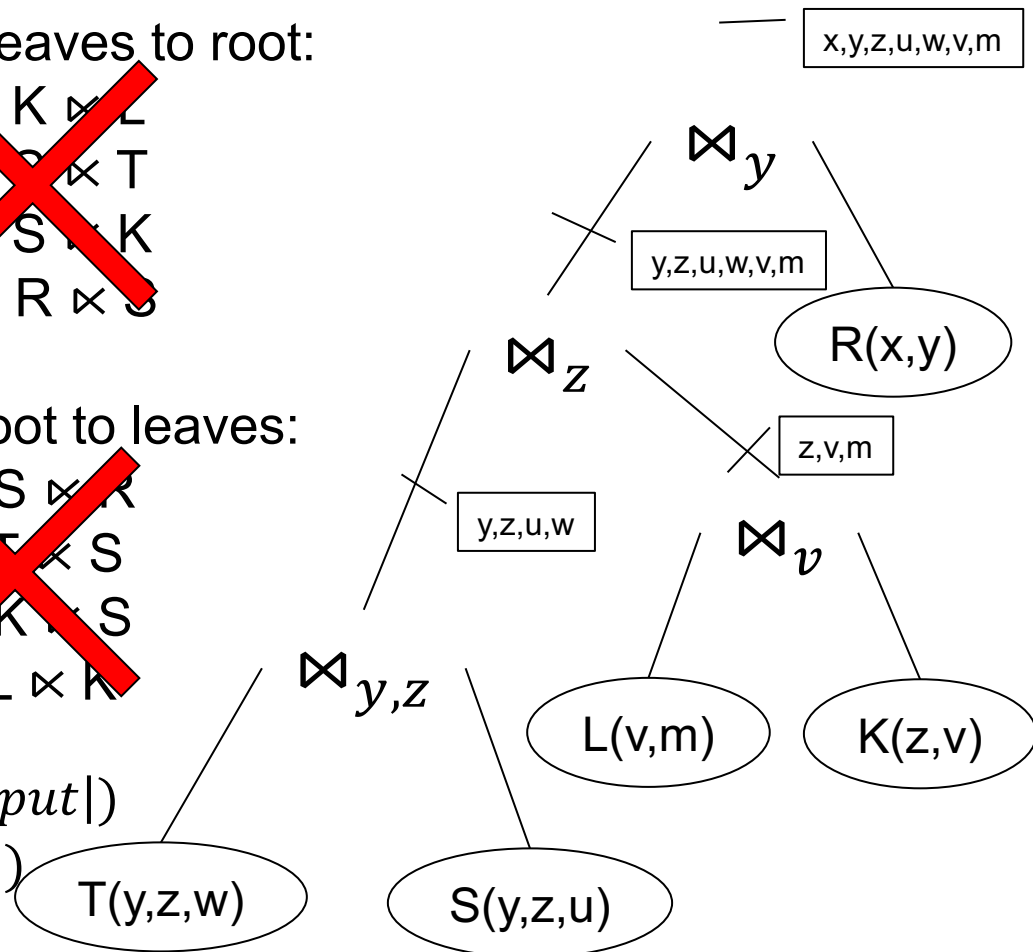


-- Leaves to root:

~~$K := K \bowtie L$   
 $S := S \bowtie T$   
 $S := S \bowtie K$   
 $K := R \bowtie S$~~

-- Root to leaves:

~~$S := S \bowtie R$   
 $T := T \bowtie S$   
 $K := K \bowtie S$   
 $L := L \bowtie K$~~



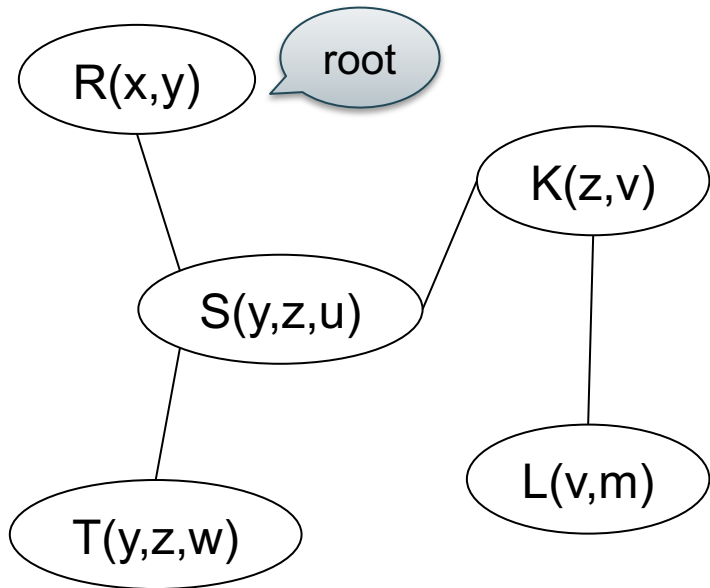
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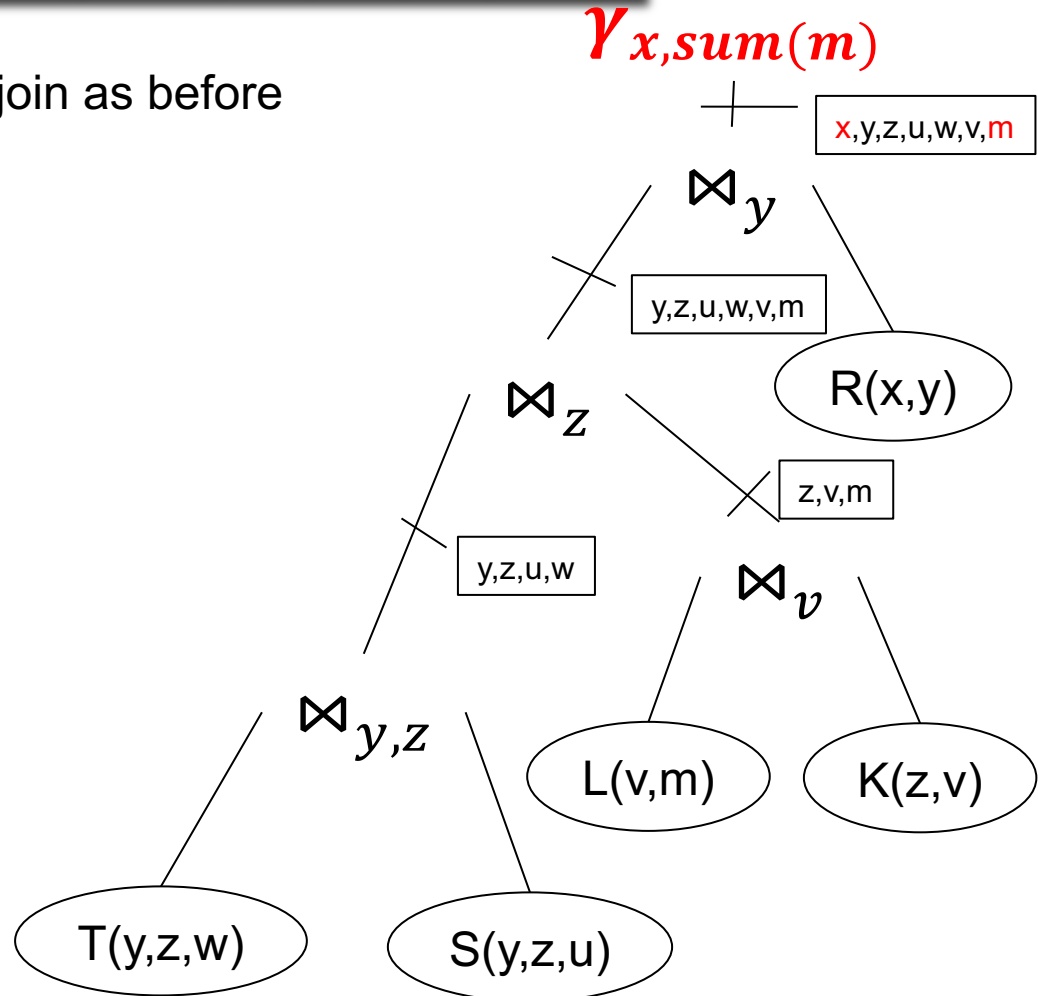


# Example: CQ with Aggregates

$Q(x, \text{sum}(m)) = R(x,y), S(y,z,u), T(y,z,w), K(z,v), L(v,m)$

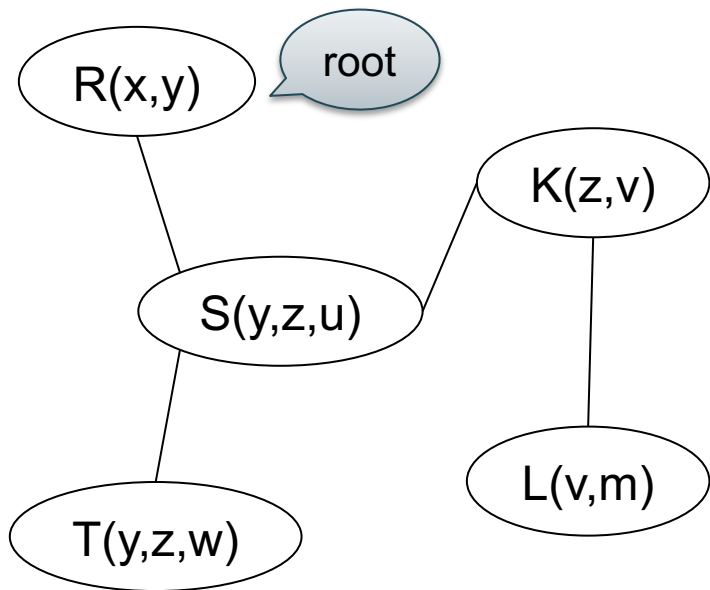


Semi-join as before

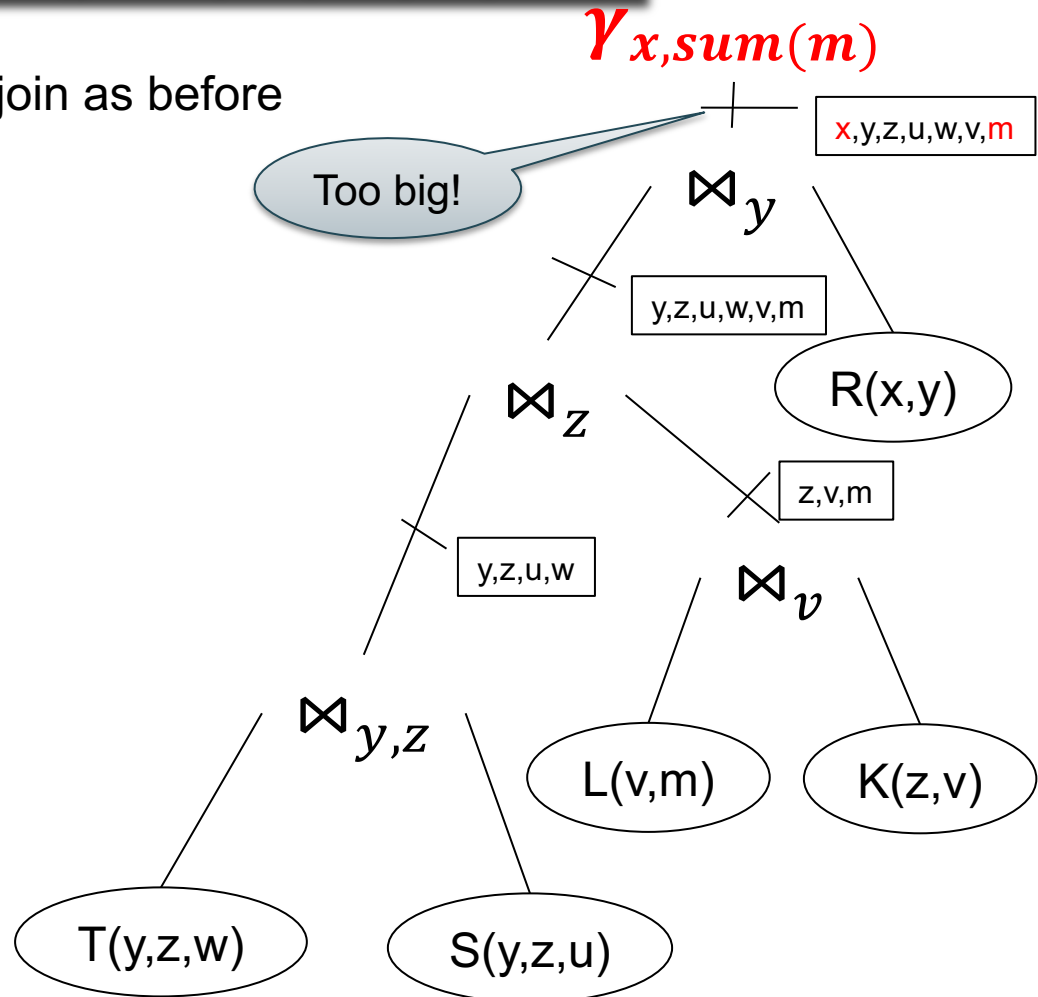


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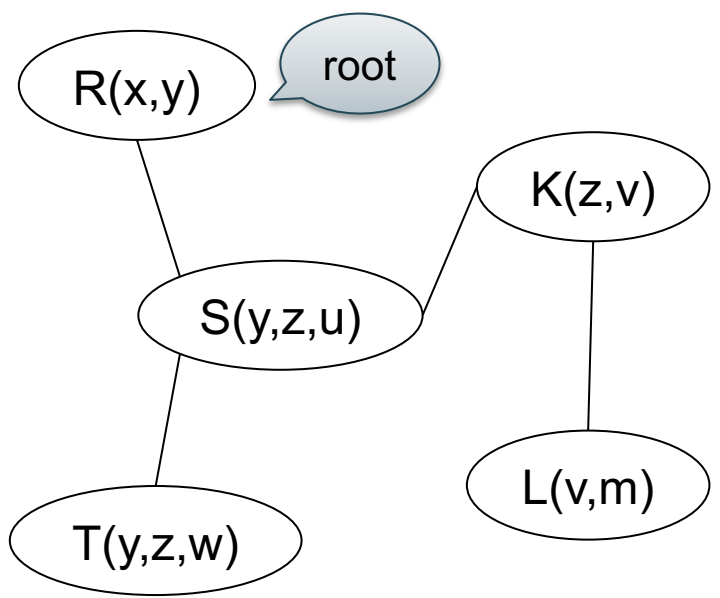


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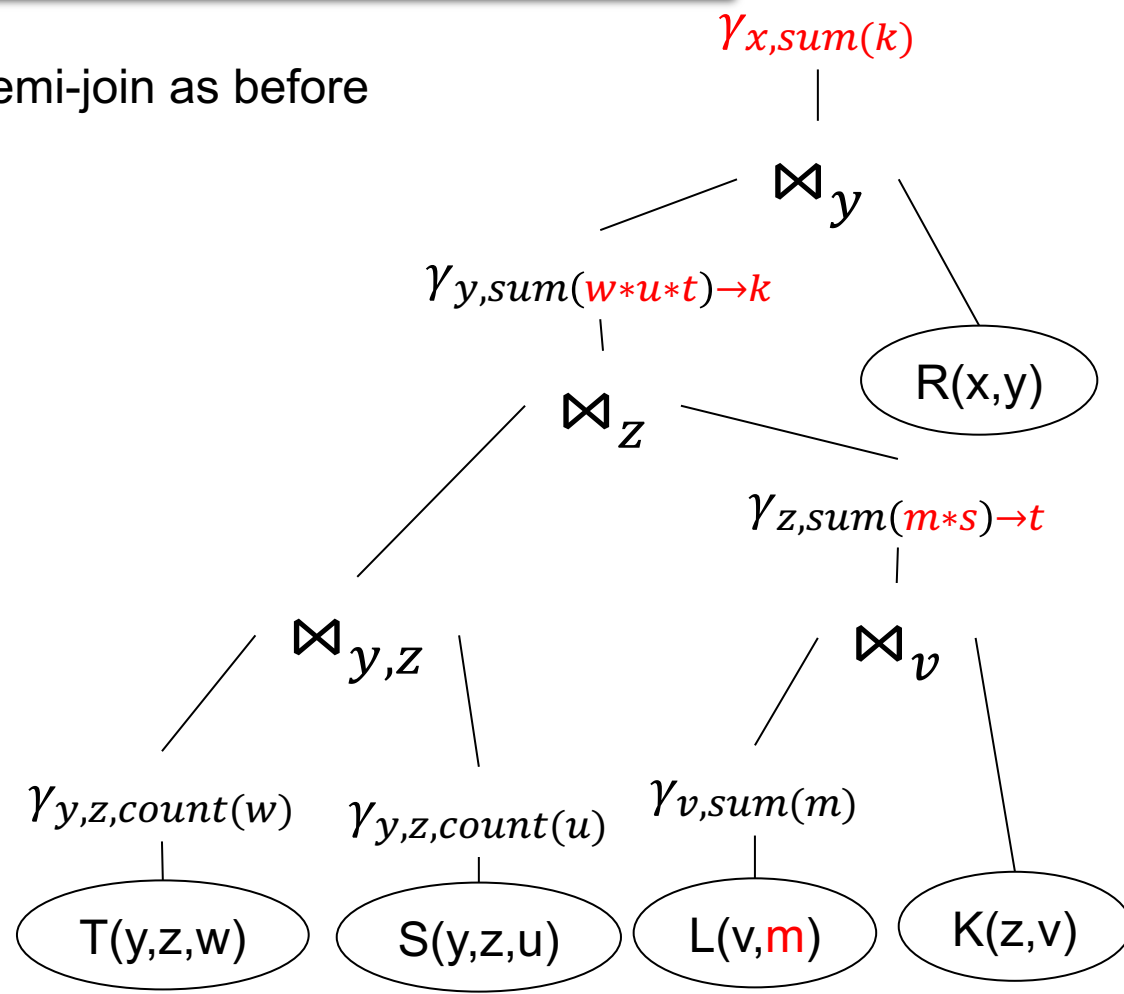


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$$Q(x, \text{sum}(m)) = R(x,y), S(y,z,u), T(y,z,w), K(z,v), L(v,m)$$

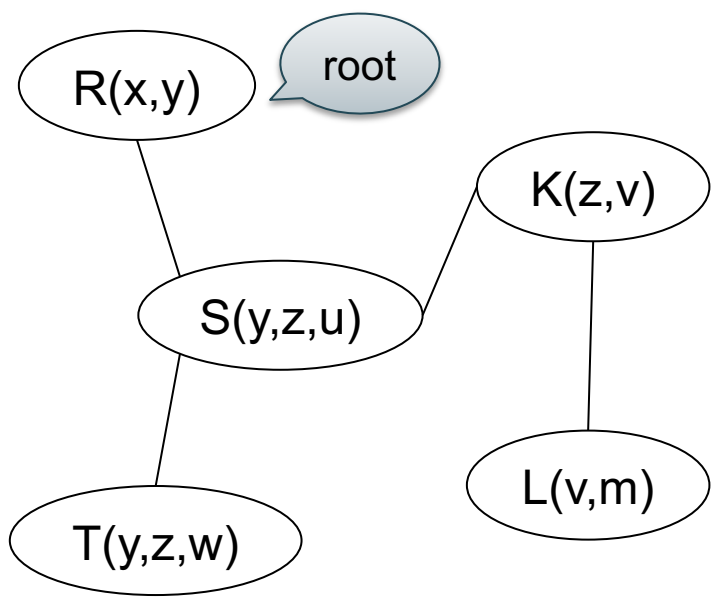


Semi-join as before

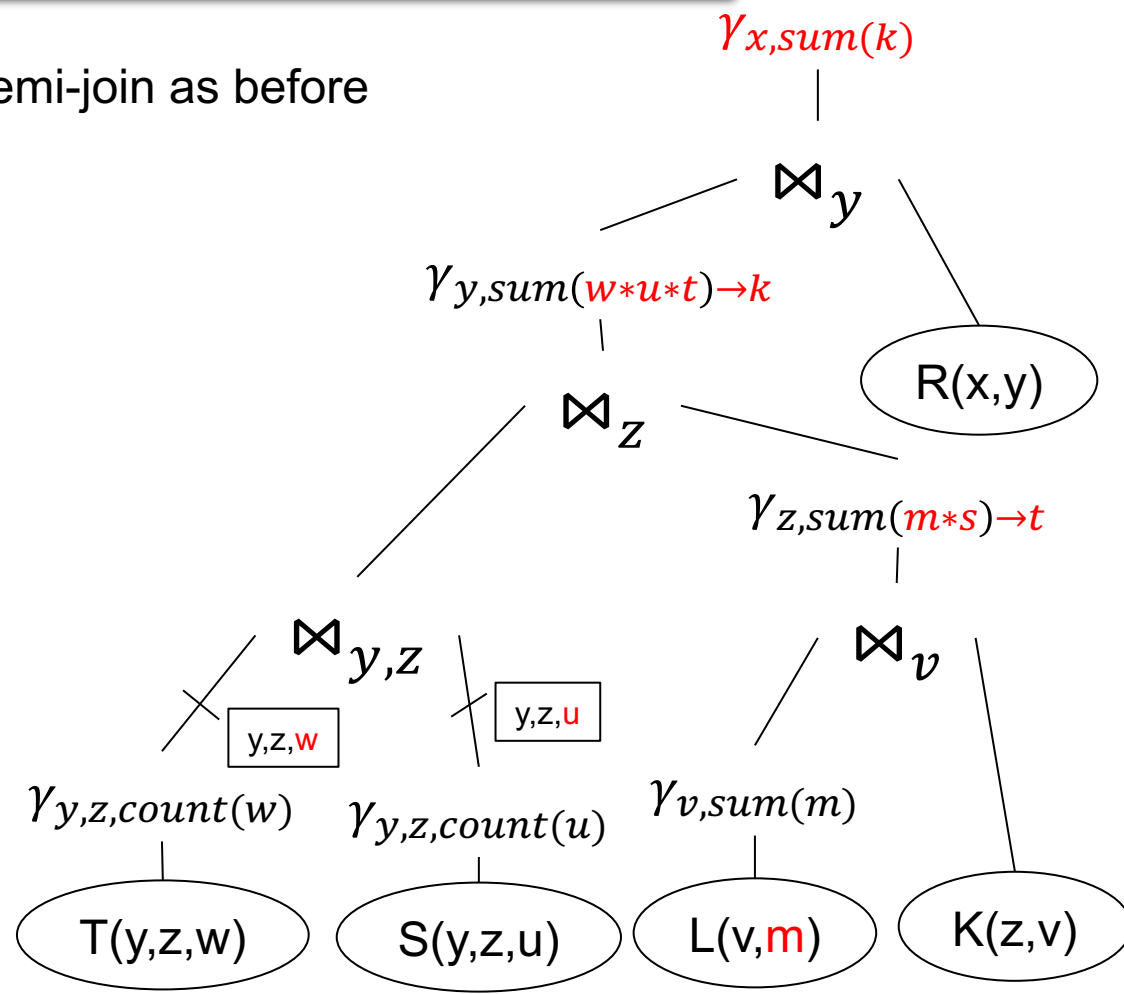


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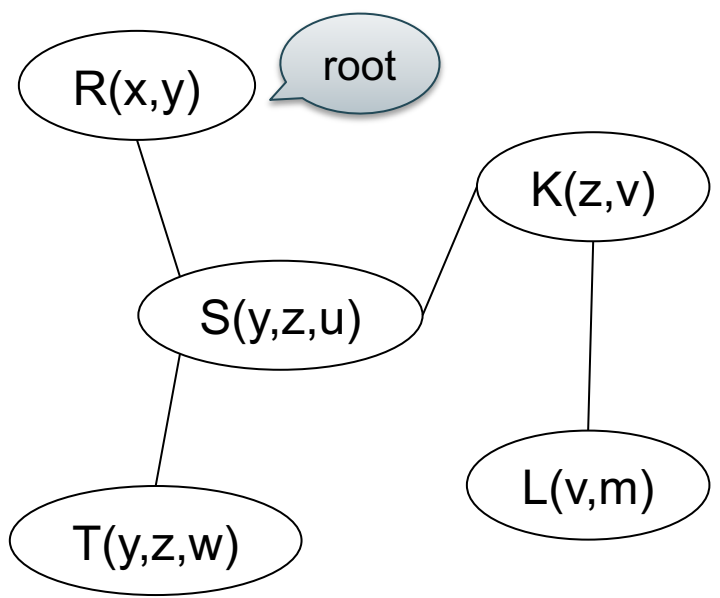


Semi-join as before

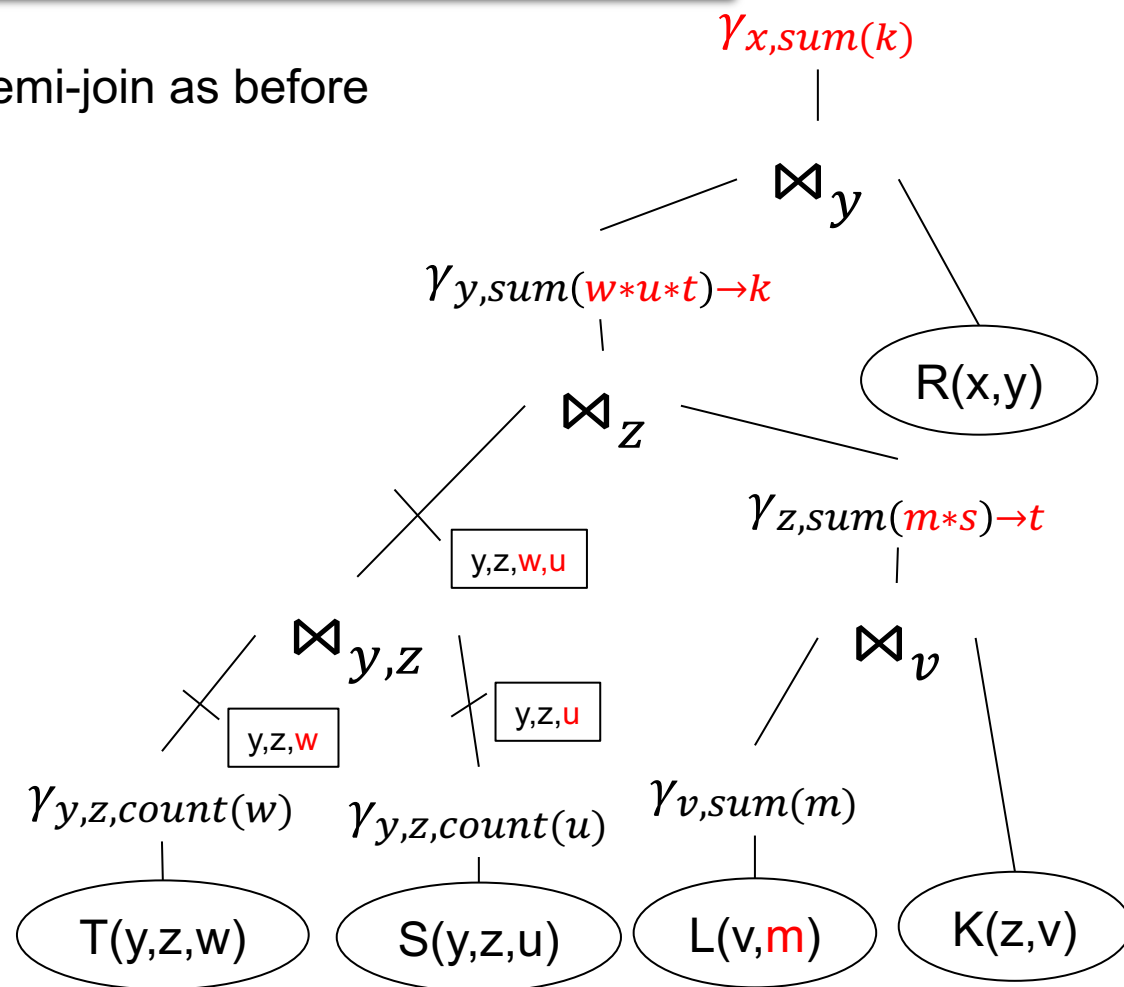


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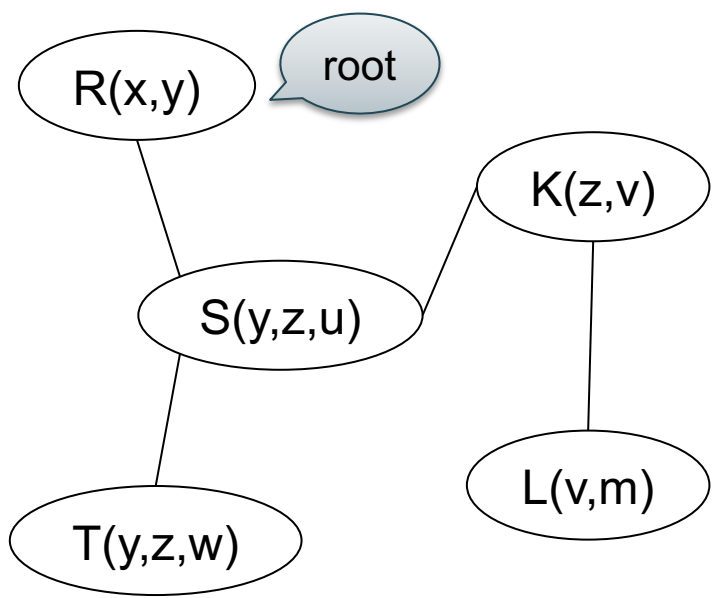


Semi-join as before

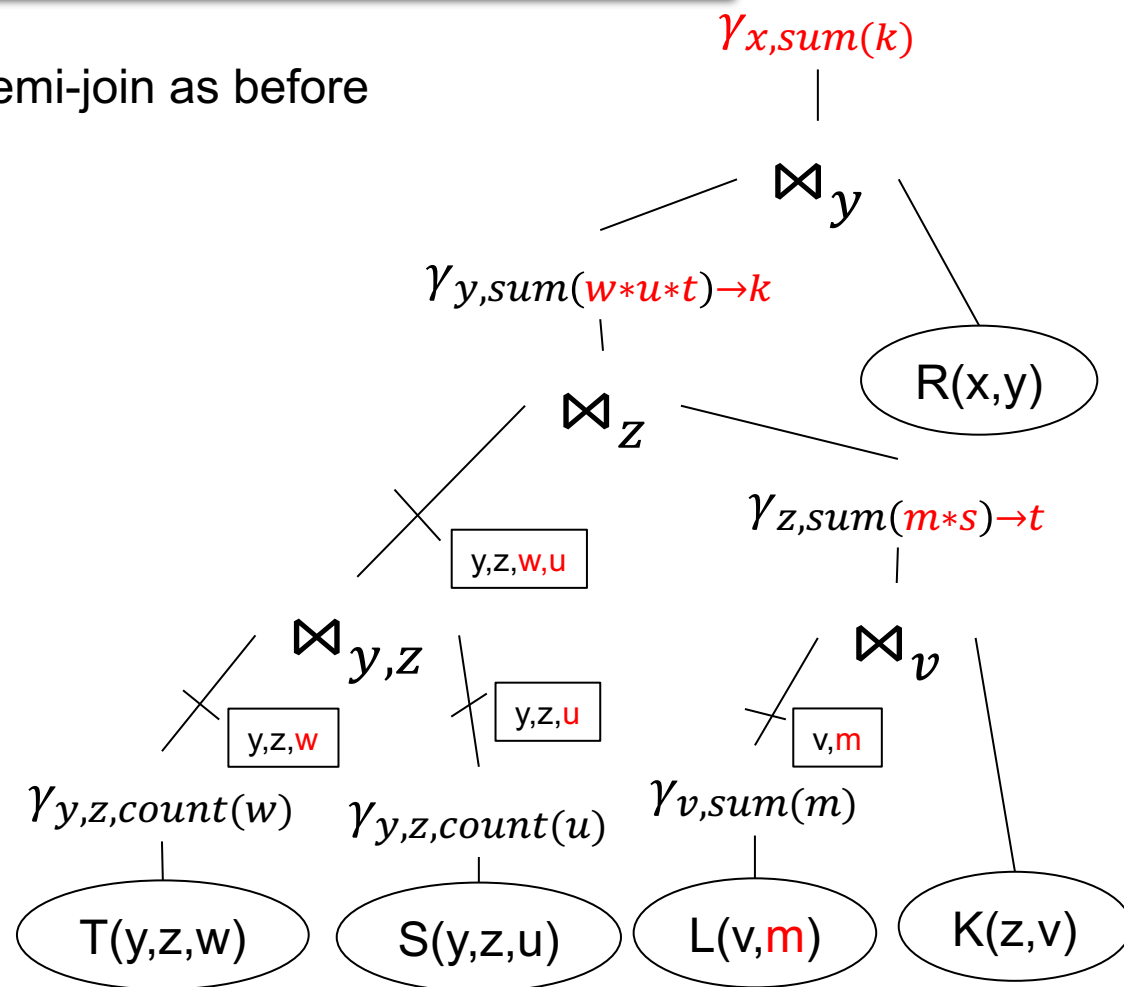


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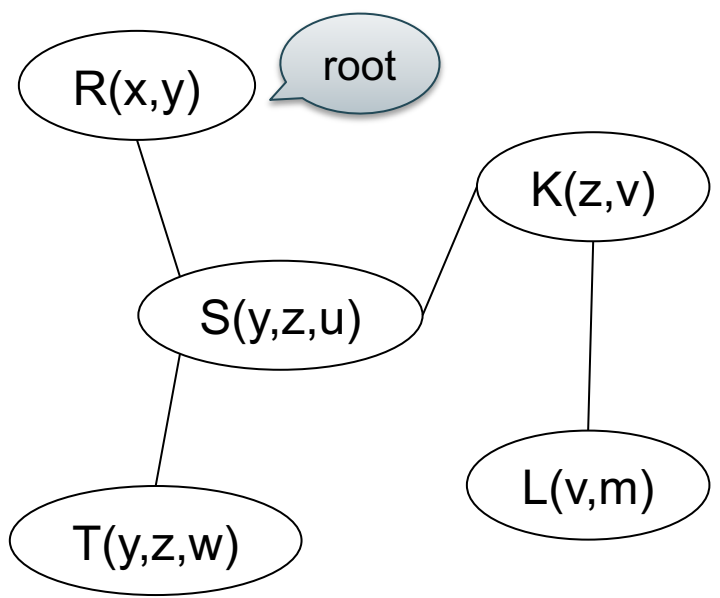


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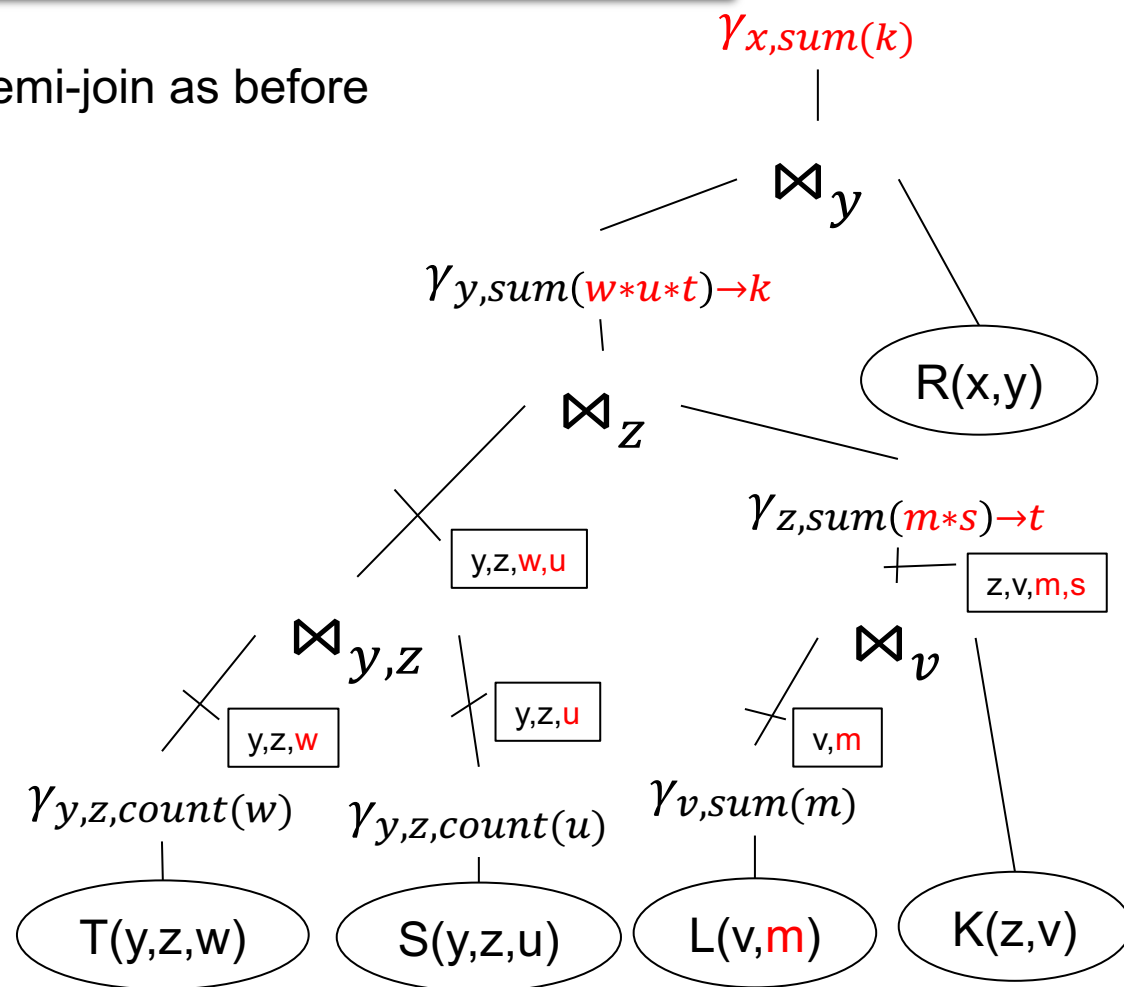


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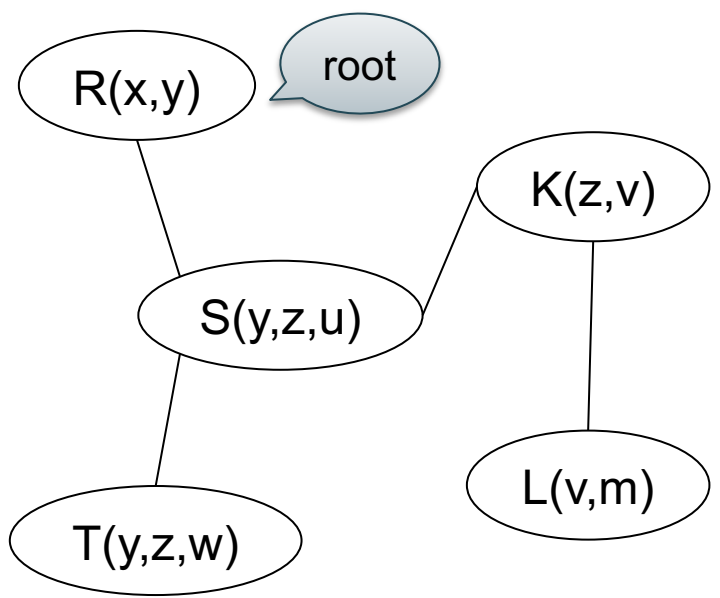


Semi-join as before

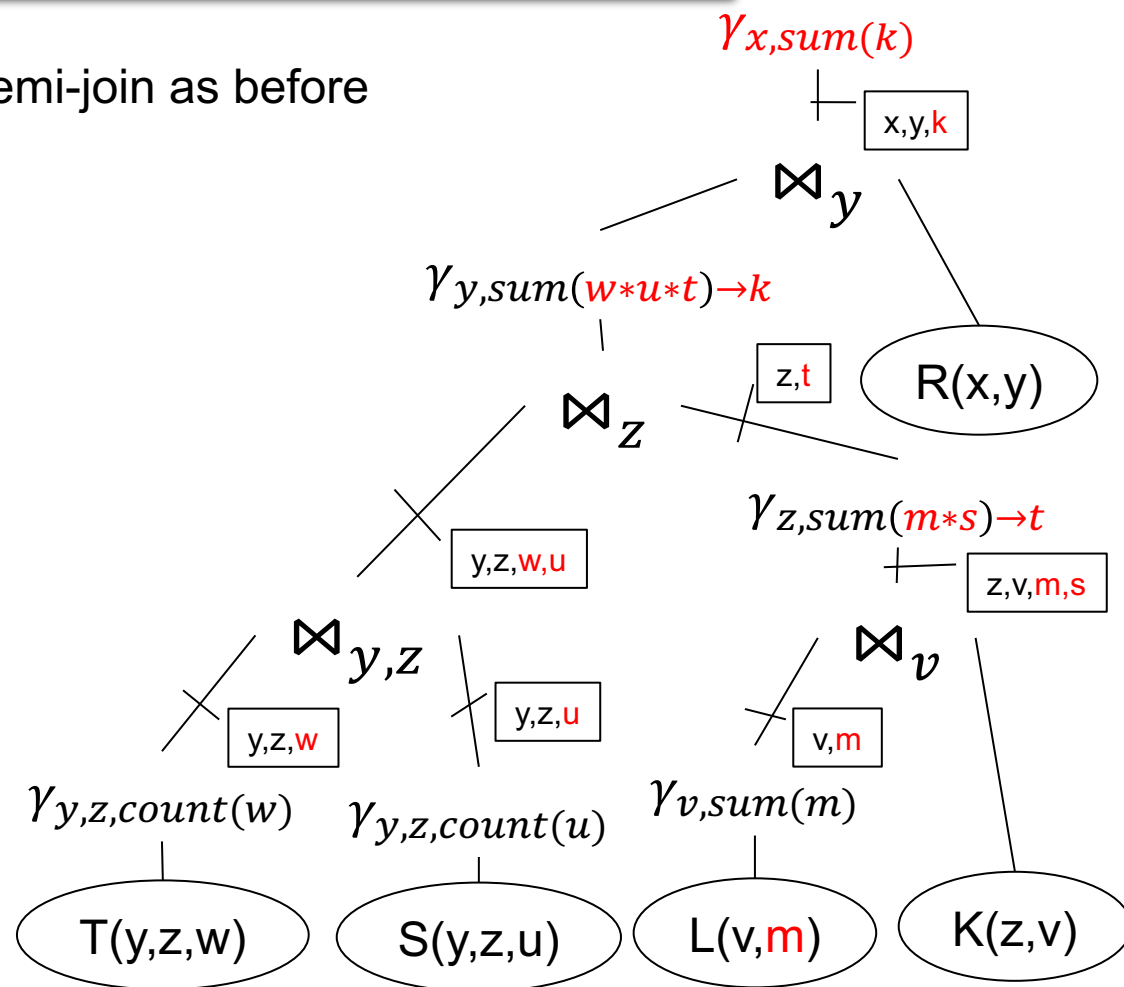


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$$Q(x, \text{sum}(m)) = R(x,y), S(y,z,u), T(y,z,w), K(z,v), L(v,m)$$



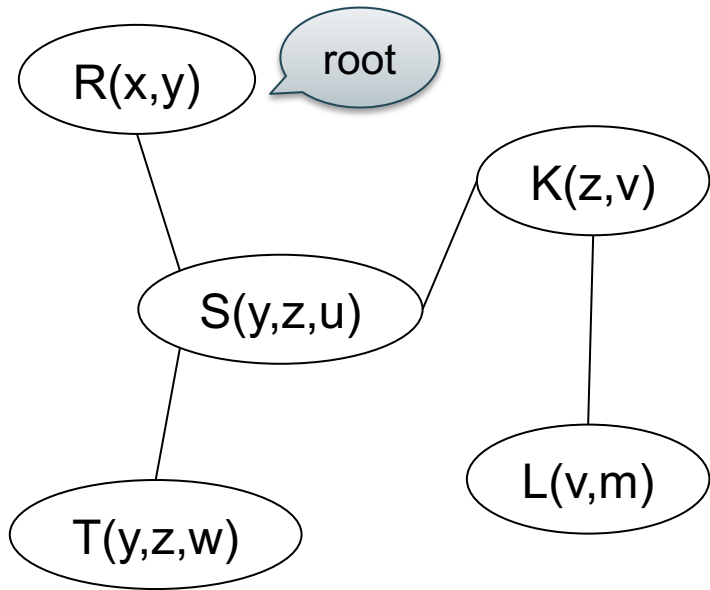
Semi-join as before



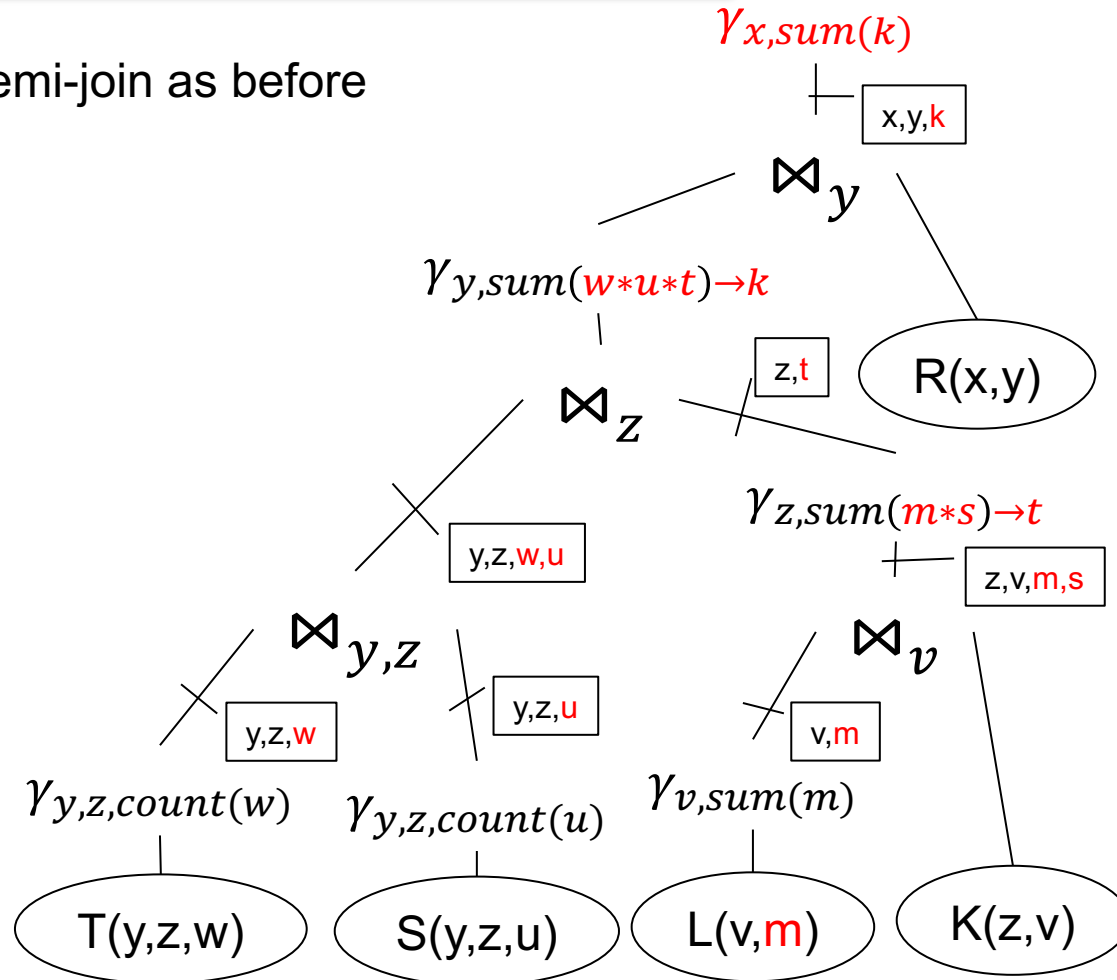


# Example: CQ with Aggregates

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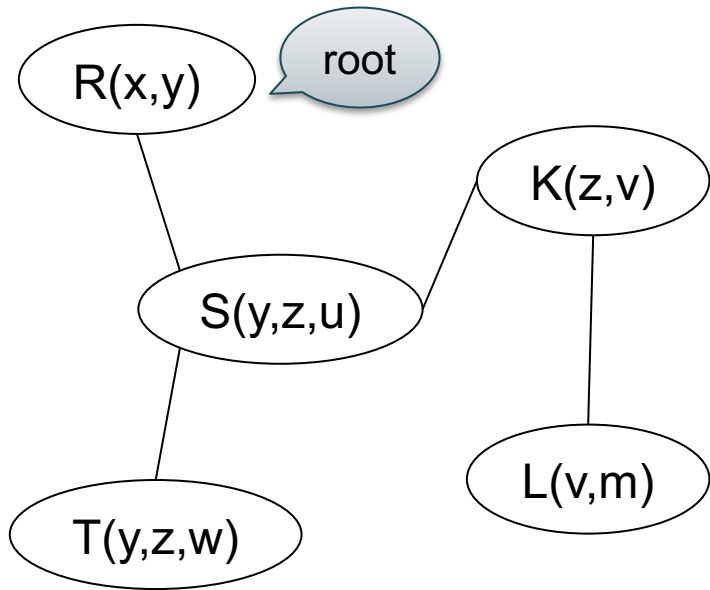


Runtime:

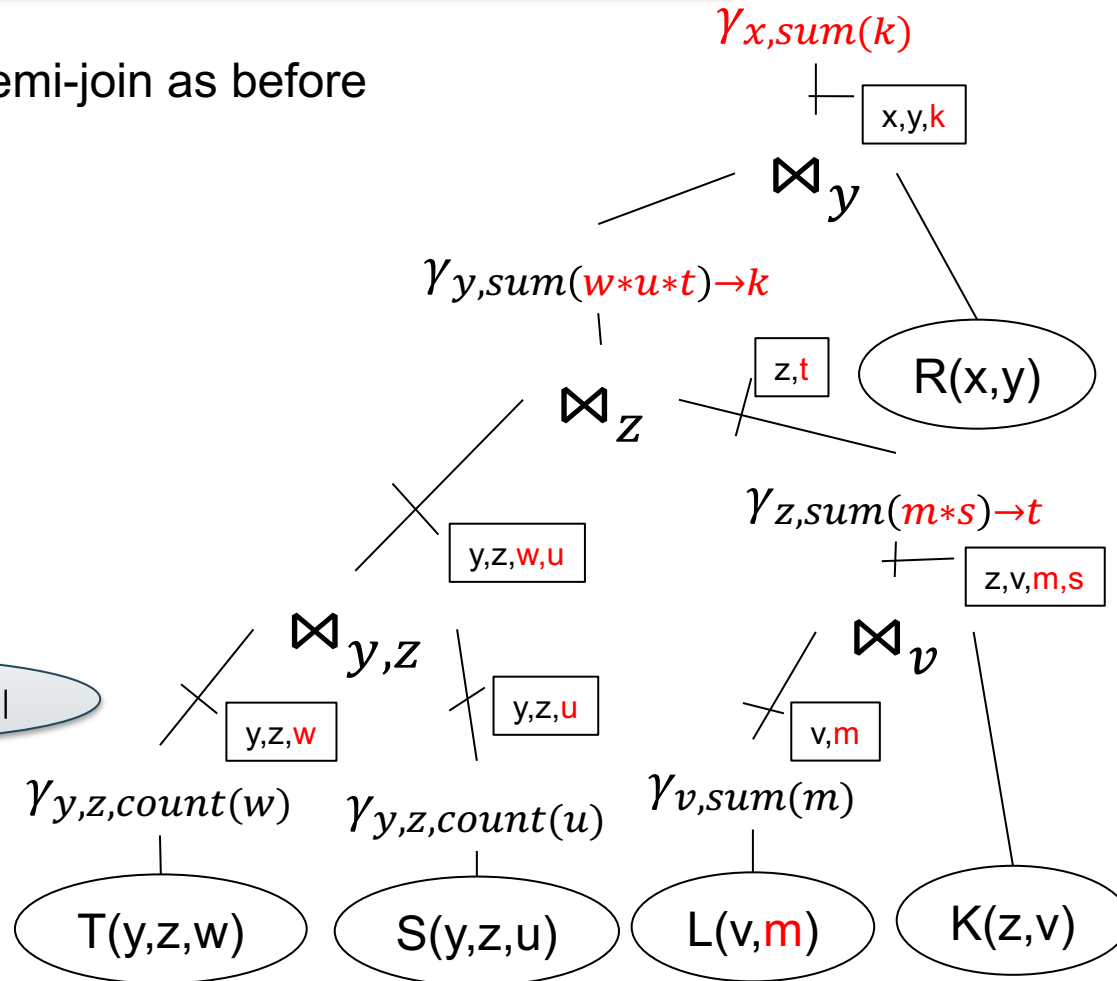
- Semi-joins:  $\tilde{O}(|Input|)$
- Join/group-by:  $\tilde{O}(|Input|)$

# Example: CQ with Aggregates

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Semi-join as before



Runtime:

- Semi-joins:  $\tilde{O}(|Input|)$
- Join/group-by:  $\tilde{O}(|Input|)$

# Discussion

What about group-by multiple attributes?

$Q(x,z,m,\text{sum}(w)) :- \dots$

- Can apply the same principle, but runtime may be polynomial in Input or Output

# Discussion

What is the query is disconnected?

```
SELECT count(*)  
FROM Author, Publication;
```

```
SELECT x.firstName, y.year, count(*)  
FROM Author x, Publication y  
GROUP By x.firstName, y.year;
```

- Simply compute each connected component separately, then take their cartesian product, or regular product, as needed.

# Discussion

Which join order do we choose?

- Yannakakis algorithm doesn't specify:  
*any* join order ensures runtime is:  
$$\tilde{O}(|Input| + |Output|)$$
- BUT: join order may impact the constant significantly, and in practice that matters

# Discussion

- Some acyclic queries have more than one join tree, and each tree has several join orders
- Example:  $Q(x) = R(x), S(x), T(x)$



# Discussion

- Database optimizers rarely do semi-join reduction
  - When they do, they sometimes call it a *magic set optimization* (we'll explain next)
- Reason: when semi-join is ineffective, then it increases cost by a factor of 3

# Discussion

- Magic set optimizations
- Semi-join reductions can also be applied to recursive datalog program
- Called *magic set optimizations*; quite complicated



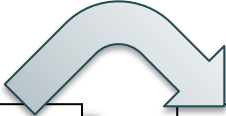
# Discussion

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```
T(x,y) :- Parent(x,y)
T(x,y) :- T(x,z),Parent(z,y)
Q(y)   :- T('Alice',y)
```

# Discussion

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T(x,y) :- Parent(x,y)
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Q(y) :- T('Alice',y)
```

```
Q(y) :- Parent('Alice',y)
Q(y) :- Q(x),Parent(x,y)
```

# Discussion

- A full reducer for Q is a sequence of semi-joins after which every tuple contributes to at least one answer
- **Theorem.** Q has a full reducer iff it is acyclic
- **Proof:** if Q is acyclic, then Yannakakis' algorithm.  
If Q is cyclic, assume w.l.o.g.:

$$Q = R(A,B) \bowtie S(B,C) \bowtie T(A,C)$$

(show in class that it has no full reducer)

# Testing if Q is Acyclic

An ear of Q is an atom  $R(X)$  with the following property:

- Let  $X' \subseteq X$  be the set of join variables (occurring some other atom)
- There exists some other atom  $S(Y)$  such that  $X' \subseteq Y$

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GYO algorithm (Graham, Yu, Özsoyoğlu) for acyclicity:

- While Q has an ear  $R(X)$ , remove  $R(X)$  from Q
- If all atoms were removed, then Q is acyclic
- If atoms remain but there is no ear, then Q is cyclic

# Outline

- Acyclic queries, Yannakakis algorithm
- Tree decomposition of cyclic queries
- Worst-case optimal algorithm; next week

# Tree Decomposition

**Def** *Tree decomposition* is  $(T, \chi)$ ,  $\chi: \text{Nodes}(T) \rightarrow 2^{\text{Vars}(Q)}$  s.t.:

- (1)  $\forall A \in \text{Atoms}(Q) \exists t \in \text{Nodes}(T), \text{Vars}(A) \subseteq \chi(t)$
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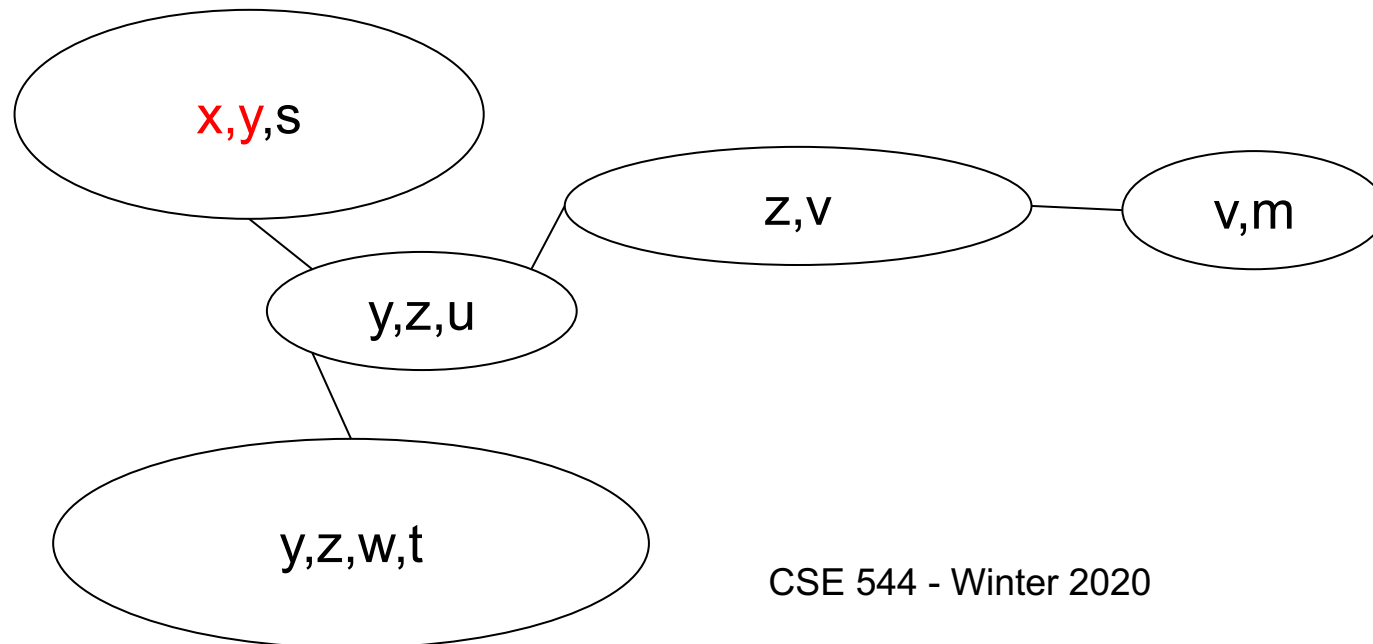
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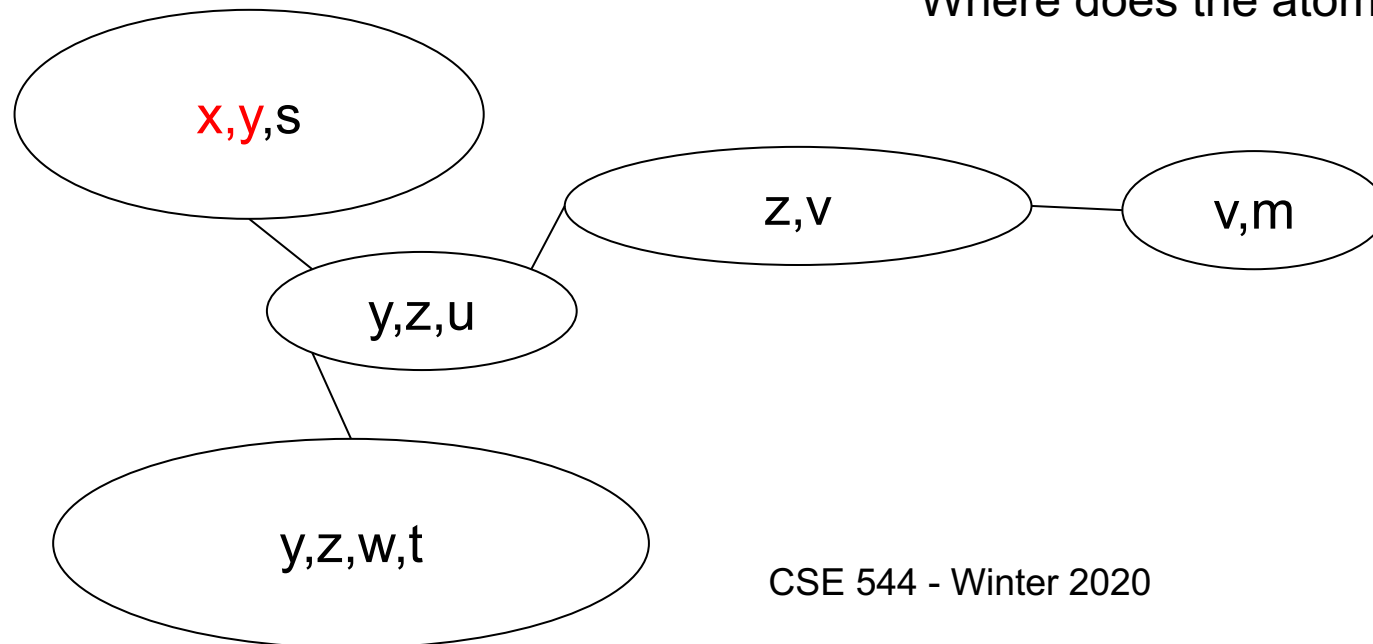
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Where does the atom  $R(x, y)$  occur?



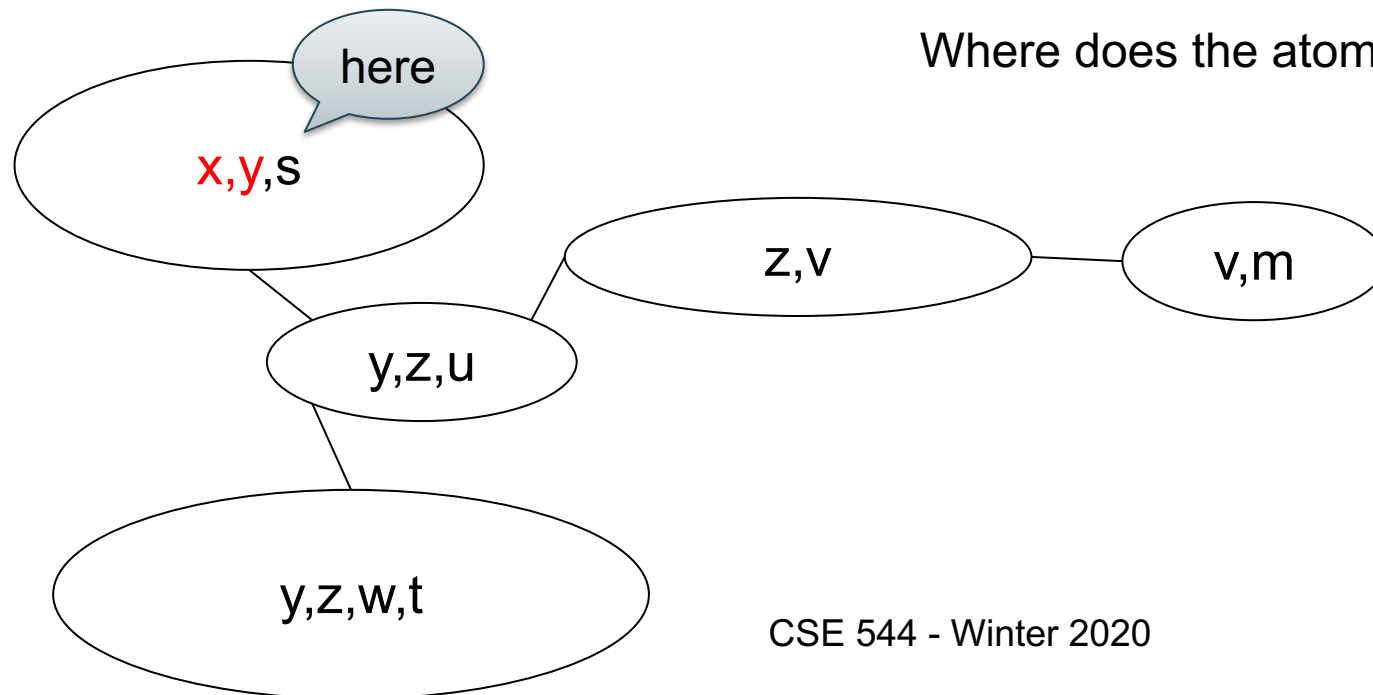
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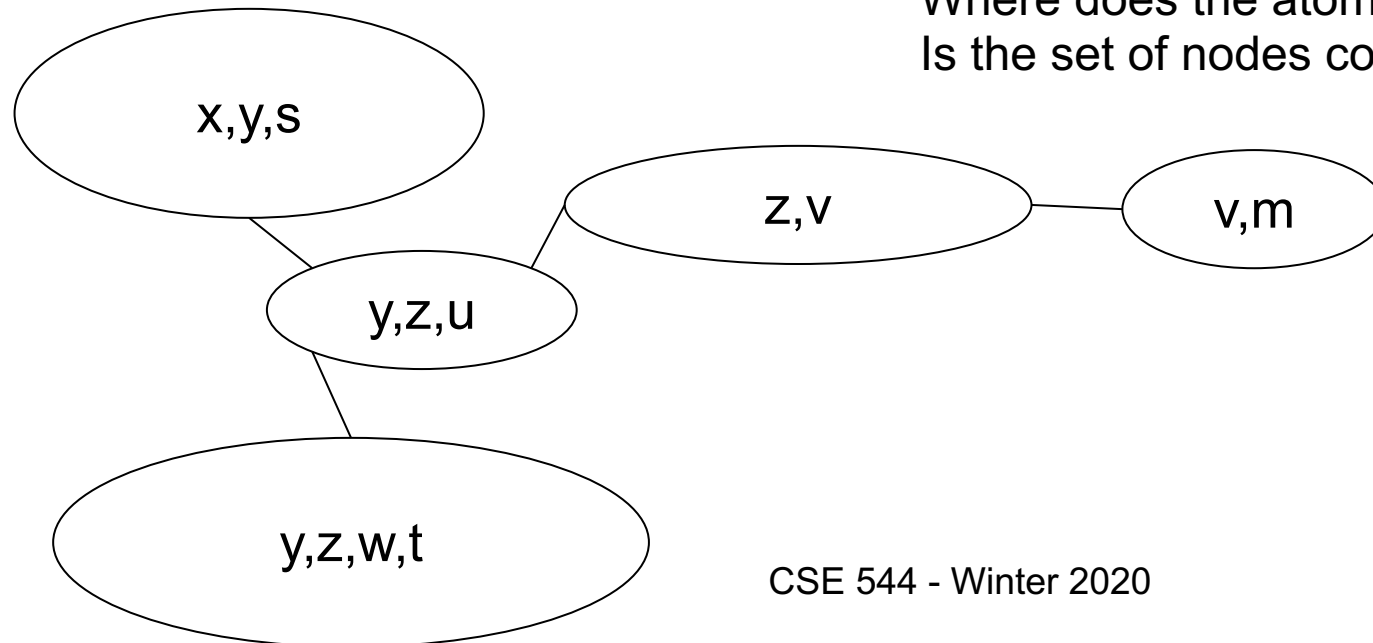
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Where does the atom  $R(x, y)$  occur?  
Is the set of nodes containing  $z$  connected?



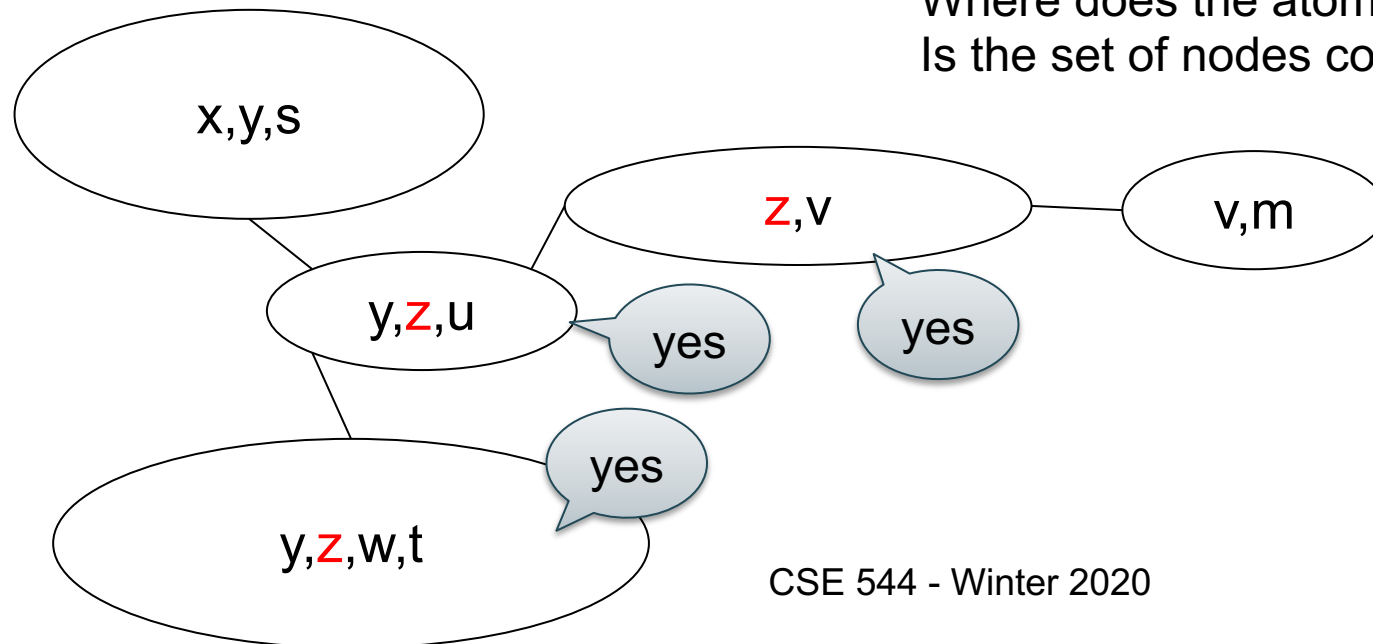
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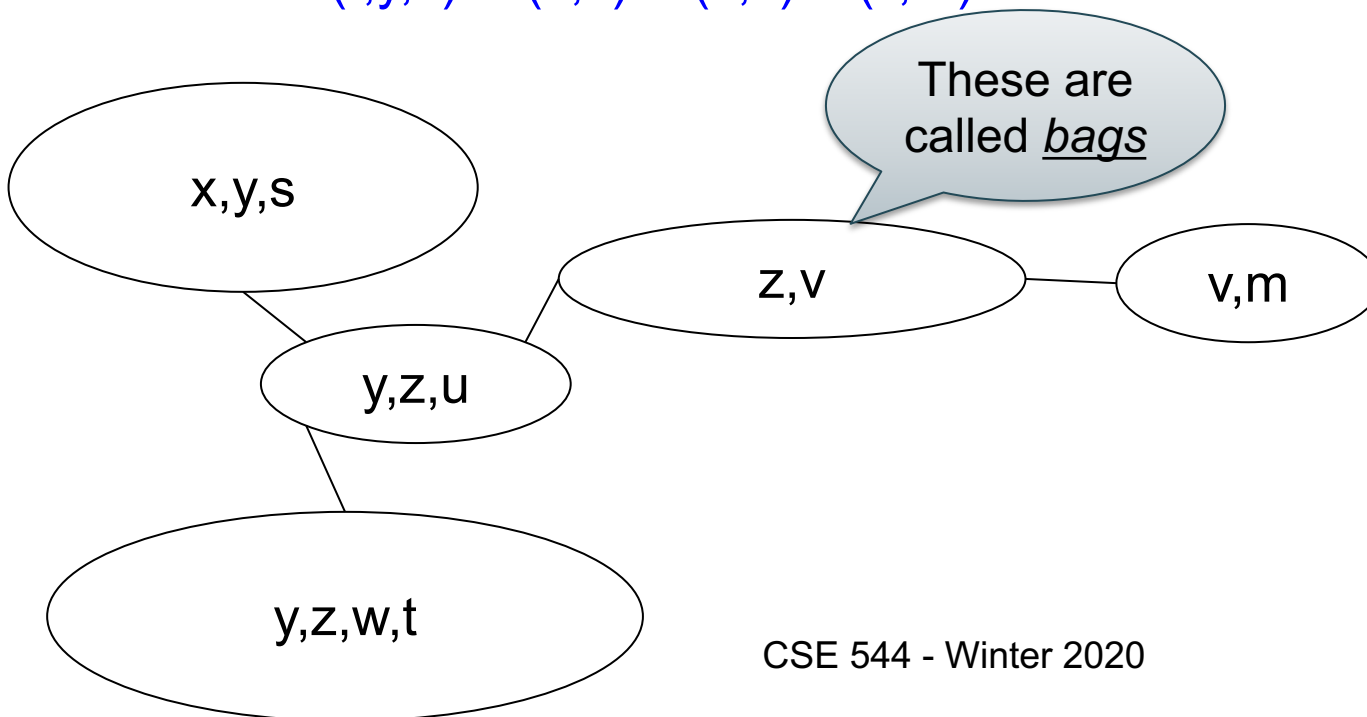


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full CQ:  $Q_t(x, y, s) = R(x, y) \wedge A(y, s) \wedge B(x, s)$

x, y, s

y, z, u

y, z, w, t

z, v

v, m

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$S(y, z, u)$

$K(z, v), F(z, v)$

$L(v, m)$

$T(y, z, w), C(z, w, t)$   
 $D(w, t, y), E(t, y, z)$



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$L(v, m)$

$T(y, z, w), C(z, w, t)$   
 $D(w, t, y), E(t, y, z)$

**Computing  $Q(D)$ :**

(1) Compute all full CQ's  $Q_t$

(2) Run Yannakakis' on the join tree

Time  $O(N^{??} + |\text{Output}|)$

# Recap

To compute a query  $Q$  proceed as follows

1. Find a tree decomposition of  $Q$
2. For each tree node (“bag”) compute its local query  $Q_t$
3. Run Yannakakis on the resulting acyclic query

Runtime is dominated step 2

Will discuss step 2 next

# Tree-width

$$\text{Def } \text{tw}(Q) = \min_T \max_{t \in \text{Nodes}(T)} |X(t)| - 1$$

# Tree-width

This is the  
standard definition  
in graph theory

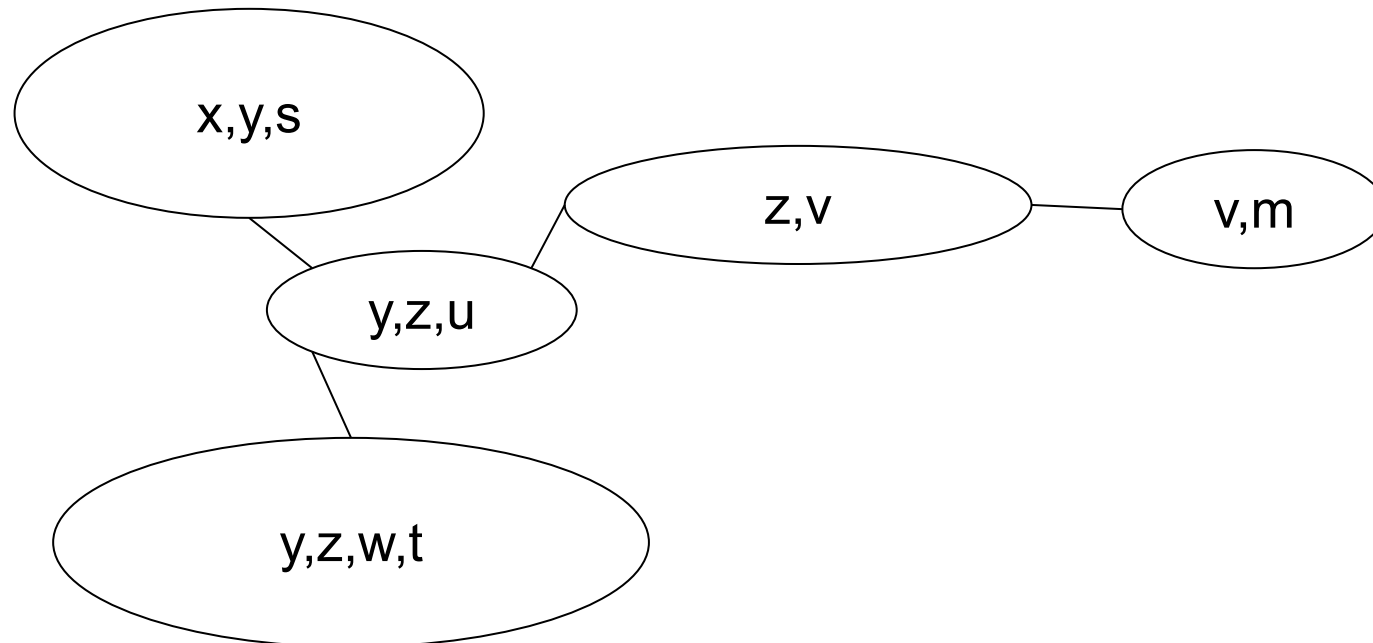
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$$\text{tw}(Q) = 3$$

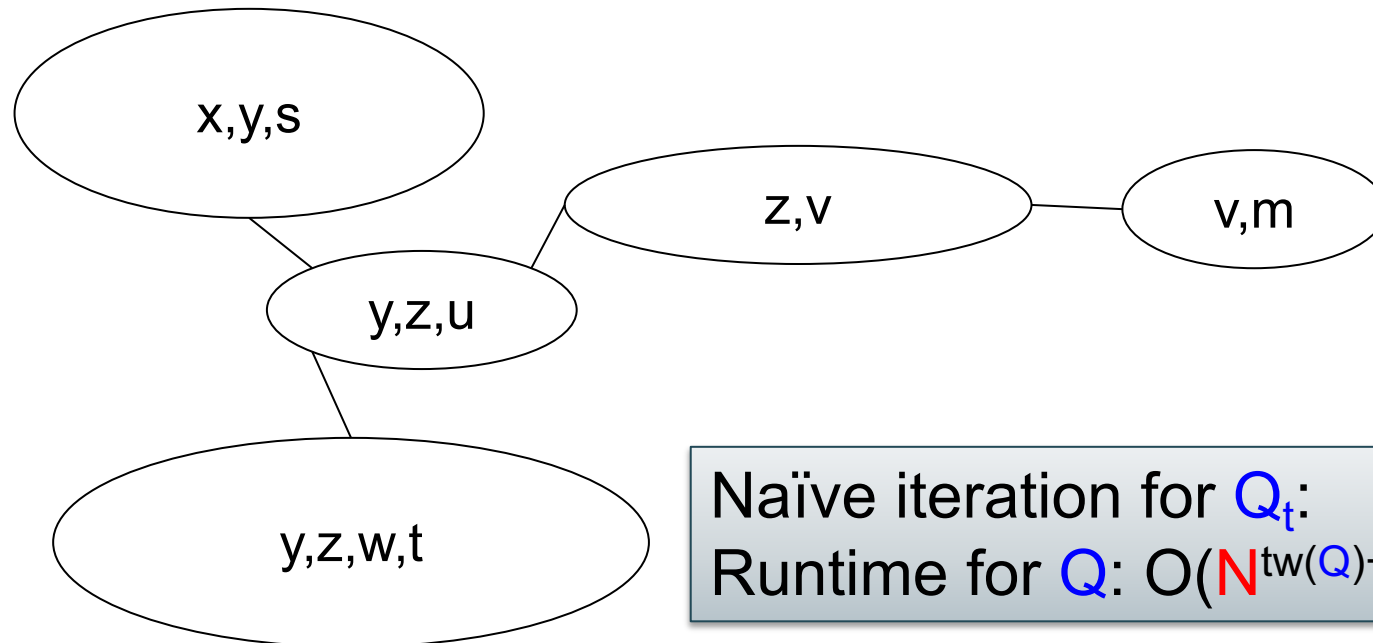


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This is the standard definition in graph theory

$$\text{Def } \text{tw}(Q) = \min_T \max_{t \in \text{Nodes}(T)} |X(t)| - 1$$

$$\text{tw}(Q) = 3$$



Naïve iteration for  $Q_t$ :  
Runtime for  $Q$ :  $O(N^{\text{tw}(Q)+1} + |\text{Output}|)$

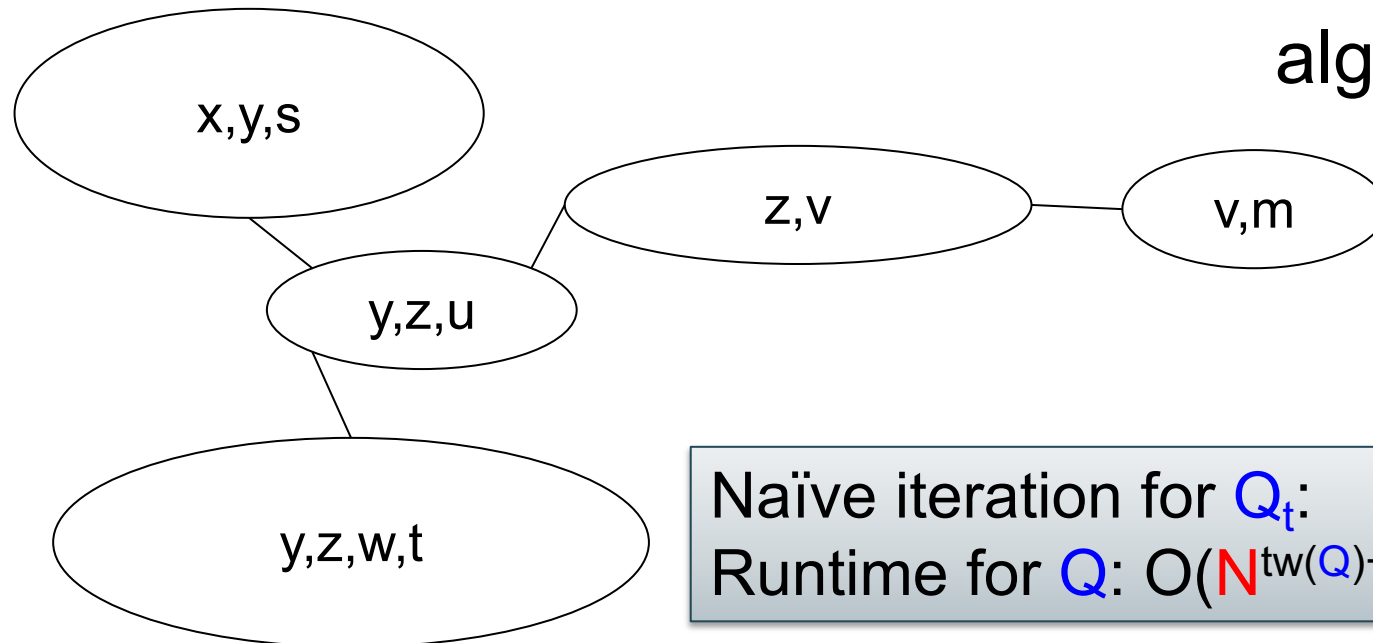
# Tree-width

This is the standard definition in graph theory

$$\text{Def } \text{tw}(Q) = \min_T \max_{t \in \text{Nodes}(T)} |X(t)| - 1$$

$$\text{tw}(Q) = 3$$

*Tree-width* gives the complexity of the most naïve algorithm



Naïve iteration for  $Q_t$ :  
Runtime for  $Q$ :  $O(N^{\text{tw}(Q)+1} + |\text{Output}|)$

# Generalized Hypertree Width

$$\mathbf{Def} \text{ ghtw}(Q) = \min_T \max_{t \in \text{Nodes}(T)} \rho(Q_t)$$

$\rho$  = edge covering  
number



# Generalized Hypertree Width

$$\text{Def } \text{ghtw}(Q) = \min_T \max_{t \in \text{Nodes}(T)} \rho(Q_t)$$

$\rho$  = edge covering number

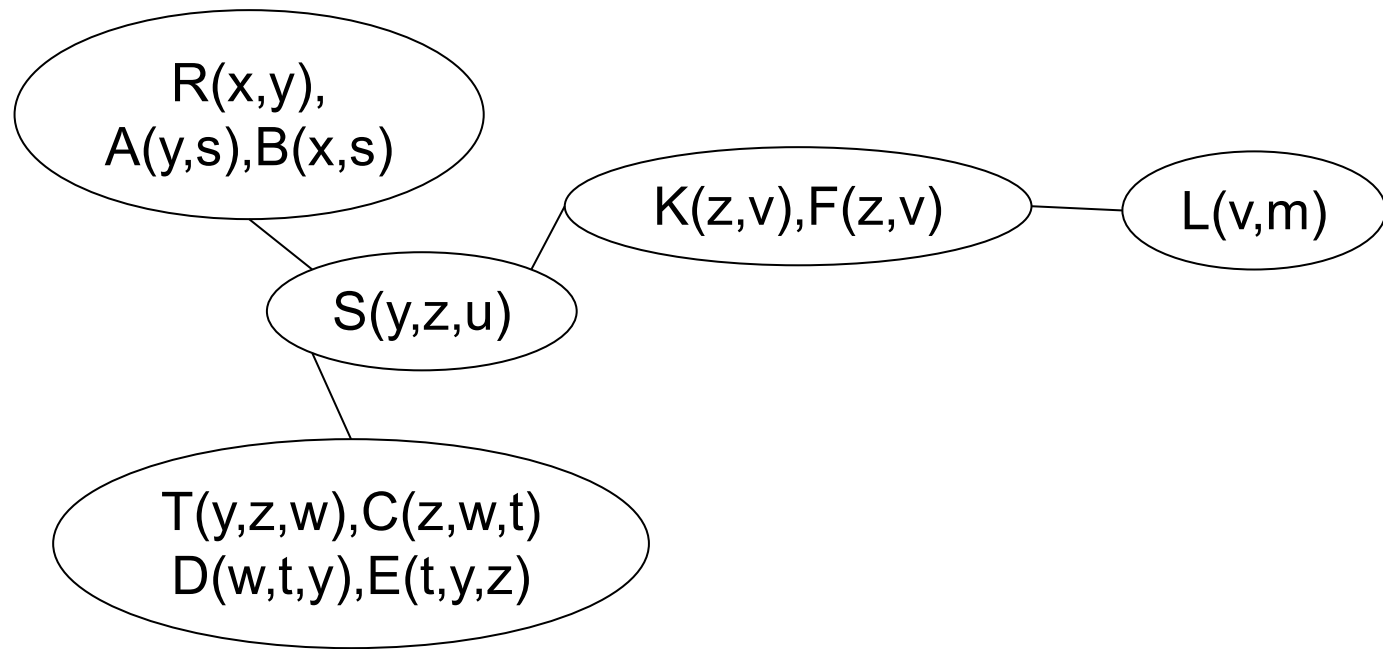
$R(x,y),$   
 $A(y,s),B(x,s)$

$S(y,z,u)$

$K(z,v),F(z,v)$

$L(v,m)$

$T(y,z,w),C(z,w,t)$   
 $D(w,t,y),E(t,y,z)$



# Generalized Hypertree Width

$$\text{Def } \text{ghtw}(Q) = \min_T \max_{t \in \text{Nodes}(T)} \rho(Q_t)$$

$\rho$  = edge covering number

$R(x,y),$   
 $A(y,s), B(x,s)$

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 $D(w,t,y), E(t,y,z)$

# Generalized Hypertree Width

$$\text{Def } \text{ghtw}(Q) = \min_T \max_{t \in \text{Nodes}(T)} \rho(Q_t)$$

$\rho$  = edge covering number

$\rho=2$

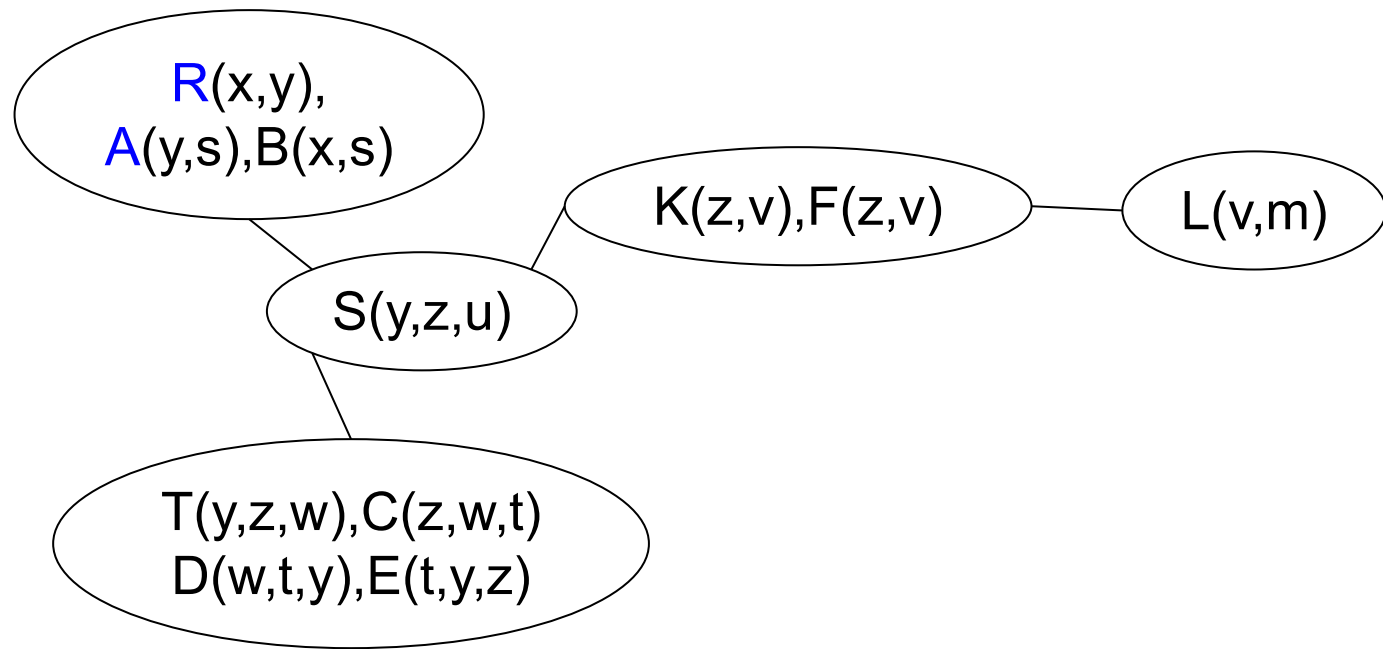
$R(x,y),$   
 $A(y,s),B(x,s)$

$S(y,z,u)$

$K(z,v),F(z,v)$

$L(v,m)$

$T(y,z,w),C(z,w,t)$   
 $D(w,t,y),E(t,y,z)$



# Generalized Hypertree Width

$$\text{Def } \text{ghtw}(Q) = \min_T \max_{t \in \text{Nodes}(T)} \rho(Q_t)$$

$\rho$  = edge covering number

$\rho=2$

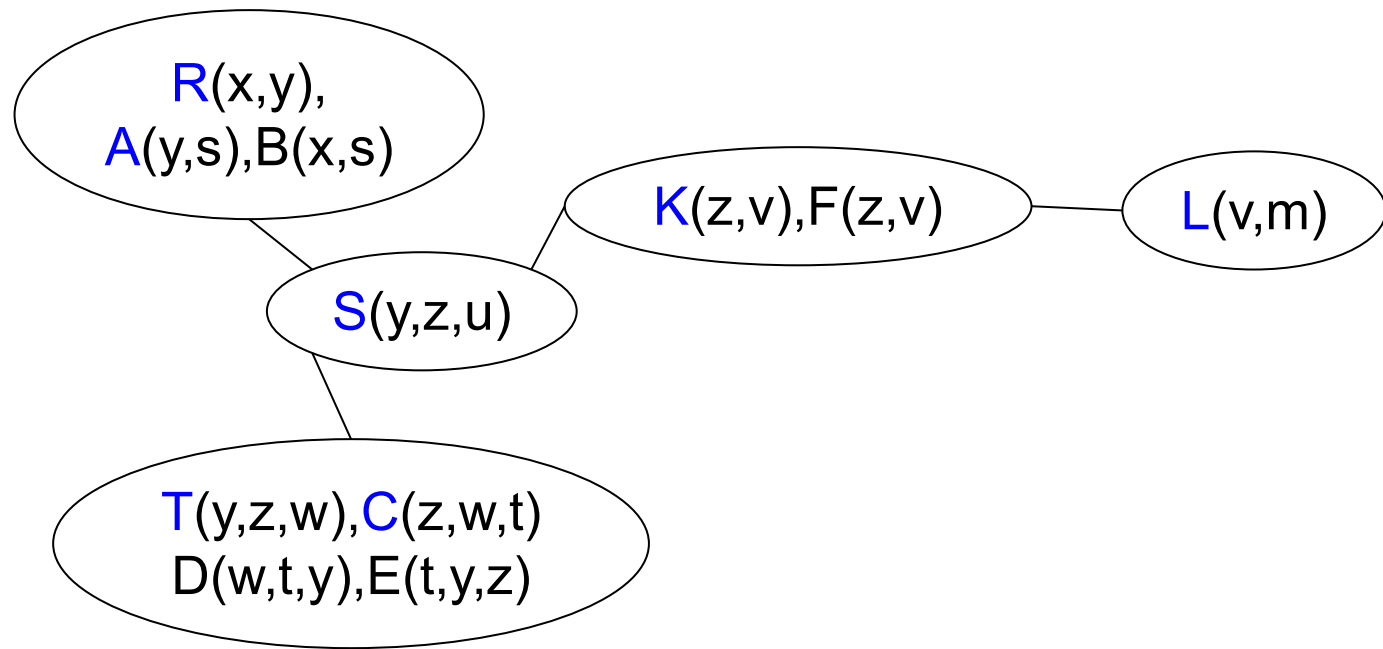
$R(x,y),$   
 $A(y,s), B(x,s)$

$S(y,z,u)$

$K(z,v), F(z,v)$

$L(v,m)$

$T(y,z,w), C(z,w,t)$   
 $D(w,t,y), E(t,y,z)$



# Generalized Hypertree Width

$$\text{Def } \text{ghtw}(Q) = \min_T \max_{t \in \text{Nodes}(T)} \rho(Q_t)$$

$\rho$  = edge covering number

$\rho=2$

$R(x,y),$   
 $A(y,s),B(x,s)$

$S(y,z,u)$

$K(z,v),F(z,v)$

$L(v,m)$

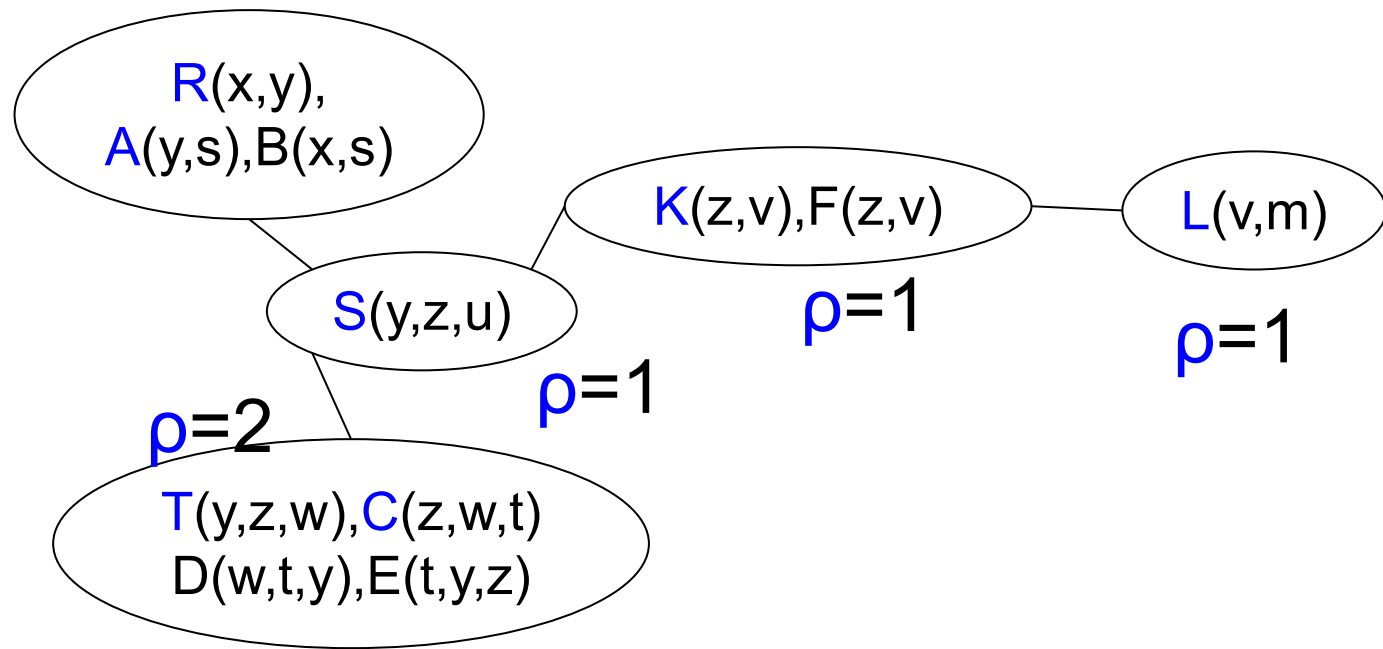
$\rho=1$

$\rho=1$

$\rho=1$

$\rho=2$

$T(y,z,w),C(z,w,t)$   
 $D(w,t,y),E(t,y,z)$



# Generalized Hypertree Width

$$\text{Def } \text{ghtw}(Q) = \min_T \max_{t \in \text{Nodes}(T)} \rho(Q_t)$$

$\rho$  = edge covering number

$$\text{ghtw}(Q) = 2$$

$$\rho=2$$

$R(x,y),$   
 $A(y,s),B(x,s)$

$S(y,z,u)$

$K(z,v),F(z,v)$

$$\rho=1$$

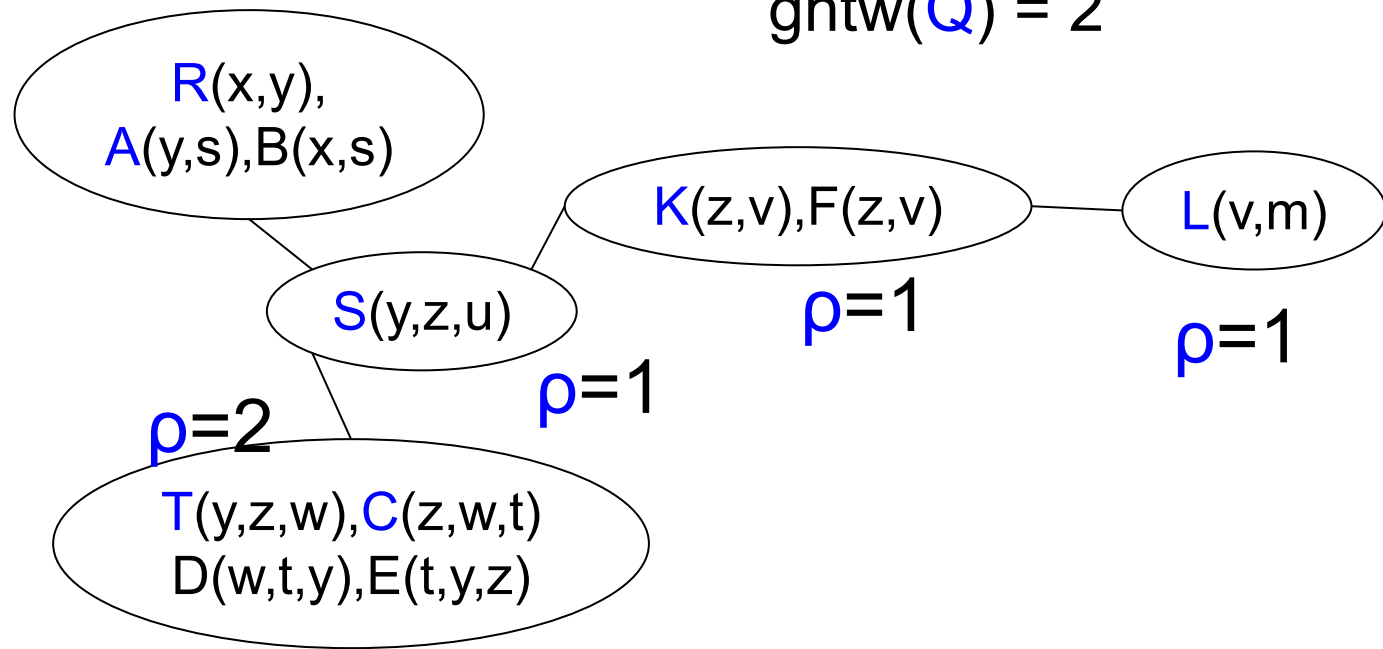
$L(v,m)$

$$\rho=1$$

$$\rho=2$$

$T(y,z,w),C(z,w,t)$   
 $D(w,t,y),E(t,y,z)$

$$\rho=1$$

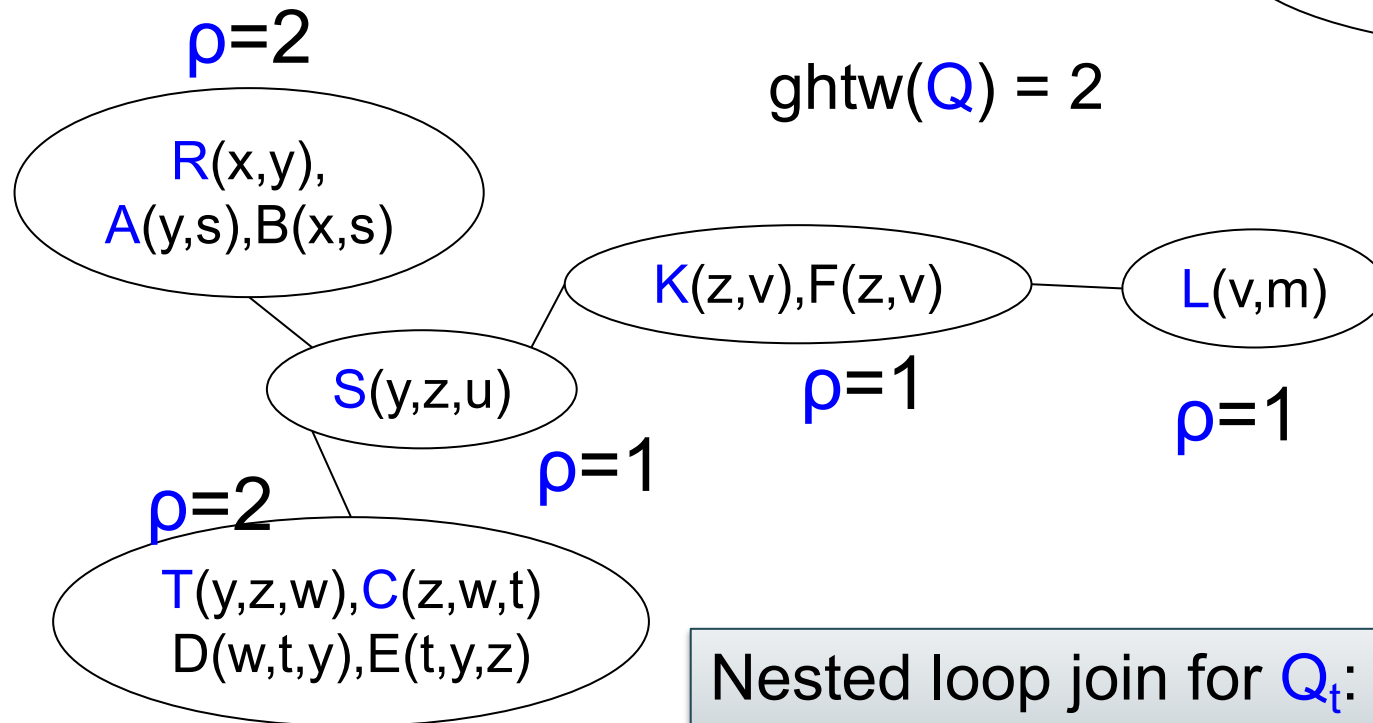


# Generalized Hypertree Width

$$\text{Def } \text{ghtw}(Q) = \min_T \max_{t \in \text{Nodes}(T)} \rho(Q_t)$$

$\rho$  = edge covering number

$$\text{ghtw}(Q) = 2$$



Nested loop join for  $Q_t$ :

Runtime for  $Q$ :  $O(N^{\text{ghtw}(Q)} + |\text{Output}|)$

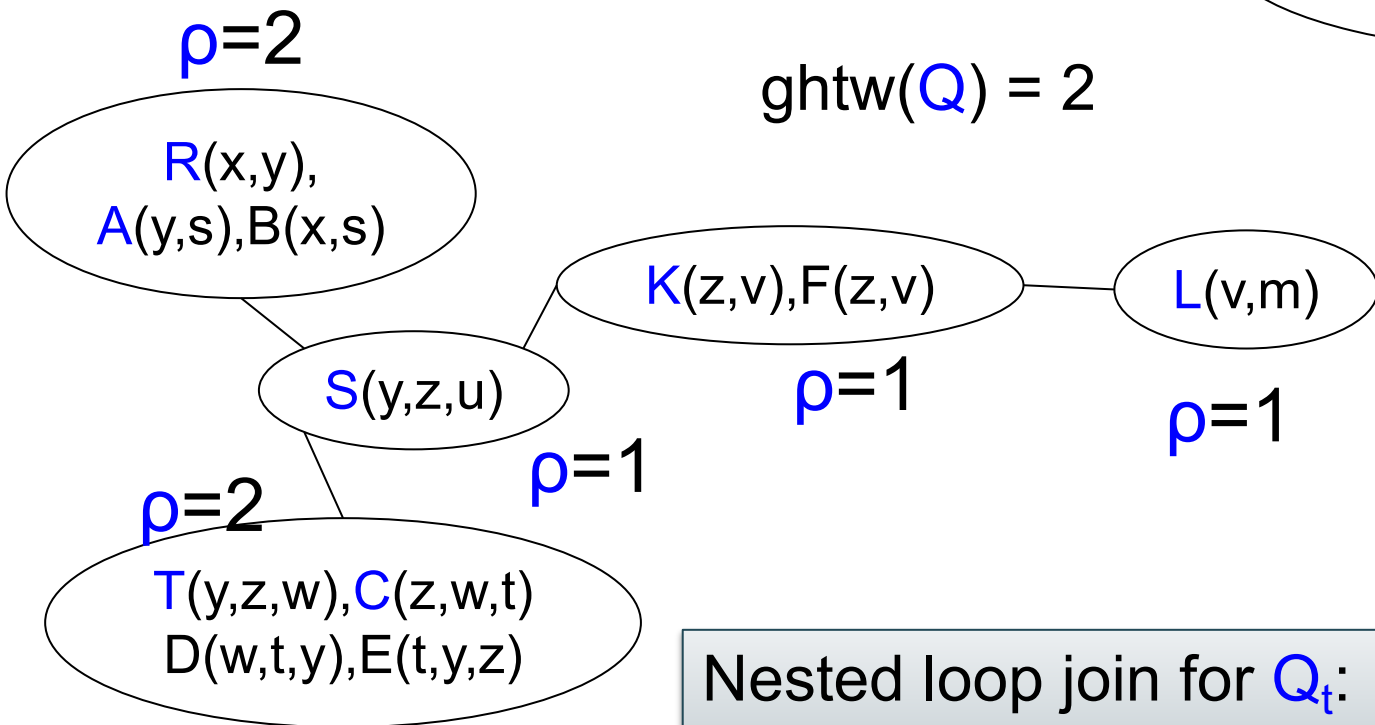
# Generalized Hypertree Width

**Def**  $ghtw(Q) = \min_T \max_{t \in \text{Nodes}(T)} \rho(Q_t)$

$\rho$  = edge covering number

$ghtw(Q) = 2$

*GHTW* is the complexity of a still naïve algorithm



Nested loop join for  $Q_t$ :  
 Runtime for  $Q$ :  $O(N^{ghtw(Q)} + |\text{Output}|)$



# Fractional Hypertree Width

**Def**  $\text{fhtw}(Q) = \min_T \max_{t \in \text{Nodes}(T)} \rho^*(Q_t)$

$\rho^*$  = Fractional edge covering number

Next lecture

$\text{fhtw}(Q) = 3/2$

$\rho^* = 3/2$

$1/2$   
R(x,y),  
A(y,s),B(x,s)  
 $1/2$   $1/2$

$1$   
S(y,z,u)

$1$   $0$   
K(z,v),F(z,v)

$\rho^* = 1$

$1$   
L(v,m)

$\rho^* = 1$

$\rho^* = 1$

$\rho^* = 4/3$

$1/3$   $1/3$   
T(y,z,w),C(z,w,t)  
D(w,t,y),E(t,y,z)  
 $1/3$   $1/3$

Generic join for  $Q_t$   
Runtime for  $Q$ :  $O(N^{\text{fhtw}(Q)} + |\text{Output}|)$

# Outline

- Acyclic queries, Yannakakis algorithm
- Tree decomposition of cyclic queries
- Worst-case optimal algorithm; next week



Wednesday!