

# CSE599d: Advanced Query Processing

## Lecture 14: The AGM Bound

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# Announcements

- HW3 due: Friday Feb. 27

## Where We Are

- Tree decomposition of a graph, tree-width  $\text{tw}(G)$ .
- Hypertree decomposition of a hypergraph/query, hypertree width  $\text{HTW}(Q)$ .
- Query evaluation in time  $O(N^{\text{HTW}} + \text{OUT})$
- Can we evaluate the bags more efficiently than  $N^{\text{HTW}}$ ?  
How large are the bags in the first place? Cardinality estimation?

# Outline

Cardinality estimation: critical for query optimization.

Alternative: upper bound on the cardinality.

Today: the AGM Bound.

Monday: Worst Case Optimal Join (WCOJ) a.k.a. Generic Join (GJ).

Later: significant extensions of the AGM bound.

# Motivation for Upper Bound

## Recap: Cardinality Estimation

Given:

- Statistics on the input relations  $R_1, R_2, \dots$
- A selection-join query (same as full conjunctive query)  $Q$

“Estimate”:

- The size  $|Q(\mathbf{D})|$ .

## Recap: Cardinality Estimation

Assumptions: uniformity, independence, preservation of values

- Selection  $\sigma_p(R)$ :

$$\text{Est}(\sigma_p(R)) = \theta_p \cdot |R|$$

$$\theta_{p_1 \wedge p_2} = \theta_{p_1} \cdot \theta_{p_2}$$

- Join:

$$\text{Est}(R(A, B) \bowtie S(B, C)) = \frac{|R| \cdot |S|}{\max(|\text{Dom}(R.B)|, |\text{Dom}(S.B)|)}$$

- 1-d histograms address uniformity.  
2-d histograms address independence, but rarely used **why???**

# The Output Bound Problem

Instead of an estimate, compute a guaranteed output bound  $B$ :

$$|Q(\mathbf{D})| \leq B$$

Challenge: make  $B$  tight (as close as possible to  $|Q(\mathbf{D})|$ ).

# Simple Examples

Assume  $|R| \leq N$ ,  $|S| \leq N$ ,  $|T| \leq N$ .

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$$\max_{\mathcal{D}} |Q(\mathcal{D})| = ?$$

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Here we use a **fractional edge cover**

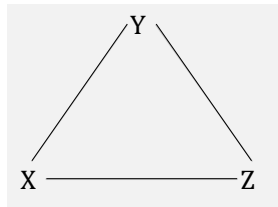
$$\max_{\mathcal{D}} |Q(\mathcal{D})| = N^{\frac{3}{2}}$$

# AGM Bound: The Statement

# Fractional Edge Covers

Query  $Q$  to hypgraph  $G = (V, E)$ .

$$R(X, Y) \wedge S(Y, Z) \wedge T(Z, X)$$



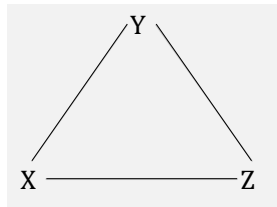
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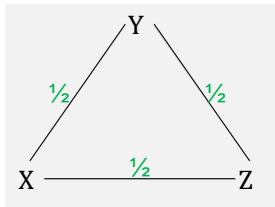


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## The AGM Bound [Atserias et al., 2013]

$$Q(\mathbf{X}) = R_1(\mathbf{Y}_1) \wedge \cdots \wedge R_m(\mathbf{Y}_m)$$

The **fractional edge covering number** is:

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**Theorem** [Upper Bound] If  $|R_1|, \dots, |R_m| \leq N$ . Then:

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**Theorem** [Lower Bound] There exists  $|R_1|, \dots, |R_m| \leq N$  and  $Q(\mathbf{D}) \geq \frac{1}{2^{|\mathbf{X}|}} N^{\rho^*}$

The AGM bound  $N^{\rho^*}$  is tight

## Examples

What are  $w$  and  $\rho^*$ ? The AGM bound? Assume  $|R| = |S| = \dots = N$ .

5-cycle:

$$R(X, Y) \wedge S(Y, Z) \wedge T(Z, U) \wedge K(U, V) \wedge L(V, X)$$

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$$\mathbf{w} = \left(\frac{1}{2}, \frac{1}{2}, \frac{1}{2}, \frac{1}{2}, \frac{1}{2}\right), \quad \text{AGM} = N^{5/2}.$$

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$$\mathbf{w} = (1, 0, 1, 1), \text{ or } (1, 1, 0, 1), \quad \text{AGM} = N^3.$$

# The General AGM Bound

$$Q(\mathbf{X}) = R_1(\mathbf{Y}_1) \wedge \cdots \wedge R_m(\mathbf{Y}_m)$$

**Theorem** [Upper Bound] For every fractional edge cover  $\mathbf{w}$ :

$$|Q| \leq |R_1|^{w_1} \cdots |R_m|^{w_m}$$

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$$R(X, Y) \wedge S(Y, Z) \wedge T(Z, X) \qquad AGM(Q) = \min \left( \begin{array}{c} (|R| \cdot |S| \cdot |T|)^{1/2} \\ |R| \cdot |S| \\ |R| \cdot |T| \\ |S| \cdot |T| \end{array} \right)$$

# Proof Outline

- Proof of upper bound: use an interesting inequality.
  
  
  
  
  
  
  
  
  
  
- Proof of lower bound: worst-case database using strong duality.

# An Inequality

# Review: Cauchy-Schwartz, Hölder

Assume non-negative vectors, or matrices  $\mathbf{a}, \mathbf{b}, \mathbf{c}, \dots$

Cauchy-Schwartz: 
$$\sum_x a_x b_x \leq (\sum_x a_x^2)^{1/2} (\sum_x b_x^2)^{1/2}$$

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If  $u + v + w + \dots \geq 1$ : 
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$G = (\{X, Y, Z\}, \{(XY), (YZ), (ZX)\}), \mathbf{a}, \mathbf{b}, \mathbf{c} \in \mathbb{R}_+^{N^2}, \mathbf{w} = \left(\frac{1}{2}, \frac{1}{2}, \frac{1}{2}\right)$ .

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$$\sum_{xyz} a_{xy}^{\frac{1}{2}} b_{yz}^{\frac{1}{2}} c_{zx}^{\frac{1}{2}} \leq \left( \sum_{xy} a_{xy} \right)^{\frac{1}{2}} \cdot \left( \sum_{yz} b_{yz} \right)^{\frac{1}{2}} \cdot \left( \sum_{zx} c_{zx} \right)^{\frac{1}{2}}$$

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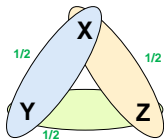
$G = (\{X, Y, Z\}, \{(XY), (YZ), (ZX)\})$ ,  $\mathbf{a}, \mathbf{b}, \mathbf{c} \in \mathbb{R}_+^{N^2}$ ,  $\mathbf{w} = \left( \frac{1}{2}, \frac{1}{2}, \frac{1}{2} \right)$ .

$$\sum_{xyz} a_{xy}^{\frac{1}{2}} b_{yz}^{\frac{1}{2}} c_{zx}^{\frac{1}{2}} \leq \left( \sum_{xy} a_{xy} \right)^{\frac{1}{2}} \cdot \left( \sum_{yz} b_{yz} \right)^{\frac{1}{2}} \cdot \left( \sum_{zx} c_{zx} \right)^{\frac{1}{2}}$$

$$\sum_{xyz} a_{xy} b_{yz} c_{zx} \leq \left( \sum_{xy} a_{xy}^2 \right)^{\frac{1}{2}} \cdot \left( \sum_{yz} b_{yz}^2 \right)^{\frac{1}{2}} \cdot \left( \sum_{zx} c_{zx}^2 \right)^{\frac{1}{2}}$$

# Example

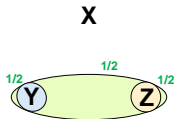
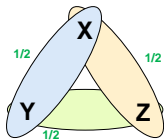
$$\sum_{xyz} a_{xy}^{\frac{1}{2}} b_{yz}^{\frac{1}{2}} c_{zx}^{\frac{1}{2}} \leq (\sum_{xy} a_{xy})^{\frac{1}{2}} \cdot (\sum_{yz} b_{yz})^{\frac{1}{2}} \cdot (\sum_{zx} c_{zx})^{\frac{1}{2}}$$



$$\sum_{xyz} a_{xy}^{\frac{1}{2}} b_{yz}^{\frac{1}{2}} c_{zx}^{\frac{1}{2}} = \sum_x \left( \sum_{yz} a_{xy}^{\frac{1}{2}} b_{yz}^{\frac{1}{2}} c_{zx}^{\frac{1}{2}} \right)$$

# Example

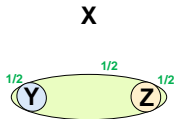
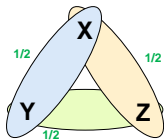
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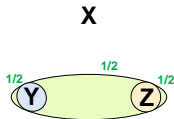
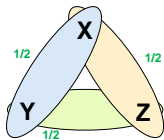
$$\sum_{xyz} a_{xy}^{\frac{1}{2}} b_{yz}^{\frac{1}{2}} c_{zx}^{\frac{1}{2}} \leq (\sum_{xy} a_{xy})^{\frac{1}{2}} \cdot (\sum_{yz} b_{yz})^{\frac{1}{2}} \cdot (\sum_{zx} c_{zx})^{\frac{1}{2}}$$



$$\begin{aligned} \sum_{xyz} a_{xy}^{\frac{1}{2}} b_{yz}^{\frac{1}{2}} c_{zx}^{\frac{1}{2}} &= \sum_x \left( \sum_{yz} a_{xy}^{\frac{1}{2}} b_{yz}^{\frac{1}{2}} c_{zx}^{\frac{1}{2}} \right) \\ &\leq \sum_x \left( \left( \sum_y a_{xy} \right)^{\frac{1}{2}} \cdot \left( \sum_{yz} b_{yz} \right)^{\frac{1}{2}} \cdot \left( \sum_z c_{zx} \right)^{\frac{1}{2}} \right) \end{aligned}$$

# Example

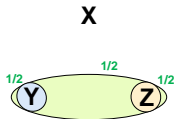
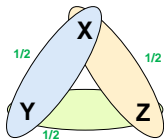
$$\sum_{xyz} a_{xy}^{\frac{1}{2}} b_{yz}^{\frac{1}{2}} c_{zx}^{\frac{1}{2}} \leq (\sum_{xy} a_{xy})^{\frac{1}{2}} \cdot (\sum_{yz} b_{yz})^{\frac{1}{2}} \cdot (\sum_{zx} c_{zx})^{\frac{1}{2}}$$



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# Example

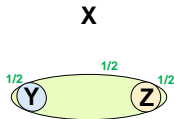
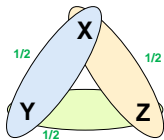
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# Example

$$\sum_{xyz} a_{xy}^{\frac{1}{2}} b_{yz}^{\frac{1}{2}} c_{zx}^{\frac{1}{2}} \leq (\sum_{xy} a_{xy})^{\frac{1}{2}} \cdot (\sum_{yz} b_{yz})^{\frac{1}{2}} \cdot (\sum_{zx} c_{zx})^{\frac{1}{2}}$$



$$\begin{aligned} \sum_{xyz} a_{xy}^{\frac{1}{2}} b_{yz}^{\frac{1}{2}} c_{zx}^{\frac{1}{2}} &= \sum_x \left( \sum_{yz} a_{xy}^{\frac{1}{2}} b_{yz}^{\frac{1}{2}} c_{zx}^{\frac{1}{2}} \right) \\ &\leq \sum_x \left( \left( \sum_y a_{xy} \right)^{\frac{1}{2}} \cdot \left( \sum_{yz} b_{yz} \right)^{\frac{1}{2}} \cdot \left( \sum_z c_{zx} \right)^{\frac{1}{2}} \right) \stackrel{\text{def}}{=} \sum_x \left( A_x^{\frac{1}{2}} \cdot B^{\frac{1}{2}} \cdot C_x^{\frac{1}{2}} \right) \\ &\leq \left( \sum_x A_x \right)^{\frac{1}{2}} \cdot B^{\frac{1}{2}} \cdot \left( \sum_x C_x \right)^{\frac{1}{2}} = \left( \sum_{xy} a_{xy} \right)^{\frac{1}{2}} \cdot \left( \sum_{yz} b_{yz} \right)^{\frac{1}{2}} \cdot \left( \sum_{zx} c_{zx} \right)^{\frac{1}{2}} \end{aligned}$$

# Proof of Friedgut's Inequality

$$\sum_{x_1, \dots, x_n} \prod_{j=1, m} \left( a_{j, \pi_{e_j}(\mathbf{x})} \right)^{w_j} \leq \prod_{j=1, m} \left( \sum_{y_j} a_{j, y_j} \right)^{w_j}$$

Induction:

## Proof of Friedgut's Inequality

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**Induction:** remove  $X_1$  from  $G$  to obtain  $G - \{X_1\}$ , keep  $w_1, \dots, w_m$ :

## Proof of Friedgut's Inequality

$$\sum_{x_1, \dots, x_n} \prod_{j=1, m} (a_{j, \pi_{e_j}(\mathbf{x})})^{w_j} \leq \prod_{j=1, m} (\sum_{y_j} a_{j, y_j})^{w_j}$$

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- Edges in  $G - \{X_1\}$  are  $e_1 - \{X_1\}, \dots, e_m - \{X_1\}$ ; still covered.

## Proof of Friedgut's Inequality

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## Proof of Friedgut's Inequality

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## Proof of Friedgut's Inequality

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# How To Impress Your Friends Doing ML

Upper bound on the trace of  $ABC$ .

Query/hypergraph:  $A(X, Y) \wedge B(Y, Z) \wedge C(Z, X)$

$$\mathbf{w} = \left(\frac{1}{2}, \frac{1}{2}, \frac{1}{2}\right)$$

$$\text{Tr}(\mathbf{a} \cdot \mathbf{b} \cdot \mathbf{c}) \stackrel{\text{def}}{=} \sum_{xyz} a_{xy} b_{yz} c_{zx} \leq \left(\sum_{xy} a_{xy}^2\right)^{\frac{1}{2}} \left(\sum_{yz} b_{yz}^2\right)^{\frac{1}{2}} \left(\sum_{zx} c_{zx}^2\right)^{\frac{1}{2}} = \|\mathbf{a}\|_2 \cdot \|\mathbf{b}\|_2 \cdot \|\mathbf{c}\|_2$$

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The tensor kernel for Loomis-Whitney:

$A(X, Y, Z) \wedge B(Y, Z, U) \wedge C(Z, U, X) \wedge D(U, X, Y)$   $\mathbf{w} = (\frac{1}{3}, \frac{1}{3}, \frac{1}{3}, \frac{1}{3})$

$$\sum_{xyz} a_{xyz} b_{yzu} c_{zux} d_{uxy} \leq \left( \sum_{xyz} (a_{xyz})^3 \right)^{\frac{1}{3}} \cdots \left( \sum_{uxy} (d_{uxy})^3 \right)^{\frac{1}{3}} = \|\mathbf{a}\|_3 \cdot \|\mathbf{b}\|_3 \cdot \|\mathbf{c}\|_3 \cdot \|\mathbf{d}\|_3$$

# How To Impress Your Friends Doing ML

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And another one

$A(X, U) \wedge B(Y, U) \wedge C(Z, U) \wedge D(X, Y, Z)$   $\mathbf{w} = (\frac{1}{3}, \frac{1}{3}, \frac{1}{3}, \frac{2}{3})$

$$\sum_{xyz} a_{xu} b_{yu} c_{zu} d_{xyz} \leq \left( \sum_{xu} a_{xu}^3 \right)^{\frac{1}{3}} \cdot \left( \sum_{yu} b_{yu}^3 \right)^{\frac{1}{3}} \cdot \left( \sum_{zu} c_{zu}^3 \right)^{\frac{1}{3}} \cdot \left( \sum_{xyz} d_{xyz}^3 \right)^{\frac{2}{3}}$$

## Proof of the AGM Upper Bound

$$Q(X_1, \dots, X_n) = R_1(\mathbf{Y}_1) \wedge \dots \wedge R_m(\mathbf{Y}_m)$$

**Theorem** For every fractional edge cover  $\mathbf{w}$ :  $|Q| \leq |R_1^D|^{w_1} \dots |R_m^D|^{w_m}$

### Proof

$N \stackrel{\text{def}}{=} \text{number of constants in the database } \mathbf{D}.$

$$\forall \mathbf{y}_j \in [N]^{e_j}, \quad a_{j, \mathbf{y}_j} \stackrel{\text{def}}{=} \begin{cases} 1 & \text{if } \mathbf{y}_j \in R_j^D \\ 0 & \text{otherwise} \end{cases}$$

$$|Q(\mathbf{D})| = \sum_{x_1, \dots, x_n} a_{1, \pi_{e_1}(\mathbf{x})} \dots a_{m, \pi_{e_m}(\mathbf{x})} \leq \left( \sum_{\mathbf{y}_1} a_{1, \mathbf{y}_1}^{w_1} \right)^{w_1} \dots \left( \sum_{\mathbf{y}_m} a_{m, \mathbf{y}_m}^{w_m} \right)^{w_m} = |R_1^D|^{w_1} \dots |R_m^D|^{w_m}$$

# Discussion

- Each fractional edge cover gives us some upper bound on  $|Q(\mathbf{D})|$ .
- Their minimum is also an upper bound.
- Next: will prove that this minimum is tight.

# Proof of the Lower Bound

## Proof of the AGM Lower Bound

By example:

$$R(X, Y) \wedge S(Y, Z) \wedge T(Z, X)$$

$$AGM(Q) = \min_{\mathbf{w}} |R|^{w_R} \cdot |S|^{w_S} \cdot |T|^{w_T}$$

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**Primal program:**

Minimize

$$w_R \log |R| + w_S \log |S| + w_T \log |T|$$

where  $\mathbf{w}$  is frac. edge cover:

$$X : \quad w_R + \quad \quad w_T \geq 1$$

$$Y : \quad w_R + w_S \quad \quad \geq 1$$

$$Z : \quad \quad w_S + w_T \geq 1$$

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$$\begin{array}{rcll} X : & w_R + & w_T & \geq 1 \\ Y : & w_R + & w_S & \geq 1 \\ Z : & w_S + & w_T & \geq 1 \end{array}$$

**Dual program:**

Maximize

$$v_X + v_Y + v_Z$$

where  $\mathbf{v}$  is “frac. vertex packing”:

$$\begin{array}{rcll} R : & v_X + & v_Y & \leq \log |R| \\ S : & & v_Y + & v_Z \leq \log |S| \\ T : & v_X + & v_Z & \leq \log |T| \end{array}$$

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Take optimum  $\mathbf{v}$ , define:  $\text{Dom}(X) \stackrel{\text{def}}{=} \lfloor \lfloor 2^{v_X} \rfloor \rfloor$ ,  $\text{Dom}(Y) \stackrel{\text{def}}{=} \lfloor \lfloor 2^{v_Y} \rfloor \rfloor$ ,  $\text{Dom}(Z) \stackrel{\text{def}}{=} \lfloor \lfloor 2^{v_Z} \rfloor \rfloor$ .

# Proof of the AGM Lower Bound

By example:

$$R(X, Y) \wedge S(Y, Z) \wedge T(Z, X)$$

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Take optimum  $\mathbf{v}$ , define:  $\text{Dom}(X) \stackrel{\text{def}}{=} \llbracket \lfloor 2^{v_X} \rrbracket \rrbracket$ ,  $\text{Dom}(Y) \stackrel{\text{def}}{=} \llbracket \lfloor 2^{v_Y} \rrbracket \rrbracket$ ,  $\text{Dom}(Z) \stackrel{\text{def}}{=} \llbracket \lfloor 2^{v_Z} \rrbracket \rrbracket$ .

Worst-case instance (cartesian products):  $R^* \stackrel{\text{def}}{=} \text{Dom}(X) \times \text{Dom}(Y)$ ,  $S^*$ ,  $T^* \stackrel{\text{def}}{=} \dots$

# Proof of the AGM Lower Bound

By example:

$$R(X, Y) \wedge S(Y, Z) \wedge T(Z, X)$$

$$AGM(Q) = \min_{\mathbf{w}} |R|^{w_R} \cdot |S|^{w_S} \cdot |T|^{w_T}$$

### Primal program:

Minimize

$$w_R \log |R| + w_S \log |S| + w_T \log |T|$$

where  $\mathbf{w}$  is frac. edge cover:

$$\begin{array}{rcll} X : & w_R + & w_T & \geq 1 \\ Y : & w_R + & w_S & \geq 1 \\ Z : & w_S + & w_T & \geq 1 \end{array}$$

### Dual program:

Maximize

$$v_X + v_Y + v_Z$$

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$$|Q^*| = \lfloor 2^{v_X} \rfloor \cdot \lfloor 2^{v_Y} \rfloor \cdot \lfloor 2^{v_Z} \rfloor \geq \frac{1}{8} 2^{v_X + v_Y + v_Z}$$

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## Special Case: $|R| = |S| = \dots = N$

Fix a hypergraph  $(V, E)$ ;  $(v_X)_{X \in V} \in \mathbb{R}_+^{|V|}$  is a **fractional vertex packing** if:

$$\forall Y \in E : \boxed{\sum_{X \in Y} v_X \leq 1}$$

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Then:  $R = [N^{v_X}] \times [N^{v_Y}]$ ,  $S = [N^{v_Y}] \times [N^{v_Z}]$ ,  $T = [N^{v_X}] \times [N^{v_Z}]$ .

$$Q = [N^{v_X}] \times [N^{v_Y}] \times [N^{v_Z}]$$

# Examples

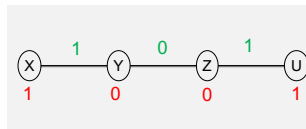
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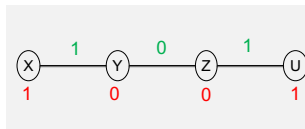


# Examples

$$|R| = |S| = \dots = N$$

$$R(X, Y) \wedge S(Y, Z) \wedge T(Z, U)$$

$$R = [N] \times [1], S = [1] \times [1], T = [1] \times [N].$$

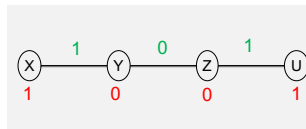


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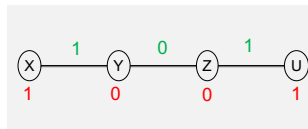
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# Examples

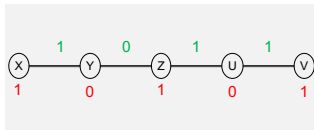
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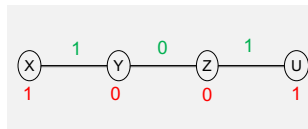


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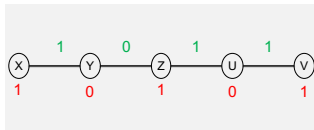
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$$R(X, Y) \wedge S(Y, Z) \wedge T(Z, U) \wedge K(U, V)$$

$$R = T = [N] \times [1], S = K = [1] \times [N]$$



## Summary of the AGM Bound

- Upper / lower bound: fractional [edge cover](#) / [vertex packing](#).
- Their equality follows from strong duality.
- The worst-case instance of the AGM bound is a [Product Database](#).
- Full CQs only. Otherwise, ignore non-head variables.

Limitation of AGM: only [cardinalities](#). Next week: extensions to [other stats](#).



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