

# CSE599d: Advanced Query Processing

## Lecture 15: The AGM Bound

Dan Suciu

University of Washington



# Today's Agenda

- AGM: the upper bound (review)
- AGM: the lower bound
- Extensions: adding keys, adding projections

# AGM Upper Bound

# Fractional Edge Covers

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

A **fractional edge cover** is  $\mathbf{w} = (w_e)_{e \in E}$ ,  $w_e \geq 0$ :

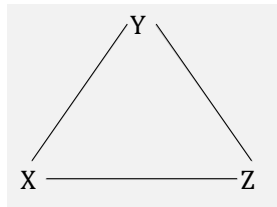
$$\forall x \in V, \sum_{e \in E: x \in e} w_e \geq 1.$$

# Fractional Edge Covers

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

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

A **fractional edge cover** is  $\mathbf{w} = (w_e)_{e \in E}$ ,  $w_e \geq 0$ :  
 $\forall X \in V, \sum_{e \in E: X \in e} w_e \geq 1$ .



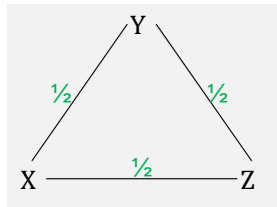
# Fractional Edge Covers

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

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

A **fractional edge cover** is  $\mathbf{w} = (w_e)_{e \in E}$ ,  $w_e \geq 0$ :

$$\forall X \in V, \sum_{e \in E: X \in e} w_e \geq 1.$$



# The Fractional Edge Covering Number

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

The fractional edge covering number is:

$$\rho^* = \min_{\mathbf{w}} (w_1 + \cdots + w_m)$$

# The AGM Bound: Uniform Cardinalities

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

$$Q(\mathbf{D}) \leq N^{\rho^*}$$

[Atserias et al., 2013]

## Simple Examples

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

5-cycle:

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

## Simple Examples

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

5-cycle:

$$R(X, Y) \wedge S(Y, Z) \wedge T(Z, U) \wedge K(U, V) \wedge L(V, X)$$
$$\mathbf{w} = (1, 0, 1, 0, 1), \quad |Q| \leq N^3.$$

# Simple Examples

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

5-cycle:

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

$$\mathbf{w} = (1, 0, 1, 0, 1),$$

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

$$|Q| \leq N^3.$$

$$AGM = N^{5/2}.$$

## Simple Examples

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

5-cycle:

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

$$\mathbf{w} = (1, 0, 1, 0, 1),$$

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

$$|Q| \leq N^3.$$

$$AGM = N^{5/2}.$$

Loomis-Whitney:

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

$$\mathbf{w} = \left(\frac{1}{3}, \frac{1}{3}, \frac{1}{3}, \frac{1}{3}\right),$$

$$AGM = N^{4/3}$$

## Simple Examples

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

5-cycle:

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

$$\mathbf{w} = (1, 0, 1, 0, 1), \quad |Q| \leq N^3.$$

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

Loomis-Whitney:

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

$$\mathbf{w} = \left(\frac{1}{3}, \frac{1}{3}, \frac{1}{3}, \frac{1}{3}\right), \quad \text{AGM} = N^{4/3}$$

Chain of length 4:

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

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

## The AGM Bound: Non-uniform Cardinalities

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

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

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

[Atserias et al., 2013]

## The AGM Bound: Non-uniform Cardinalities

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

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

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

The AGM bound is the minimum over all fractional edge covers  $\mathbf{w}$ :

$$AGM(Q) := \min_{\mathbf{w}} |R_1|^{w_1} \cdots |R_m|^{w_m}$$

[Atserias et al., 2013]

## The AGM Bound: Non-uniform Cardinalities

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

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

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

The AGM bound is the minimum over all fractional edge covers  $\mathbf{w}$ :

$$AGM(Q) := \min_{\mathbf{w}} |R_1|^{w_1} \cdots |R_m|^{w_m}$$

When  $|R_1| = |R_2| = \cdots = N$  then  $AGM(Q) = N^{\min_{\mathbf{w}} \sum w_i} = N^{\rho^*}$

[Atserias et al., 2013]

# Example

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

## Example

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

Assuming  $T$  is the largest among  $R, S, T$ :

$$AGM(Q) = \begin{cases} (|R| \cdot |S| \cdot |T|)^{1/2} & \text{when } |T| \leq |R| \cdot |S| \\ |R| \cdot |S| & \text{otherwise} \end{cases}$$

# Proof of the Upper Bound

There are two ways to prove the upper bound:

- Using a class of inequalities introduced by Friedgut. Reviewed next.
  
- Using entropic inequalities: Next week.

## Review: Generalized Hölder's Inequality

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

If  $u + v + w + \dots \geq 1$  then:

$$\sum_x a_x^u b_x^v c_x^w \dots \leq (\sum_x a_x)^u (\sum_x b_x)^v (\sum_x c_x)^w \dots$$

## Review: Generalized Hölder's Inequality

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

If  $u + v + w + \dots \geq 1$  then:

$$\sum_x a_x^u b_x^v c_x^w \dots \leq (\sum_x a_x)^u (\sum_x b_x)^v (\sum_x c_x)^w \dots$$

Condition  $u + v + w + \dots \geq 1$  says that  $(u, v, \dots)$  is a fractional edge cover of the hypergraph  $(V, E)$ , where:

$$V := \{X\}, E := \{\{X\}, \{X\}, \dots\}$$

## Friedgut's Inequality [Friedgut, 2004]

Hypergraph  $G = (V = \{X_1, \dots, X_n\}, E = \{e_1, \dots, e_m\})$ .

$N \in \mathbb{N}$ , multi-dimensional vectors:  $\mathbf{a}_1 \in \mathbb{R}_+^{N^{e_1}}, \dots, \mathbf{a}_m \in \mathbb{R}_+^{N^{e_m}}$ .

For any fractional edge cover  $w_1, \dots, w_m$  of  $G$ , the following holds:

$$\sum_{\mathbf{x} \in [N]^n} \left( \prod_{j=1, m} \mathbf{a}_{j, \pi_{e_j}(\mathbf{x})}^{w_j} \right) \leq \prod_{j=1, m} \left( \sum_{\mathbf{y}_j \in [N]^{e_j}} \mathbf{a}_{j, \mathbf{y}_j} \right)^{w_j}$$

## Friedgut's Inequality [Friedgut, 2004]

Hypergraph  $G = (V = \{X_1, \dots, X_n\}, E = \{e_1, \dots, e_m\})$ .

$N \in \mathbb{N}$ , multi-dimensional vectors:  $\mathbf{a}_1 \in \mathbb{R}_+^{N^{e_1}}, \dots, \mathbf{a}_m \in \mathbb{R}_+^{N^{e_m}}$ .

For any fractional edge cover  $w_1, \dots, w_m$  of  $G$ , the following holds:

$$\sum_{\mathbf{x} \in [N]^n} \left( \prod_{j=1, m} \mathbf{a}_{j, \pi_{e_j}(\mathbf{x})}^{w_j} \right) \leq \prod_{j=1, m} \left( \sum_{\mathbf{y}_j \in [N]^{e_j}} \mathbf{a}_{j, \mathbf{y}_j} \right)^{w_j}$$

E.g. 
$$\sum_{x,y,z \in [N]} a_{xy}^{\frac{1}{2}} b_{yz}^{\frac{1}{2}} c_{zx}^{\frac{1}{2}} \leq \left( \sum_{x,y \in [N]} a_{xy} \right)^{\frac{1}{2}} \cdot \left( \sum_{y,z \in [N]} b_{yz} \right)^{\frac{1}{2}} \cdot \left( \sum_{z,x \in [N]} c_{zx} \right)^{\frac{1}{2}}$$

## Proof of Friedgut's Inequality

$$\sum_{\mathbf{x} \in [N]^n} \left( \prod_{j=1, m} a_{j, \pi_{e_j}(\mathbf{x})}^{w_j} \right) \leq \prod_{j=1, m} \left( \sum_{\mathbf{y}_j \in [N]^{e_j}} a_{j, \mathbf{y}_j} \right)^{w_j}$$

Induction:

## Proof of Friedgut's Inequality

$$\sum_{\mathbf{x} \in [N]^n} \left( \prod_{j=1, m} a_{j, \pi_{e_j}(\mathbf{x})}^{w_j} \right) \leq \prod_{j=1, m} \left( \sum_{\mathbf{y}_j \in [N]^{e_j}} a_{j, \mathbf{y}_j} \right)^{w_j}$$

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

## Proof of Friedgut's Inequality

$$\sum_{\mathbf{x} \in [N]^n} \left( \prod_{j=1, m} a_{j, \pi_{e_j}(\mathbf{x})}^{w_j} \right) \leq \prod_{j=1, m} \left( \sum_{\mathbf{y}_j \in [N]^{e_j}} a_{j, \mathbf{y}_j} \right)^{w_j}$$

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

- $X_1$  is covered in  $G$

## Proof of Friedgut's Inequality

$$\sum_{\mathbf{x} \in [N]^n} \left( \prod_{j=1, m} a_{j, \pi_{e_j}(\mathbf{x})}^{w_j} \right) \leq \prod_{j=1, m} \left( \sum_{\mathbf{y}_j \in [N]^{e_j}} a_{j, \mathbf{y}_j} \right)^{w_j}$$

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

- $X_1$  is covered in  $G$
- Edges in  $G - \{X_1\}$  are  $e_1 - \{X_1\}, \dots, e_m - \{X_1\}$ ;  $\mathbf{w}$  still a cover.

## Proof of Friedgut's Inequality

$$\sum_{\mathbf{x} \in [N]^n} \left( \prod_{j=1, m} a_{j, \pi_{e_j}(\mathbf{x})}^{w_j} \right) \leq \prod_{j=1, m} \left( \sum_{\mathbf{y}_j \in [N]^{e_j}} a_{j, \mathbf{y}_j} \right)^{w_j}$$

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

- $X_1$  is covered in  $G$
- Edges in  $G - \{X_1\}$  are  $e_1 - \{X_1\}, \dots, e_m - \{X_1\}$ ;  $\mathbf{w}$  still a cover.

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

## Proof of Friedgut's Inequality

$$\sum_{\mathbf{x} \in [N]^n} \left( \prod_{j=1, m} a_{j, \pi_{e_j}(\mathbf{x})}^{w_j} \right) \leq \prod_{j=1, m} \left( \sum_{\mathbf{y}_j \in [N]^{e_j}} a_{j, \mathbf{y}_j} \right)^{w_j}$$

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

- $X_1$  is covered in  $G$
- Edges in  $G - \{X_1\}$  are  $e_1 - \{X_1\}, \dots, e_m - \{X_1\}$ ;  $\mathbf{w}$  still a cover.

$$\sum_{x_1, \dots, x_n} \left( \prod_{j=1, m} a_{j, \pi_{e_j}(\mathbf{x})}^{w_j} \right) = \sum_{x_1} \left( \sum_{x_2, \dots, x_n} \left( \prod_{j=1, m} a_{j, \pi_{e_j}(\mathbf{x})}^{w_j} \right) \right)$$

## Proof of Friedgut's Inequality

$$\sum_{\mathbf{x} \in [N]^n} \left( \prod_{j=1, m} a_{j, \pi_{e_j}(\mathbf{x})}^{w_j} \right) \leq \prod_{j=1, m} \left( \sum_{\mathbf{y}_j \in [N]^{e_j}} a_{j, \mathbf{y}_j} \right)^{w_j}$$

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

- $X_1$  is covered in  $G$
- Edges in  $G - \{X_1\}$  are  $e_1 - \{X_1\}, \dots, e_m - \{X_1\}$ ;  $\mathbf{w}$  still a cover.

$$\begin{aligned} \sum_{x_1, \dots, x_n} \left( \prod_{j=1, m} a_{j, \pi_{e_j}(\mathbf{x})}^{w_j} \right) &= \sum_{x_1} \left( \sum_{x_2, \dots, x_n} \left( \prod_{j=1, m} a_{j, \pi_{e_j}(\mathbf{x})}^{w_j} \right) \right) \\ &\stackrel{\text{Induction}}{\leq} \sum_{x_1} \left( \prod_{j=1, m} \left( \sum_{\mathbf{y}_j - \{x_1\}} a_{j, \mathbf{y}_j} \right)^{w_j} \right) \end{aligned}$$

## Proof of Friedgut's Inequality

$$\sum_{\mathbf{x} \in [N]^n} \left( \prod_{j=1, m} a_{j, \pi_{e_j}(\mathbf{x})}^{w_j} \right) \leq \prod_{j=1, m} \left( \sum_{\mathbf{y}_j \in [N]^{e_j}} a_{j, \mathbf{y}_j} \right)^{w_j}$$

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

- $X_1$  is covered in  $G$
- Edges in  $G - \{X_1\}$  are  $e_1 - \{X_1\}, \dots, e_m - \{X_1\}$ ;  $\mathbf{w}$  still a cover.

$$\begin{aligned} \sum_{x_1, \dots, x_n} \left( \prod_{j=1, m} a_{j, \pi_{e_j}(\mathbf{x})}^{w_j} \right) &= \sum_{x_1} \left( \sum_{x_2, \dots, x_n} \left( \prod_{j=1, m} a_{j, \pi_{e_j}(\mathbf{x})}^{w_j} \right) \right) \\ &\stackrel{\text{Induction}}{\leq} \sum_{x_1} \left( \prod_{j=1, m} \left( \sum_{\mathbf{y}_j - \{x_1\}} a_{j, \mathbf{y}_j} \right)^{w_j} \right) \\ &\stackrel{\text{Hölder}}{\leq} \prod_{j=1, m} \left( \sum_{\mathbf{y}_j} a_{j, \mathbf{y}_j} \right)^{w_j} \end{aligned}$$

# Moving Exponents from Left to Right

Cauchy-Schwartz becomes a special case:

$$\sum_x a_x^{\frac{1}{2}} b_x^{\frac{1}{2}} \leq \left( \sum_x a_x \right)^{\frac{1}{2}} \left( \sum_x b_x \right)^{\frac{1}{2}}$$

$$\sum_x a_x b_x \leq \left( \sum_x a_x^2 \right)^{\frac{1}{2}} \left( \sum_x b_x^2 \right)^{\frac{1}{2}}$$

## Moving Exponents from Left to Right

Cauchy-Schwartz becomes a special case:

$$\sum_x a_x^{\frac{1}{2}} b_x^{\frac{1}{2}} \leq \left( \sum_x a_x \right)^{\frac{1}{2}} \left( \sum_x b_x \right)^{\frac{1}{2}}$$
$$\sum_x a_x b_x \leq \left( \sum_x a_x^2 \right)^{\frac{1}{2}} \left( \sum_x b_x^2 \right)^{\frac{1}{2}}$$

For a triangle, we obtain:

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

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

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

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

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

# 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

## Statement of the Lower Bound: Uniform Cardinalities

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

### **Theorem** [Lower Bound]

For every  $N$ , there exists an instance  $\mathbf{D} = (R_1^D, \dots, R_m^D)$  s.t.

$$|R_1^D|, \dots, |R_m^D| \leq N \quad \text{and} \quad Q(\mathbf{D}) \geq \frac{1}{2^{|\mathbf{X}|}} N^{\rho^*}$$

## Statement of the Lower Bound: Uniform Cardinalities

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

**Theorem** [Lower Bound]

For every  $N$ , there exists an instance  $\mathbf{D} = (R_1^D, \dots, R_m^D)$  s.t.

$$|R_1^D|, \dots, |R_m^D| \leq N \quad \text{and} \quad Q(\mathbf{D}) \geq \frac{1}{2^{|\mathbf{X}|}} N^{\rho^*}$$

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

The case of non-uniform cardinalities is similar; will discuss shortly.

# Background: The Duality Theorem for Linear Programming

**Primal Linear Program:**Minimize  $\mathbf{c}^T \mathbf{x}$ where  $\mathbf{x} \geq 0$  satisfies:

$$\mathbf{Ax} \geq \mathbf{b}$$

# Background: The Duality Theorem for Linear Programming

**Primal Linear Program:**Minimize  $\mathbf{c}^T \mathbf{x}$ where  $\mathbf{x} \geq \mathbf{0}$  satisfies:

$$\mathbf{Ax} \geq \mathbf{b}$$

**Dual Linear Program:**Maximize  $\mathbf{y}^T \mathbf{b}$ where  $\mathbf{y} \geq \mathbf{0}$  satisfies:

$$\mathbf{y}^T \mathbf{A} \leq \mathbf{c}^T$$

# Background: The Duality Theorem for Linear Programming

**Primal Linear Program:**Minimize  $\mathbf{c}^T \mathbf{x}$ where  $\mathbf{x} \geq \mathbf{0}$  satisfies:

$$\mathbf{Ax} \geq \mathbf{b}$$

**Dual Linear Program:**Maximize  $\mathbf{y}^T \mathbf{b}$ where  $\mathbf{y} \geq \mathbf{0}$  satisfies:

$$\mathbf{y}^T \mathbf{A} \leq \mathbf{c}^T$$

**Weak Duality Theorem** If  $\mathbf{x}, \mathbf{y}$  are feasible solutions then  $\mathbf{y}^T \mathbf{b} \leq \mathbf{c}^T \mathbf{x}$

proof in class

# Background: The Duality Theorem for Linear Programming

**Primal Linear Program:**Minimize  $\mathbf{c}^T \mathbf{x}$ where  $\mathbf{x} \geq \mathbf{0}$  satisfies:

$$\mathbf{Ax} \geq \mathbf{b}$$

**Dual Linear Program:**Maximize  $\mathbf{y}^T \mathbf{b}$ where  $\mathbf{y} \geq \mathbf{0}$  satisfies:

$$\mathbf{y}^T \mathbf{A} \leq \mathbf{c}^T$$

**Weak Duality Theorem** If  $\mathbf{x}, \mathbf{y}$  are feasible solutions then  $\mathbf{y}^T \mathbf{b} \leq \mathbf{c}^T \mathbf{x}$

proof in class

**Strong Duality Theorem** If  $\mathbf{x}^*, \mathbf{y}^*$  are optimal solutions then  $(\mathbf{y}^*)^T \mathbf{b} = \mathbf{c}^T \mathbf{x}^*$

# Fractional Edge Cover v.s. Fractional Vertex Packing

$$G = (V, E)$$

**Fractional Edge Cover:**

Minimize  $\sum_{e \in E} w_e$

where  $\mathbf{w} = (w_e)_{e \in E} \geq \mathbf{0}$  satisfies:

$$\forall x \in V : \sum_{e: x \in e} w_e \geq 1$$

# Fractional Edge Cover v.s. Fractional Vertex Packing

$$G = (V, E)$$

## Fractional Edge Cover:

$$\text{Minimize } \sum_{e \in E} w_e$$

where  $\mathbf{w} = (w_e)_{e \in E} \geq \mathbf{0}$  satisfies:

$$\forall x \in V : \sum_{e: x \in e} w_e \geq 1$$

## Fractional Vertex Packing:

$$\text{Maximize } \sum_{x \in V} v_x$$

where  $\mathbf{v} = (v_x)_{x \in V} \geq \mathbf{0}$  satisfies:

$$\forall e \in E : \sum_{x: x \in e} v_x \leq 1$$

# Fractional Edge Cover v.s. Fractional Vertex Packing

$$G = (V, E)$$

## Fractional Edge Cover:

$$\text{Minimize } \sum_{e \in E} w_e$$

where  $\mathbf{w} = (w_e)_{e \in E} \geq \mathbf{0}$  satisfies:

$$\forall x \in V : \sum_{e: x \in e} w_e \geq 1$$

## Fractional Vertex Packing:

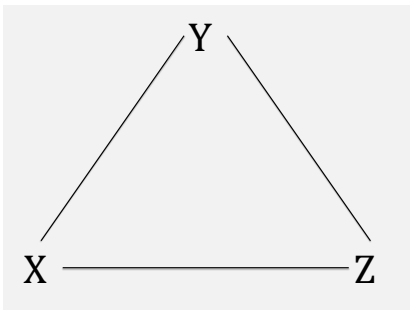
$$\text{Maximize } \sum_{x \in V} v_x$$

where  $\mathbf{v} = (v_x)_{x \in V} \geq \mathbf{0}$  satisfies:

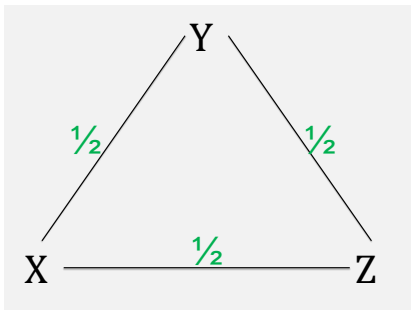
$$\forall e \in E : \sum_{x: x \in e} v_x \leq 1$$

The optimal values are equal:  $\sum_e w_e^* = \sum_x v_x^*$ .

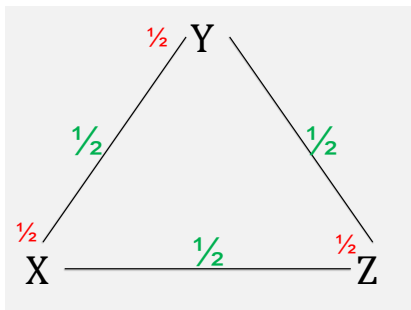
# Example



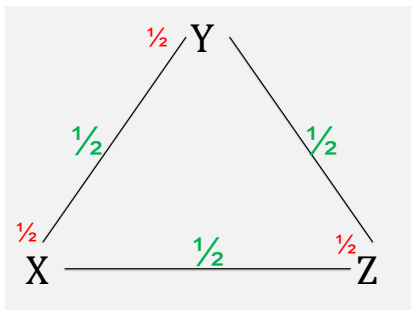
# Example



# Example



# Example



The optimal values are equal:  $\sum_e w_e^* = \sum_x v_x^* = 3/2$ .

We use  $\mathbf{v}^*$  to construct a worst-case instance for the query  $Q$ .

## The Worst-Case Instance By Example

$$R(X, Y) \wedge S(Y, Z) \wedge T(Z, X), \quad |R|, |S|, |T| \leq N, \quad \text{AGM}(Q) = N^{3/2}$$

# The Worst-Case Instance By Example

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

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

$$AGM(Q) = N^{3/2}$$

**Fractional Edge Cover:**Minimize  $w_R + w_S + w_T$ where  $\mathbf{w} \geq \mathbf{0}$  satisfies:

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

# The Worst-Case Instance By Example

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

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

$$AGM(Q) = N^{3/2}$$

## Fractional Edge Cover:

Minimize  $w_R + w_S + w_T$

where  $\mathbf{w} \geq \mathbf{0}$  satisfies:

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

## Fractional Vertex Packing:

Maximize  $v_X + v_Y + v_Z$

where  $\mathbf{v} \geq \mathbf{0}$  satisfies:

$$\begin{array}{rcll} R : & v_X + & v_Y & \leq 1 \\ S : & & v_Y + & v_Z \leq 1 \\ T : & v_X + & & v_Z \leq 1 \end{array}$$

# The Worst-Case Instance By Example

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

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

$$AGM(Q) = N^{3/2}$$

## Fractional Edge Cover:

Minimize  $w_R + w_S + w_T$

where  $\mathbf{w} \geq \mathbf{0}$  satisfies:

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

## Fractional Vertex Packing:

Maximize  $v_X + v_Y + v_Z$

where  $\mathbf{v} \geq \mathbf{0}$  satisfies:

$$\begin{array}{rcll} R : & v_X + & v_Y & \leq 1 \\ S : & & v_Y + & v_Z \leq 1 \\ T : & v_X + & & v_Z \leq 1 \end{array}$$

By strong duality:  $\rho^* = w_R^* + w_S^* + w_T^* = v_X^* + v_Y^* + v_Z^*$ .

# The Worst-Case Instance By Example

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

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

$$AGM(Q) = N^{3/2}$$

### Fractional Edge Cover:

Minimize  $w_R + w_S + w_T$

where  $\mathbf{w} \geq \mathbf{0}$  satisfies:

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

### Fractional Vertex Packing:

Maximize  $v_X + v_Y + v_Z$

where  $\mathbf{v} \geq \mathbf{0}$  satisfies:

$$\begin{array}{rcll} R : & v_X + & v_Y & \leq 1 \\ S : & & v_Y + & v_Z \leq 1 \\ T : & v_X + & & v_Z \leq 1 \end{array}$$

By strong duality:  $\rho^* = w_R^* + w_S^* + w_T^* = v_X^* + v_Y^* + v_Z^*$ .

Set  $s_x := \lfloor N^{v_x^*} \rfloor$ ,  $s_y := \lfloor N^{v_y^*} \rfloor$ ,  $s_z := \lfloor N^{v_z^*} \rfloor$ .

# The Worst-Case Instance By Example

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

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

$$AGM(Q) = N^{3/2}$$

### Fractional Edge Cover:

Minimize  $w_R + w_S + w_T$

where  $\mathbf{w} \geq \mathbf{0}$  satisfies:

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

### Fractional Vertex Packing:

Maximize  $v_X + v_Y + v_Z$

where  $\mathbf{v} \geq \mathbf{0}$  satisfies:

$$\begin{array}{rcll} R : & v_X + & v_Y & \leq 1 \\ S : & & v_Y + & v_Z \leq 1 \\ T : & v_X + & v_Z & \leq 1 \end{array}$$

By strong duality:  $\rho^* = w_R^* + w_S^* + w_T^* = v_X^* + v_Y^* + v_Z^*$ .

Set  $s_x := \lfloor N^{v_x^*} \rfloor$ ,  $s_y := \lfloor N^{v_y^*} \rfloor$ ,  $s_z := \lfloor N^{v_z^*} \rfloor$ .

$\mathbf{D}$  is a cross product:

$$R^D := [s_x] \times [s_y], S^D := [s_y] \times [s_z], T^D := [s_z] \times [s_x]$$

# The Worst-Case Instance By Example

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

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

$$AGM(Q) = N^{3/2}$$

### Fractional Edge Cover:

Minimize  $w_R + w_S + w_T$

where  $\mathbf{w} \geq \mathbf{0}$  satisfies:

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

### Fractional Vertex Packing:

Maximize  $v_X + v_Y + v_Z$

where  $\mathbf{v} \geq \mathbf{0}$  satisfies:

$$\begin{array}{rcll} R : & v_X + & v_Y & \leq 1 \\ S : & & v_Y + & v_Z \leq 1 \\ T : & v_X + & v_Z & \leq 1 \end{array}$$

By strong duality:  $\rho^* = w_R^* + w_S^* + w_T^* = v_X^* + v_Y^* + v_Z^*$ .

Set  $s_x := \lfloor N^{v_x^*} \rfloor$ ,  $s_y := \lfloor N^{v_y^*} \rfloor$ ,  $s_z := \lfloor N^{v_z^*} \rfloor$ .

$\mathbf{D}$  is a cross product:

$$R^D := [s_x] \times [s_y], S^D := [s_y] \times [s_z], T^D := [s_z] \times [s_x]$$

$$|Q(\mathbf{D})| = s_x \cdot s_y \cdot s_z \geq \frac{1}{8} N^{v_x^* + v_y^* + v_z^*} = \frac{1}{8} N^{\rho^*}$$

## Proof of the Lower Bound

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

**Theorem** [Lower Bound]

For every  $N$ , there exists an instance  $\mathbf{D} = (R_1^D, \dots, R_m^D)$  s.t.

$$|R_1^D|, \dots, |R_m^D| \leq N \quad \text{and} \quad Q(\mathbf{D}) \geq \frac{1}{2^{|\mathbf{X}|}} N^{\rho^*}$$

## Proof of the Lower Bound

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

**Theorem** [Lower Bound]

For every  $N$ , there exists an instance  $\mathbf{D} = (R_1^D, \dots, R_m^D)$  s.t.

$$|R_1^D|, \dots, |R_m^D| \leq N \quad \text{and} \quad Q(\mathbf{D}) \geq \frac{1}{2^{|\mathbf{X}|}} N^{\rho^*}$$

**Proof** By strong duality:  $\rho^* = \sum_j w_j^* = \sum_i v_i^*$ .

## Proof of the Lower Bound

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

**Theorem** [Lower Bound]

For every  $N$ , there exists an instance  $\mathbf{D} = (R_1^D, \dots, R_m^D)$  s.t.

$$|R_1^D|, \dots, |R_m^D| \leq N \quad \text{and} \quad Q(\mathbf{D}) \geq \frac{1}{2^{|\mathbf{X}|}} N^{\rho^*}$$

**Proof** By strong duality:  $\rho^* = \sum_j w_j^* = \sum_i v_i^*$ .

For every variable  $X_i$  set  $s_i := \lfloor N^{v_i^*} \rfloor$

## Proof of the Lower Bound

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

**Theorem** [Lower Bound]

For every  $N$ , there exists an instance  $\mathbf{D} = (R_1^D, \dots, R_m^D)$  s.t.

$$|R_1^D|, \dots, |R_m^D| \leq N \quad \text{and} \quad Q(\mathbf{D}) \geq \frac{1}{2^{|\mathbf{X}|}} N^{\rho^*}$$

**Proof** By strong duality:  $\rho^* = \sum_j w_j^* = \sum_i v_i^*$ .

For every variable  $X_i$  set  $s_i := \lfloor N^{v_i^*} \rfloor$

Define  $\mathbf{D}$  the cross product database:  $R_j^D := \times_{i: X_i \in \text{Vars}(R_j)} [s_i]$

$$|Q(\mathbf{D})| = \prod_{i=1, n} s_i \geq \frac{1}{2^n} \prod_i N^{v_i^*} = \frac{1}{2^n} N^{\rho^*}$$

# Statement of the Lower Bound: Non-Uniform Cardinalities

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

## Theorem [Lower Bound]

For all numbers  $N_1, \dots, N_m$  there exists  $\mathbf{D} = (R_1^D, \dots, R_m^D)$  s.t:

$$|R_1^D| \leq N_1, \dots, |R_m^D| \leq N_m \quad \text{and} \quad Q(\mathbf{D}) \geq \frac{1}{2^{|\mathbf{X}|}} \min_{\mathbf{w}} N_j^{w_j}$$

# Statement of the Lower Bound: Non-Uniform Cardinalities

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

## Theorem [Lower Bound]

For all numbers  $N_1, \dots, N_m$  there exists  $\mathbf{D} = (R_1^D, \dots, R_m^D)$  s.t:

$$|R_1^D| \leq N_1, \dots, |R_m^D| \leq N_m \quad \text{and} \quad Q(\mathbf{D}) \geq \frac{1}{2^{|\mathbf{X}|}} \min_{\mathbf{w}} N_j^{w_j}$$

The AGM bound  $\min_{\mathbf{w}} N_j^{w_j}$  is tight

# Proof of the AGM Lower Bound: Non-uniform Cardinalities

By example:

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

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

# Proof of the AGM Lower Bound: Non-uniform Cardinalities

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:

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

# Proof of the AGM Lower Bound: Non-uniform Cardinalities

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

$$v_X + v_Y + v_Z$$

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

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

# Proof of the AGM Lower Bound: Non-uniform Cardinalities

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

$$v_X + v_Y + v_Z$$

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

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

By strong duality:  $w_R^* + w_S^* + w_T^* = v_X^* + v_Y^* + v_Z^*$ .

Worst-case instance is the product database:

$$s_x \stackrel{\text{def}}{=} \lfloor 2^{v_X^*} \rfloor, s_y \stackrel{\text{def}}{=} \lfloor 2^{v_Y^*} \rfloor, s_z \stackrel{\text{def}}{=} \lfloor 2^{v_Z^*} \rfloor, R^D := [s_x] \times [s_y], S^D := [s_y] \times [s_z], T^D := [s_z] \times [s_x]$$

# Proof of the AGM Lower Bound: Non-uniform Cardinalities

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

$$v_X + v_Y + v_Z$$

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

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

By strong duality:  $w_R^* + w_S^* + w_T^* = v_X^* + v_Y^* + v_Z^*$ .

Worst-case instance is the product database:

$$s_x \stackrel{\text{def}}{=} \lfloor 2^{v_x^*} \rfloor, s_y \stackrel{\text{def}}{=} \lfloor 2^{v_y^*} \rfloor, s_z \stackrel{\text{def}}{=} \lfloor 2^{v_z^*} \rfloor, R^D := [s_x] \times [s_y], S^D := [s_y] \times [s_z], T^D := [s_z] \times [s_x]$$

$$|Q(D)| = s_x s_y s_z \geq \frac{1}{8} 2^{v_x^* + v_y^* + v_z^*}$$

# Proof of the AGM Lower Bound: Non-uniform Cardinalities

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

$$v_X + v_Y + v_Z$$

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

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

By strong duality:  $w_R^* + w_S^* + w_T^* = v_X^* + v_Y^* + v_Z^*$ .

Worst-case instance is the product database:

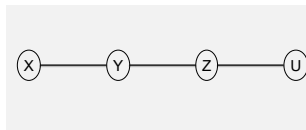
$$s_x \stackrel{\text{def}}{=} \lfloor 2^{v_X^*} \rfloor, s_y \stackrel{\text{def}}{=} \lfloor 2^{v_Y^*} \rfloor, s_z \stackrel{\text{def}}{=} \lfloor 2^{v_Z^*} \rfloor, R^D := [s_x] \times [s_y], S^D := [s_y] \times [s_z], T^D := [s_z] \times [s_x]$$

$$|Q(D)| = s_x s_y s_z \geq \frac{1}{8} 2^{v_X^* + v_Y^* + v_Z^*} = \frac{1}{8} 2^{w_R^* + w_S^* + w_T^*} = \frac{1}{8} AGM(Q)$$

# Examples

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

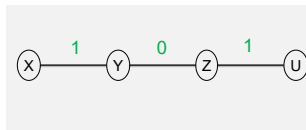
$$R(X, Y) \wedge S(Y, Z) \wedge T(Z, U)$$



# Examples

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

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

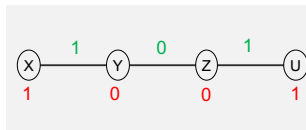


# Examples

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

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

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

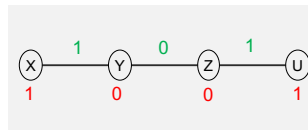


# Examples

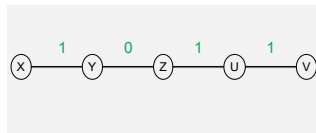
$$|R| = |S| = \dots = N$$

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

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



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

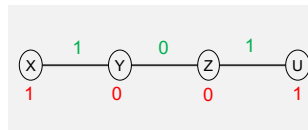


# Examples

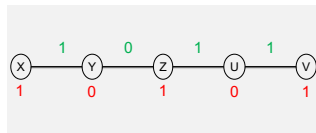
$$|R| = |S| = \dots = N$$

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

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



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

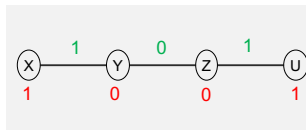


# Examples

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

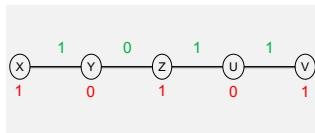
$$R(X, Y) \wedge S(Y, Z) \wedge T(Z, U)$$

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



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

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



# Extensions of the AGM Bound

# Extensions

- Use more stats: functional dependencies, domain cardinalities, and others.
  
  
  
  
  
  
  
  
  
  
- Beyond full queries: projections (same as Group-By).

## Domain Sizes

If  $|\text{Dom}(X)|$  is known, simply add it to the hypergraph.

## Domain Sizes

If  $|\text{Dom}(X)|$  is known, simply add it to the hypergraph.

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

We know  $|R|, |S|, |T|, |\text{Dom}(X)|$ .

## Domain Sizes

If  $|\text{Dom}(X)|$  is known, simply add it to the hypergraph.

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

We know  $|R|, |S|, |T|, |\text{Dom}(X)|$ .

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

## Domain Sizes

If  $|\text{Dom}(X)|$  is known, simply add it to the hypergraph.

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

We know  $|R|, |S|, |T|, |\text{Dom}(X)|$ .

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

Fractional edge covers:  $(0, \frac{1}{2}, \frac{1}{2}, \frac{1}{2})$  and  $(1, 0, 1, 0)$

$$AGM(Q) = \min((|R| \cdot |S| \cdot |T|)^{1/2}, |A| \cdot |S|)$$

## Simple Functional Dependencies

Given FDs,  $|Q| \ll AGM(Q)$ .

E.g.  $R(X, Y) \wedge S(Y, Z)$ :  $N^2$  becomes  $N$  when  $Y \rightarrow Z$ .

## Simple Functional Dependencies

Given FDs,  $|Q| \ll AGM(Q)$ .

E.g.  $R(X, Y) \wedge S(Y, Z)$ :  $N^2$  becomes  $N$  when  $Y \rightarrow Z$ .

$U \rightarrow V$  is **simple** if  $|U| = 1$ .

## Simple Functional Dependencies

Given FDs,  $|Q| \ll AGM(Q)$ .

E.g.  $R(X, Y) \wedge S(Y, Z)$ :  $N^2$  becomes  $N$  when  $Y \rightarrow Z$ .

$U \rightarrow V$  is **simple** if  $|U| = 1$ .

Method [Khamis et al., 2016]:

- **Expand**  $Q$  to  $Q^+$  by replacing each atom  $R(Y)$  with  $R'(Y^+)$ .
- Return  $AGM(Q^+)$ .
- This bound is tight. **Proof: very useful exercise.**

## Example

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

Fractional edge covers:  $(1, 1, 0)$ ,  $(1, 0, 1)$ ,  $(0, 1, 1)$ ,  $(1/2, 1/2, 1/2)$

$$|Q| \leq \min(|R| \cdot |S|, |R| \cdot |T|, |S| \cdot |T|, \sqrt{|R| \cdot |S| \cdot |T|})$$

## Example

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

Fractional edge covers:  $(1, 1, 0)$ ,  $(1, 0, 1)$ ,  $(0, 1, 1)$ ,  $(1/2, 1/2, 1/2)$

$$|Q| \leq \min(|R| \cdot |S|, |R| \cdot |T|, |S| \cdot |T|, \sqrt{|R| \cdot |S| \cdot |T|})$$

Assume that  $S.Y$  is a key:

$$Y \rightarrow Z$$

## Example

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

Fractional edge covers:  $(1, 1, 0)$ ,  $(1, 0, 1)$ ,  $(0, 1, 1)$ ,  $(1/2, 1/2, 1/2)$

$$|Q| \leq \min(|R| \cdot |S|, |R| \cdot |T|, |S| \cdot |T|, \sqrt{|R| \cdot |S| \cdot |T|})$$

Assume that  $S.Y$  is a key:  $Y \rightarrow Z$

$$Q^+(X, Y, Z) = R'(X, Y, Z) \wedge S(Y, Z) \wedge T(Z, X)$$

## Example

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

Fractional edge covers:  $(1, 1, 0)$ ,  $(1, 0, 1)$ ,  $(0, 1, 1)$ ,  $(1/2, 1/2, 1/2)$

$$|Q| \leq \min(|R| \cdot |S|, |R| \cdot |T|, |S| \cdot |T|, \sqrt{|R| \cdot |S| \cdot |T|})$$

Assume that  $S.Y$  is a key:  $Y \rightarrow Z$

$$Q^+(X, Y, Z) = R'(X, Y, Z) \wedge S(Y, Z) \wedge T(Z, X)$$

Fractional edge covers:  $(1, 0, 0)$ ,  $(0, 1, 1)$

$$|Q| \leq \min(|R|, |S| \cdot |T|)$$

# Discussion

The expansion procedure is very easy, but limited only to simple FDs:

$AGM(Q^+)$  is always an upper bound on  $Q$ 's output, but may not be tight.

## 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**. Later: extensions to **other stats**.



Atserias, A., Grohe, M., and Marx, D. (2013).  
Size bounds and query plans for relational joins.  
*SIAM J. Comput.*, 42(4):1737–1767.



Friedgut, E. (2004).  
Hypergraphs, entropy, and inequalities.  
*Am. Math. Mon.*, 111(9):749–760.



Khamis, M. A., Ngo, H. Q., and Suciu, D. (2016).  
Computing join queries with functional dependencies.  
In Milo, T. and Tan, W., editors, *Proceedings of the 35th ACM SIGMOD-SIGACT-SIGAI Symposium on Principles of Database Systems, PODS 2016, San Francisco, CA, USA, June 26 - July 01, 2016*, pages 327–342. ACM.