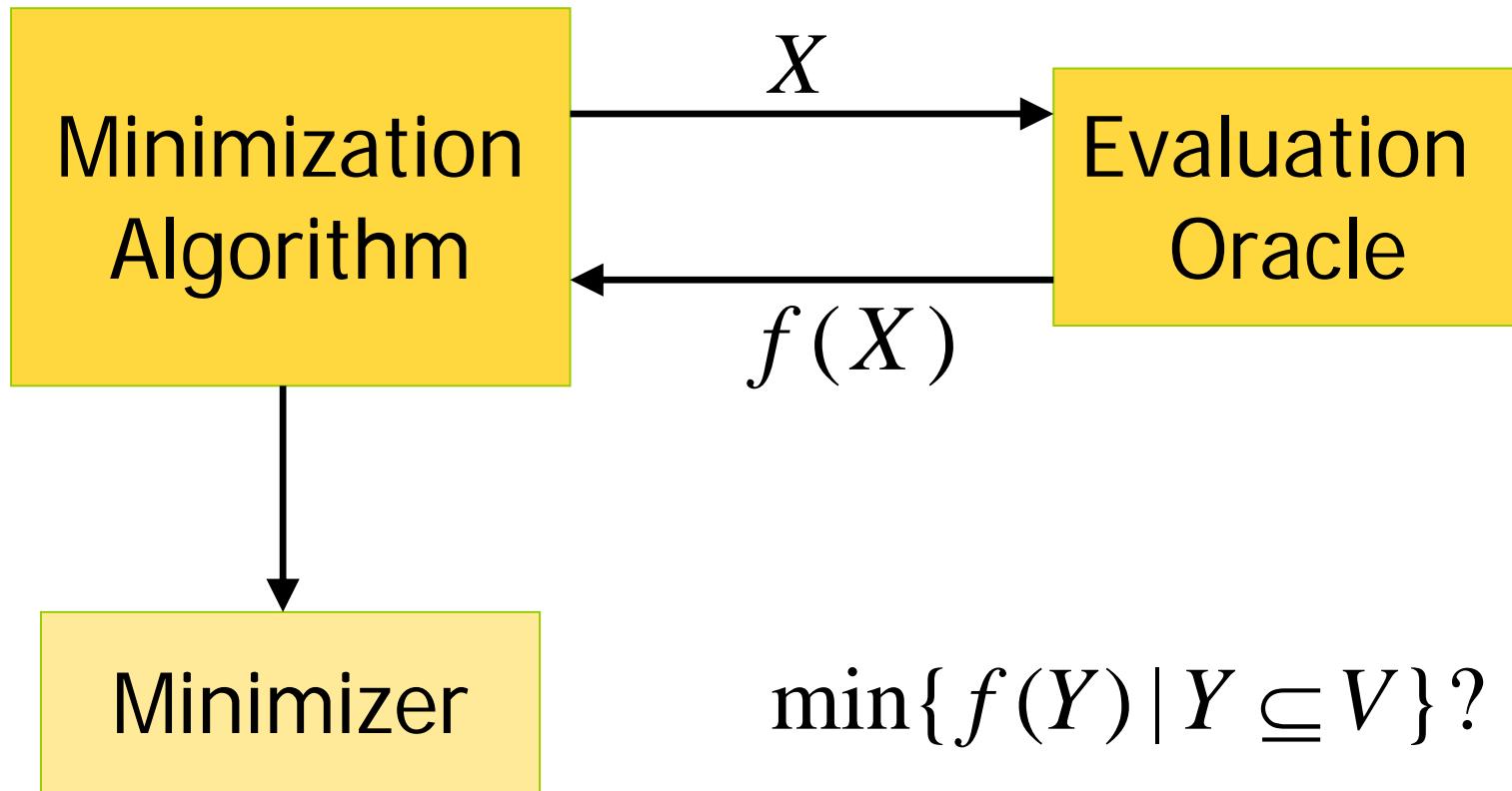


# Submodular Function Minimization

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# Submodular Function Minimization

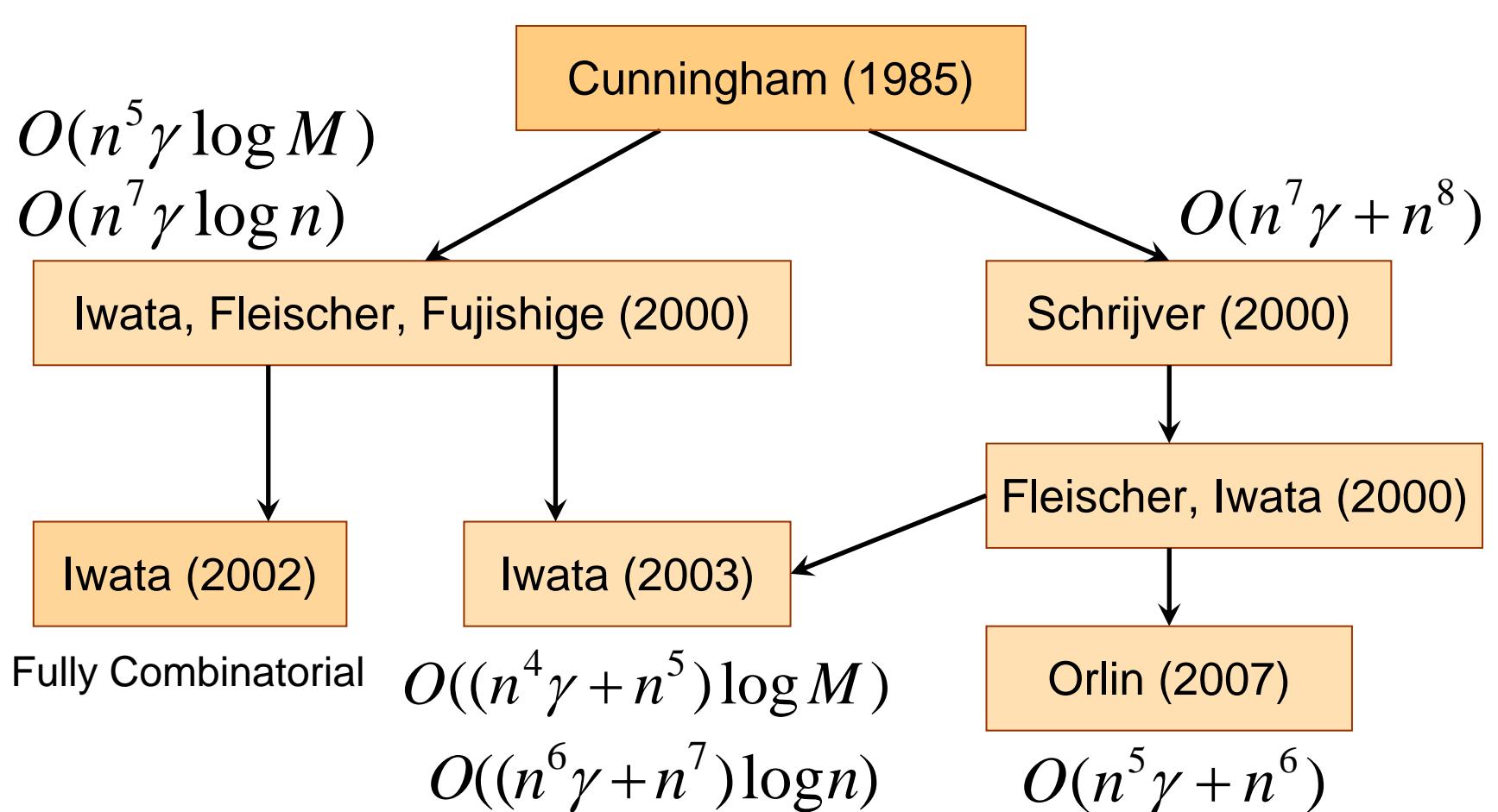
Assumption:  $f(\emptyset) = 0$



# Submodular Function Minimization

Grötschel, Lovász, Schrijver (1981, 1988)

Ellipsoid Method

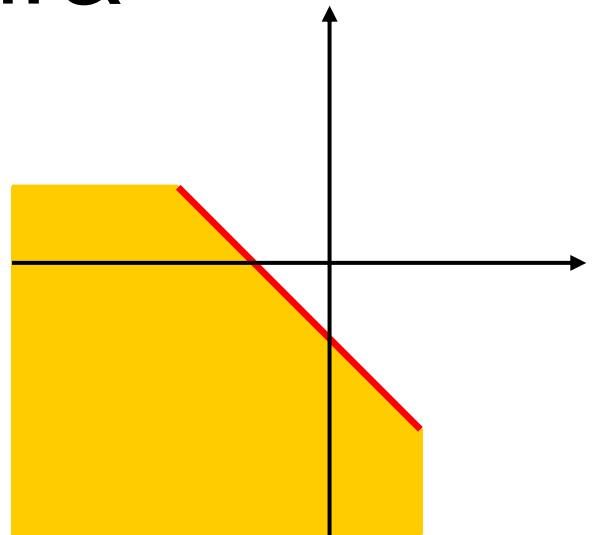


# Base Polyhedra

$$\mathbf{R}^V = \{x \mid V \rightarrow \mathbf{R}\}$$

$$x(Y) = \sum_{v \in Y} x(v)$$

$$f(\phi) = 0$$



Submodular Polyhedron

$$P(f) = \{x \mid x \in \mathbf{R}^V, \forall Y \subseteq V, x(Y) \leq f(Y)\}$$

Base Polyhedron

$$B(f) = \{x \mid x \in P(f), x(V) = f(V)\}$$

# Min-Max Theorem

Theorem

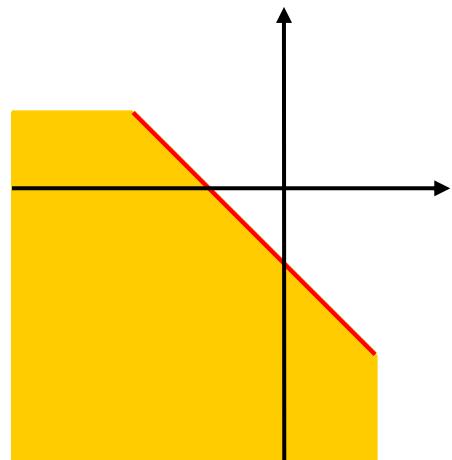
Edmonds (1970)

$$\begin{aligned}\min_{Y \subseteq V} f(Y) &= \max\{ z(V) \mid z \in P(f), z \leq 0 \} \\ &= \max\{ x^-(V) \mid x \in B(f) \}\end{aligned}$$

$$x^-(v) := \min\{0, x(v)\}$$

$$z(V) \leq z(Y) \leq f(Y)$$

$$x^-(V) \leq x(Y) \leq f(Y)$$



# Min-Max Theorem

$$\begin{aligned}\min_{Y \subseteq V} f(Y) &= \max\{ z(V) \mid z \in P(f), z \leq 0 \} \\ &= \max\{ x^-(V) \mid x \in B(f) \}\end{aligned}$$

$f^\circ(X) := \min\{f(Y) \mid Y \subseteq X\}$  Submodular

$$P(f^\circ) \subseteq P(f)$$

$$z \in B(f^\circ) \Rightarrow z(V) = f^\circ(V) = \min_{Y \subseteq V} f(Y)$$

# Combinatorial Approach

Extreme Base  $y_i \in B(f)$

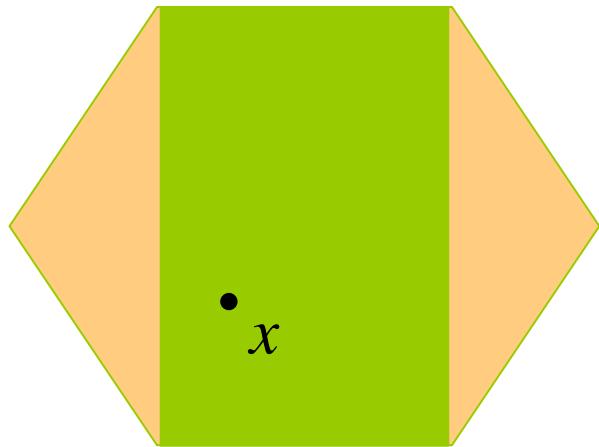
Convex Combination

$$x = \sum_{i \in I} \lambda_i y_i$$

Cunningham (1985)

$$O(n^6 M \gamma \log n M)$$

$$M = \max_{X \subseteq V} |f(X)|$$



Gaussian Elimination

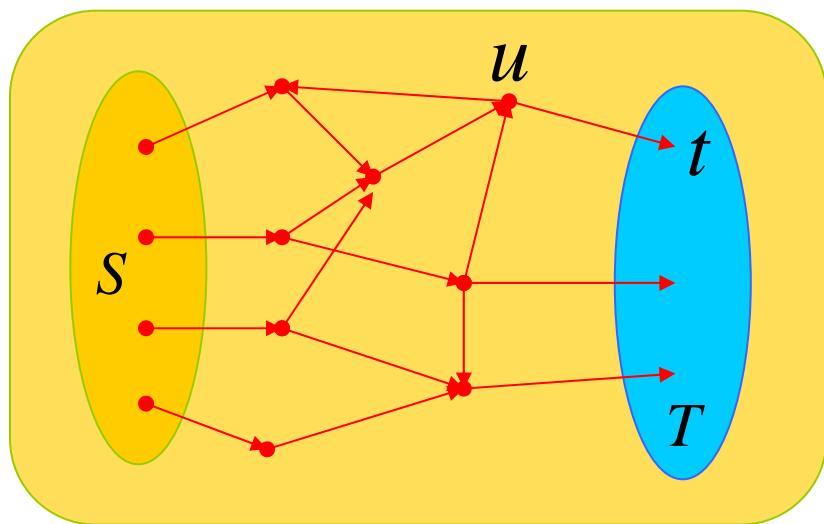
# Schrijver's Algorithm

$$x = \sum_{i \in I} \lambda_i y_i$$



$$G_I = (V, A_I)$$

$$A_I = \{(u, v) \mid \exists i \in I, v \prec_i u\}$$



$$S = \{v \mid x(v) < 0\} \quad T = \{v \mid x(v) > 0\}$$

$d(v)$ : distance from  $S$

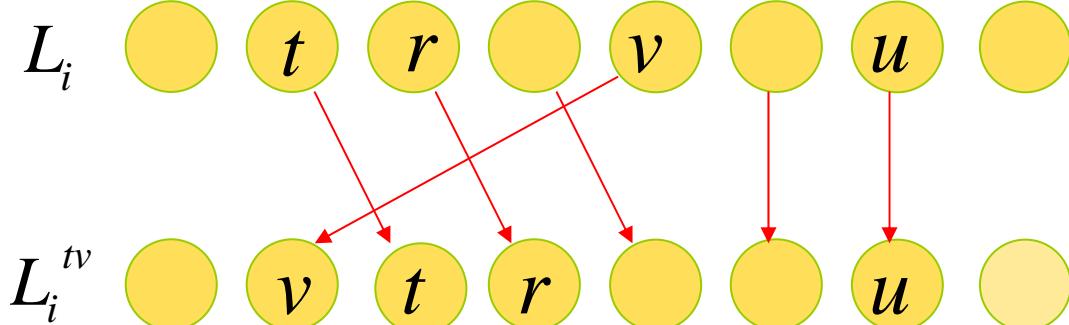
$d(t)$ : maximum in  $T$

$$d(u) = d(t) - 1$$

$$x(u) := x(u) + \alpha$$

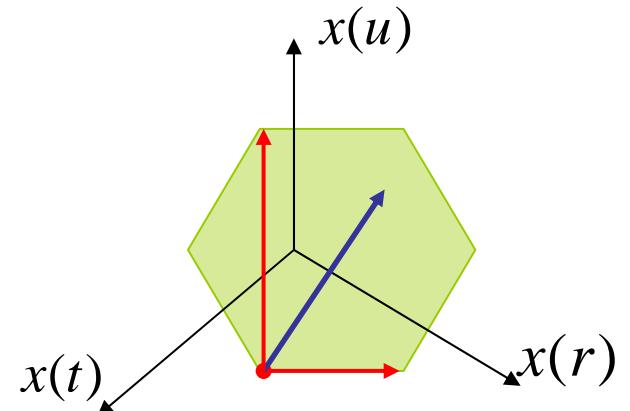
$$x(t) := x(t) - \alpha$$

# Schrijver's Algorithm



$$\begin{array}{cc}
 r & u \\
 \begin{matrix}
 t & - & \cdots & - \\
 r & - & \cdots & - \\
 0 & + & \ddots & \vdots \\
 \vdots & \ddots & \ddots & - \\
 u & 0 & \cdots & 0
 \end{matrix} &
 \begin{matrix}
 \xi_r \\
 \vdots \\
 \xi_u
 \end{matrix} = \begin{matrix}
 -1 \\
 0 \\
 \vdots \\
 0 \\
 1
 \end{matrix}
 \end{array}$$

$y_i^{tv} - y_i$ 
 $y_i^{tv} - y_i$ 
 $\xi \geq 0$



$$\alpha := \min\{x(u), \frac{\lambda_i}{\sum_v \xi_v}\}$$

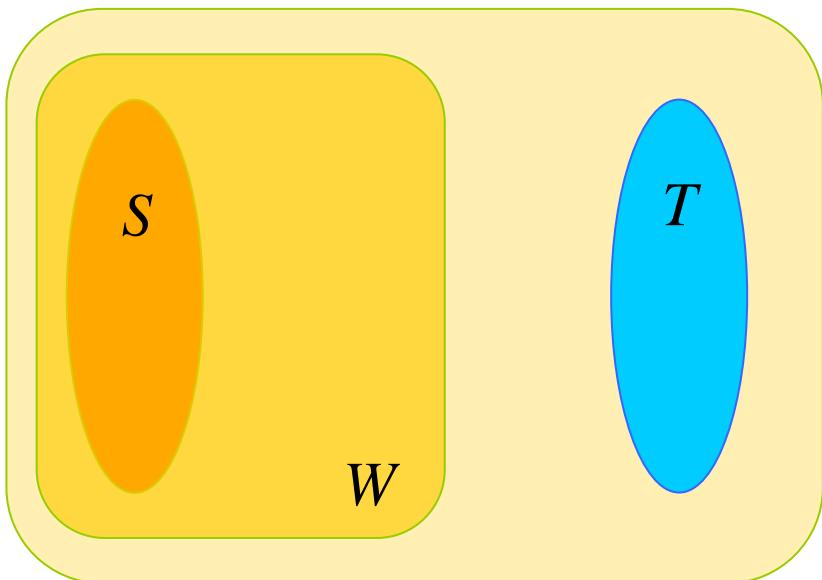
$$\lambda_{i_v} := \alpha \xi_v$$

$$\lambda_i := \lambda_i - \alpha \sum_v \xi_v$$

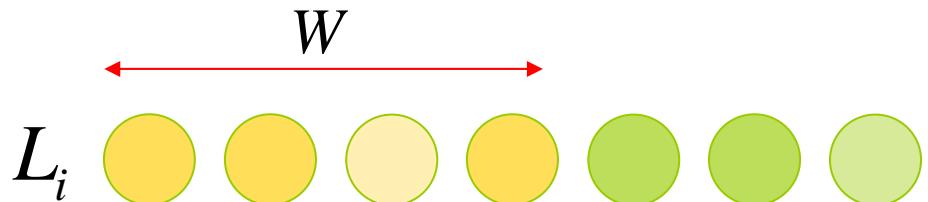
# Schrijver's Algorithm

No Path from  $S$  to  $T$

$$W := \{v \mid v : \text{Reachable from } S\}$$



$$S = \{v \mid x(v) < 0\} \quad T = \{v \mid x(v) > 0\}$$



$$y_i(W) = f(W), \quad \forall i \in I$$

$$\frac{x^-(V) = x(W)}{= \sum_{i \in I} \lambda_i y_i(W) = \underline{f(W)}}$$

$$O(n^7\gamma + n^8)$$

# IFF Scaling Algorithm

$$\delta \approx \frac{M}{n^2} \longrightarrow \delta < \frac{1}{n^2}$$

$\left\lfloor \frac{f_\delta(X)}{\delta} \right\rfloor$ : Submodular

$$f_\delta(X) = f(X) + \underline{\delta |X| \cdot |V \setminus X|}$$

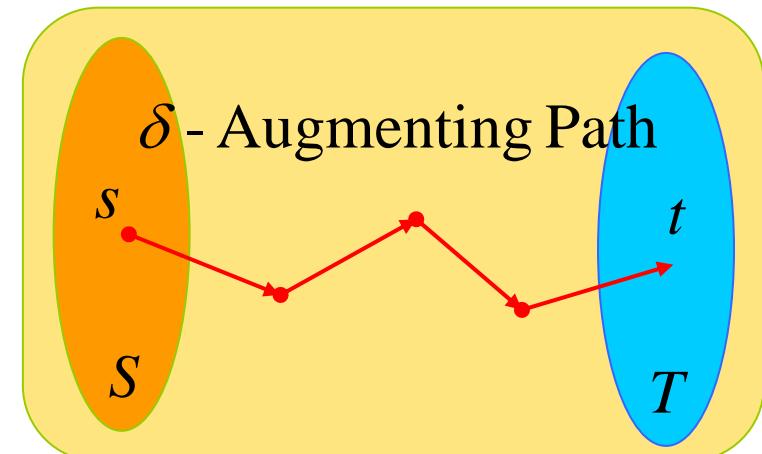
Cut Function

$\varphi$ : Flow in the Complete Digraph  $\varphi(u, v) \leq \delta$

$$x = \sum_{i \in I} \lambda_i y_i$$

$$z = x + \partial \varphi$$

Increase  $z^-(V)$

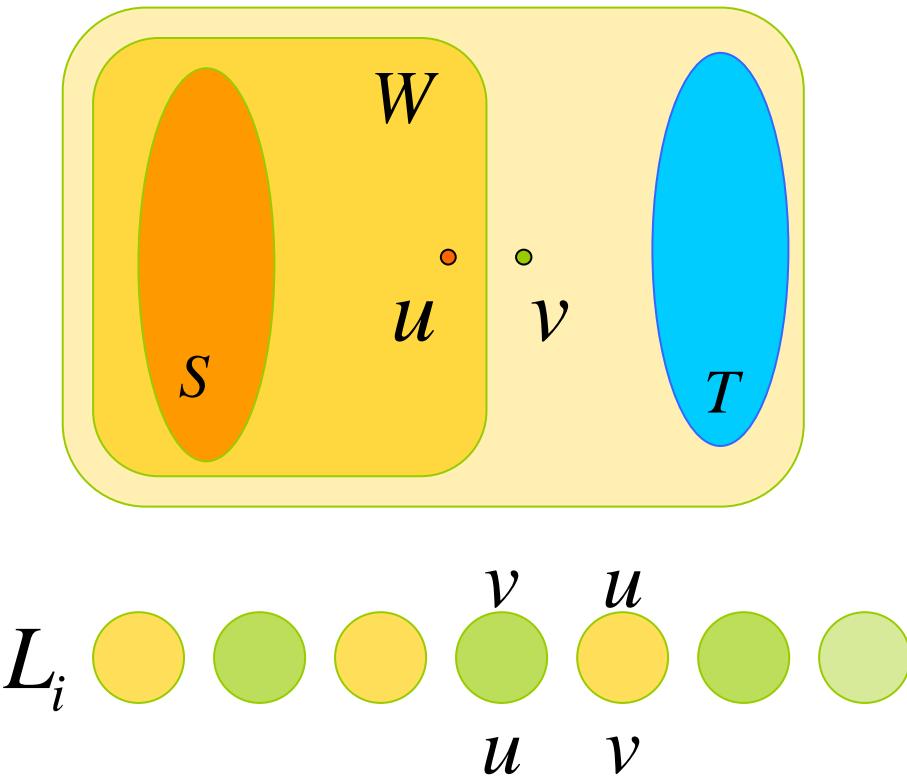


$$S = \{v \mid z(v) < -\delta\} \quad T = \{v \mid z(v) > \delta\}$$

# IFF Scaling Algorithm

No Path from  $S$  to  $T$

$$W := \{v \mid v : \text{Reachable from } S\}$$



Double-Exchange

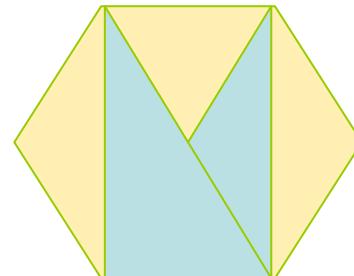
$$y_i := y_i + \beta(\chi_u - \chi_v)$$

$$x := x + \alpha(\chi_u - \chi_v)$$

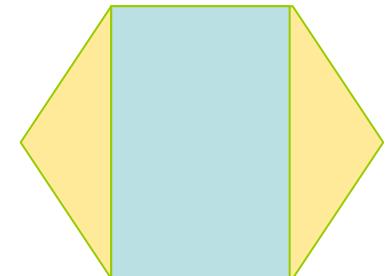
$$\alpha := \min\{\delta, \lambda_i \beta\}$$

$$\alpha = \lambda_i \beta$$

$$\alpha < \lambda_i \beta$$



Saturating

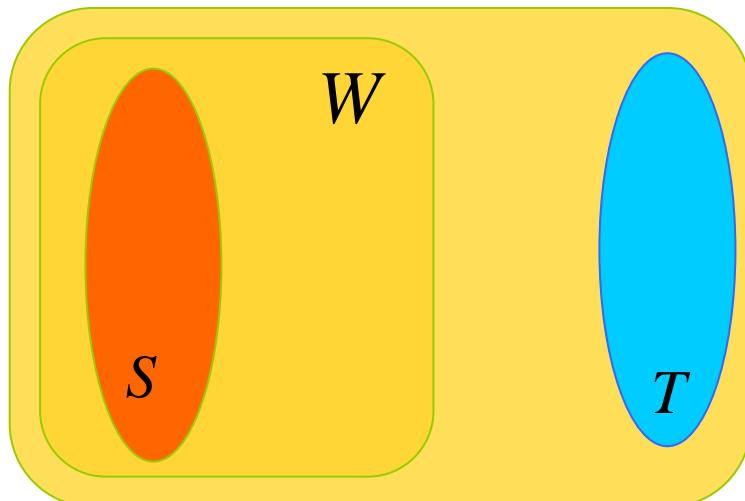


Nonsaturating

# IFF Scaling Algorithm

No Path from  $S$  to  $T$

No Active Triple( $i, u, v$ )



$$y_i(W) = f(W), \forall i \in I$$

$$x(W) = \sum_{i \in I} \lambda_i y_i(W) = f(W)$$

$$z^-(V) \geq f(W) - n\delta$$

$$x^-(V) \geq f(W) - n^2\delta$$

$$\delta < \frac{1}{n^2} \implies f(W) : \text{Min}$$

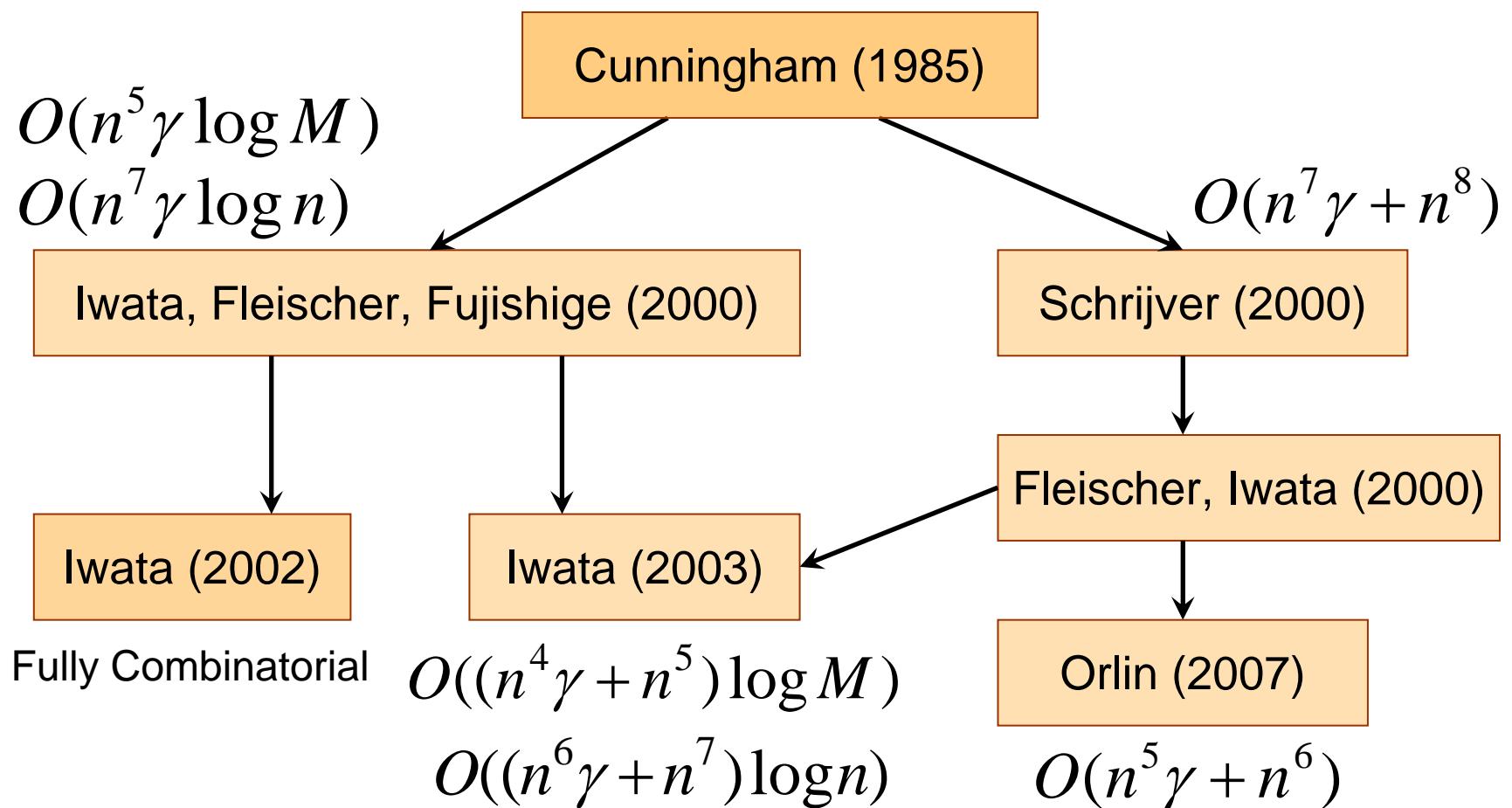
$$L_i$$

$$O(n^5 \gamma \log M)$$

# Submodular Function Minimization

Grötschel, Lovász, Schrijver (1981, 1988)

Ellipsoid Method



# Open Problems

- A Fully Combinatorial SFM Algorithm without Scaling.
- A Lower Bound on the Number of Oracle Calls for General SFM.

# Fully Combinatorial Algorithm

Addition, Subtraction, Comparison  
Oracle Call (Function Evaluation)

Multiplication  $\alpha \in \mathbf{R}, \mu \in \mathbf{Z} \longrightarrow$  Compute  $\mu\alpha$

Division  $\alpha \in \mathbf{R}, \beta \in \mathbf{R} \longrightarrow$  Compute  $q = \lceil \alpha / \beta \rceil$

- Neglect the **Gaussian Elimination Step**.
- Use **Nonnegative Integer** Combination Instead of Convex Combination.
- Choose a **Step Length** Appropriately.

# Pointers

- S. Fujishige: *Submodular Functions and Optimization*, North-Holland, 2005.
- S. T. McCormick: Submodular Function Minimization, *Discrete Optimization*, K. Aardal, G. Nemhauser, R. Weismantel, eds., Handbooks in Operations Research, Elsevier, 2005.

# Pointers

- S. Iwata: Submodular Function Minimization,  
*Mathematical Programming*, 112 (2008),  
pp. 45-64