

Optimal Distributed Covering Algorithms

Ran Ben-Basat ¹, Guy Even ², Ken-ichi Kawarabayashi ³, Gregory Schwartzman ³

¹Harvard

²Tel-Aviv U.

³ NII

accepted to DISC

Min Weight Hitting Set (Hypergraph Vertex Cover)

Input:

n - # elements

$e_1, \dots, e_m \subseteq [n]$ - sets

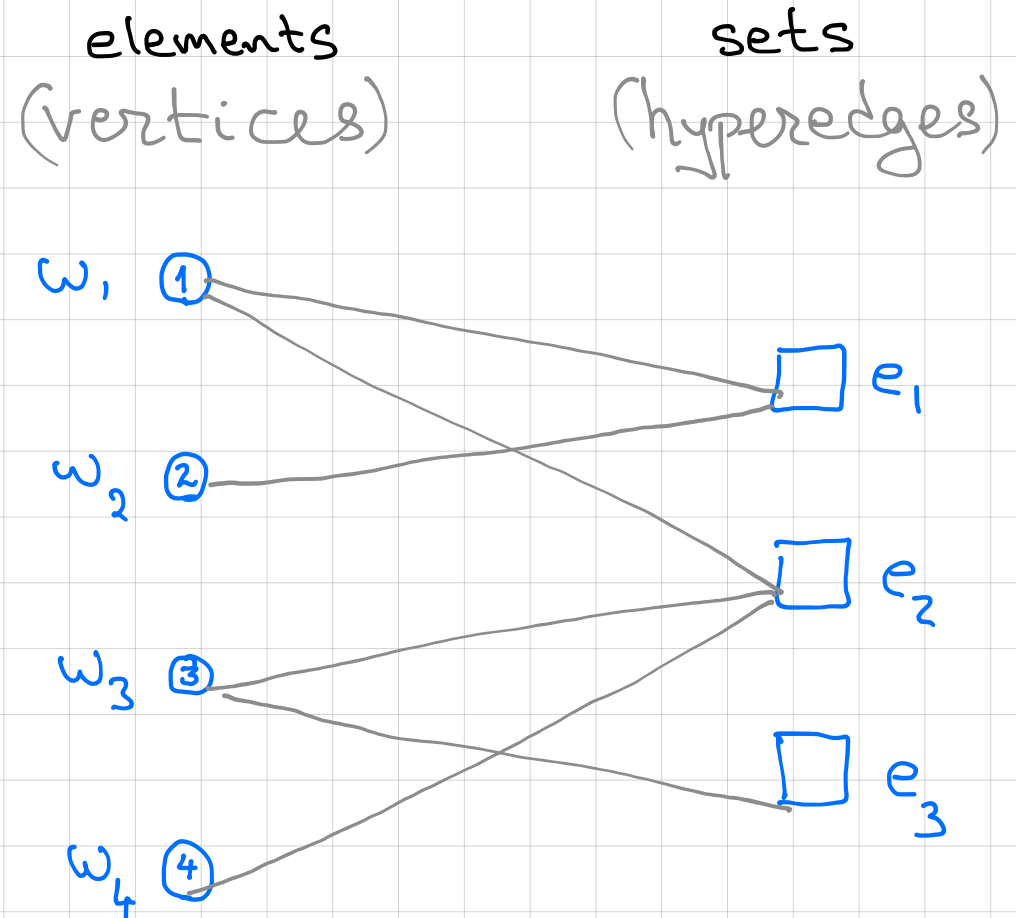
$w_1, \dots, w_n \in \mathbb{N}^+$ - weights

Hitting Set:

$X \subseteq [n]$ s.t.

$\forall i \ e_i \cap X \neq \emptyset$

Goal: $\min w(X)$



Min Weight Hitting Set (Hypergraph Vertex Cover)

distributed setting:

- bipartite network of vertices & hyperedges equipped with "ports"

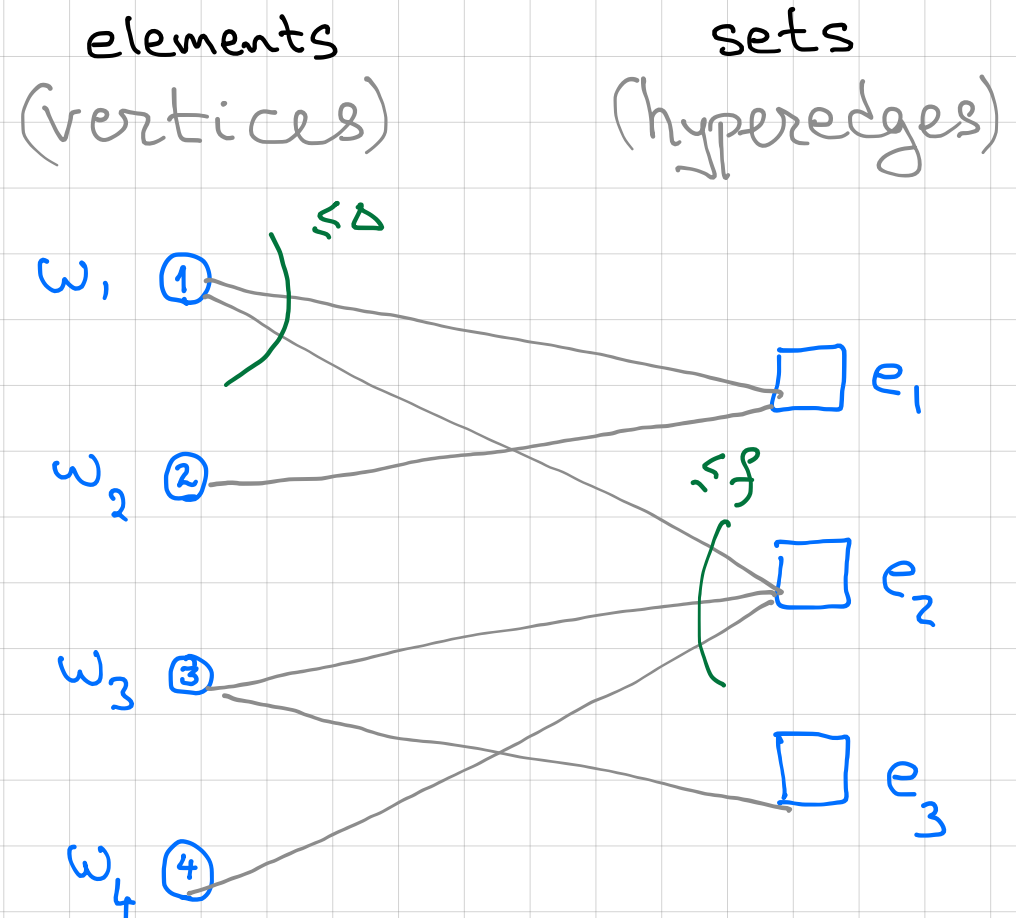
- local input: w_i

- know: Δ, f, ε

(can do without Δ, f)

- CONGEST

$\ast m = \text{poly}(n), w_{\max} = \text{poly}(n)$



Theorem ([KMW16])

Any distributed constant-factor approximation algorithm requires $\Omega(\log \Delta / \log \log \Delta)$ rounds to terminate.

lower bound holds for:

- every constant approximation ratio
- unweighted graphs, and
- even if the message lengths are not bounded.

Weighted Graph Vertex Cover Results

det.	weighted	approximation	time	algorithm
yes	no	3	$O(\Delta)$	[PS09]
yes	no	2	$O(\Delta^2)$	[Ast+09]
yes	yes	2	$O(1)$ for $\Delta \leq 3$	[Ast+09]
yes	yes	2	$O(\Delta + \log^* n)$	[PR01]
yes	yes	2	$O(\Delta + \log^* W)$	[AS10]
yes	yes	2	$O(\log^2 n)$	[KVY94]
yes	yes	2	$O(\log n \log \Delta / \log^2 \log \Delta)$	[Ben+18]
no	yes	2	$O(\log n)$	[GKP08; KY11]
yes	yes	2	$O(\log n)$	This work
yes	yes	$2 + \epsilon$	$O(\epsilon^{-4} \log(W \cdot \Delta))$	[Hoc82; KMW06]
yes	yes	$2 + \epsilon$	$O(\log \epsilon^{-1} \log n)$	[KVY94]
yes	yes	$2 + \epsilon$	$O(\epsilon^{-1} \log \Delta / \log \log \Delta)$	[BCS17; EGM18]
yes	yes	$2 + \epsilon$	$O\left(\frac{\log \Delta}{\log \log \Delta} + \frac{\log \epsilon^{-1} \log \Delta}{\log^2 \log \Delta}\right)$	[Ben+18]
yes	yes	$2 + \epsilon$	$O\left(\frac{\log \Delta}{\log \log \Delta} + \log \epsilon^{-1} \cdot (\log \Delta)^{0.001}\right)$	This work
yes	yes	$2 + \frac{\log \log \Delta}{c \cdot \log \Delta}$	$O(\log \Delta / \log \log \Delta)$	[BCS17], $\forall c = O(1)$
yes	yes	$2 + (\log \Delta)^{-c}$	$O(\log \Delta / \log \log \Delta)$	[Ben+18], $\forall c = O(1)$
yes	yes	$2 + 2^{-c \cdot (\log \Delta)^{0.99}}$	$O(\log \Delta / \log \log \Delta)$	This work, $\forall c = O(1)$

Weighted Hypergraph Vertex Cover Results

weighted	approximation	time	algorithm
yes	f	$O(f^2 \Delta^2 + f \Delta \log^* W)$	[AS10]
yes	f	$O(f \log^2 n)$	[KVV94]
yes	f	$O(f \log n)$	This work
no	$f + \epsilon$	$O\left(\epsilon^{-1} \cdot f \cdot \frac{\log(f\Delta)}{\log \log(f\Delta)}\right)$	[EGM18] ¹
yes	$f + \epsilon$	$O(f \cdot \log(f/\epsilon) \cdot \log n)$	[KVV94]
yes	$f + \epsilon$	$O(\epsilon^{-4} \cdot f^4 \cdot \log f \cdot \log(W \cdot \Delta))$	[KMW06]
yes	$f + \epsilon$	$O\left(f \cdot \log(f/\epsilon) \cdot (\log \Delta)^{0.001} + \frac{\log \Delta}{\log \log \Delta}\right)$	This work
no	$f + 1/c$	$O(\log \Delta / \log \log \Delta)$	[EGM18], $\forall f, c = O(1)$
yes	$f + 2^{-c \cdot (\log \Delta)^{0.99}}$	$O(\log \Delta / \log \log \Delta)$	This work, $\forall f, c = O(1)$

No depend. on n, W
 depend. on ϵ : $\lg \frac{1}{\epsilon}$

Primal Dual Scheme

LP relaxation:

$$\begin{aligned} & \min \sum_v \omega_v \cdot x_v \\ \text{s.t.} \quad & \sum_{v \in e} x_v \geq 1 \quad \forall e \quad \leftarrow \text{covering constraint} \\ & (1 \geq) \quad x_v \geq 0 \quad \forall v \end{aligned}$$

Dual LP:

$$\begin{aligned} & \max \sum_e \delta_e \\ \text{s.t.} \quad & \sum_{e \ni v} \delta_e \leq \omega_v \quad \forall v \quad \leftarrow \text{packing constr.} \\ & \delta_e \geq 0 \quad \forall e \end{aligned}$$

Primal Dual Scheme

DEF: v is β -tight if

$$\beta \cdot w(v) \geq w(v) - \sum_{e \ni v} \delta(e) \quad \leftarrow \text{slack}(v)$$

$$\text{set } \beta \stackrel{\Delta}{=} \frac{\varepsilon}{f + \varepsilon}$$

THM: if $\{\delta_e\}_{e \in E}$ feas

then $w(\beta\text{-Tight}) \leq (f + \varepsilon) \cdot \text{OPT}_{\text{frac}}$

ALG: find $\{\delta_e\}_e$ feas s.t. β -Tight is a VC
apx ratio (ALG) $\leq f + \varepsilon$

$$\min \sum_v w_v \cdot x_v$$

$$\text{s.t. } \sum_{v \in e} x_v \geq 1 \quad \forall e$$

$$x_v \geq 0 \quad \forall v$$

$$\max \sum_e \delta_e$$

$$\text{s.t. } \sum_{e \ni v} \delta_e \leq w_v \quad \forall v$$

$$\delta_e \geq 0 \quad \forall e$$

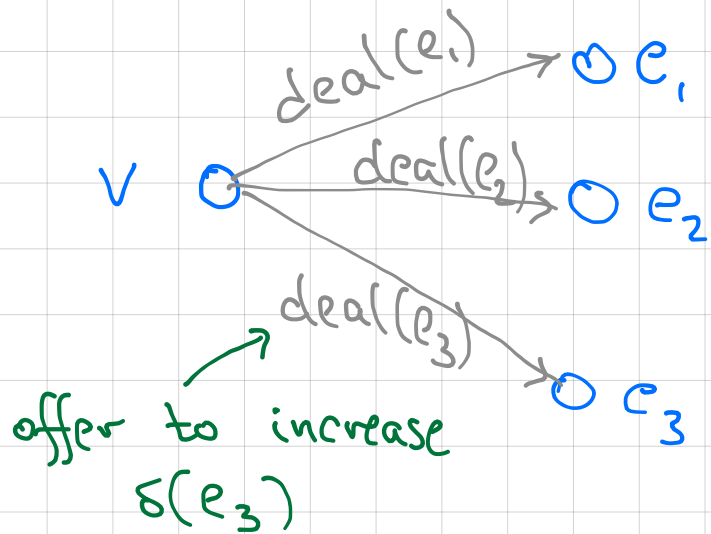
ALG Termination Rules

- * v joins cover & terminates if becomes tight
- * e terminates if $\exists v \in e \cap \text{cover}$
- * v terminates (& does not join cover)
if $\forall e \ni v$ terminated

ALG terminates if $\forall v$ & $\forall e$ terminated

Dual Solution Dynamics

$$\sum_{e \ni v} \delta_e \leq w_v \quad \forall v$$



v sends $deal(e_i)$ to e_i .
 update $\delta(e_i) \leftarrow \delta(e_i) + deal(e_i)$

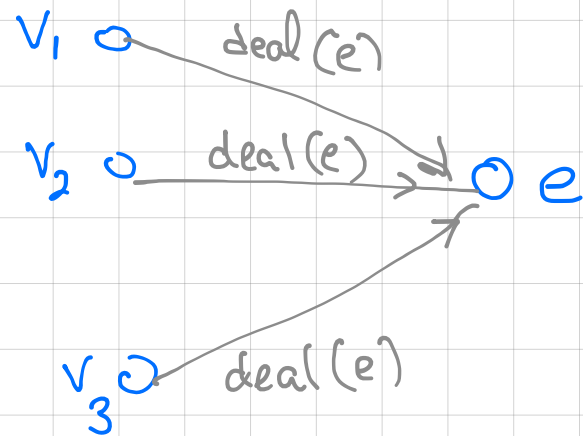
is v -safe if

$$\sum_{e_i \ni v} \delta(e_i) + \sum_i deal(e_i) \leq w(v)$$

e recv's $deal(e)$

$$\delta(e) \leftarrow \delta(e) + deal(e)$$

* e recv's same deal $\forall v \in e$



deal(e) Dynamics

$$\left. \begin{array}{l} \text{Init:} \\ \text{deal}(e) \end{array} \right\} \beta \cdot \min_{v \in e} \frac{w(v)}{\deg(v)} \quad \left. \begin{array}{l} \text{max} \\ \text{deal}(e) \end{array} \right\} \beta \cdot w(v) \quad \frac{\text{max}}{\text{init}} \leq \Delta$$

$$\text{deal}_{i+1}(e) = \begin{cases} \alpha \cdot \text{deal}_i(e) & \text{if } \forall v \in e \setminus \text{Tight} \\ & \sum_{\substack{e' \ni v \\ e' \text{ not covered}}} \text{deal}_i(e') \leq \frac{\beta}{\alpha} \cdot w(v) \\ \text{deal}_i(e) & \text{Otherwise} \end{cases}$$

RAISE:

$$\text{deal}_{i+1}(e) = \alpha \cdot \text{deal}_i(e)$$

STUCK:

$$\text{deal}_{i+1}(e) = \text{deal}_i(e)$$

$$\text{deal}_{i+1}(e) = \begin{cases} \alpha \cdot \text{deal}_i(e) & \text{if } \forall v \in e \setminus \text{Tight} \\ & \sum_{\substack{e' \ni v \\ e' \text{ not covered}}} \text{deal}_i(e') \leq \frac{\beta}{\alpha} \cdot w(v) \\ \text{deal}_i(e) & \text{Otherwise} \end{cases}$$

$$\# \text{ Raises} \leq \log_{\alpha} \frac{\text{MAX}}{\text{INIT}} \leq \log_{\alpha} \Delta$$

$$\# \text{ STUCKS} \leq f \cdot \frac{\alpha}{\beta} \quad (\forall v \text{ becomes tight after } \frac{\alpha}{\beta})$$

$$\Rightarrow \# \text{ RNDS} \leq \log_{\alpha} \Delta + f \cdot \frac{\alpha}{\beta}$$

$$\alpha = \frac{\lg \Delta}{\lg \lg \Delta}$$

$$\Rightarrow \# \text{ RNDS} \leq O\left(\frac{f^2}{\lg \lg \Delta} \cdot \frac{\lg \Delta}{\lg \lg \Delta}\right)$$

Reduced Depend. $\frac{1}{\epsilon} \rightarrow \lg \frac{1}{\epsilon}$

* $\text{slack}(v) \in [\beta \cdot w(v), w(v)]$

* $\lg \frac{1}{\beta}$ "levels" per v accord. to $\log \frac{w(v)}{\text{slack}(v)}$

* within each level $\# \text{stuck}(v) \leq \alpha$

$\Rightarrow \# \text{stuck} \leq \# \text{levels} \times |e| \times \alpha$

$\leq (\lg \frac{1}{\beta}) \cdot f \cdot \alpha$

LIE!

$\Rightarrow \# \text{RNDS} \leq \lg_{\alpha} \Delta + (\lg \frac{1}{\beta}) \cdot f \cdot \alpha$

$\Rightarrow \alpha \approx \frac{\lg \Delta}{\lg \lg \Delta} : \# \text{RNDS} \leq O\left(f \cdot \lg \left(\frac{1}{\beta}\right) \cdot \frac{\lg \Delta}{\lg \lg \Delta}\right)$

$$f - \alpha p x$$

$$\text{If } \varepsilon \approx \frac{1}{n \cdot w_{\max}} = \frac{1}{\text{poly}(n)}$$

then $(f + \varepsilon) - \alpha p x$ is $f - \alpha p x$.

$$\# \text{ rnds} : O(f \cdot \lg n)$$

(Req. fine tuning of α)

