

# Matroid Intersection under Restricted Oracles

Yutaro Yamaguchi

Collaborators: Kristóf Bérczi, Tamás Király, Yu Yokoi

Special Thanks: Mihály Bárász, Yuni Iwamasa, Taihei Oki

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**Q. Is “Matroid Intersection” tractable? In what sense?**

K. Bérczi, T. Király, Y. Yamaguchi, Y. Yokoi:

**Matroid Intersection under Restricted Oracles.**

*SIAM Journal on Discrete Mathematics (SIDMA).* To appear. (arXiv:2209.14516)

M. Bárász, K. Bérczi, T. Király, Y. Yamaguchi, Y. Yokoi:

**Matroid Intersection under Minimum Rank Oracle.** In preparation.

(Including and Extending

M. Bárász: **Matroid Intersection for the Min-Rank Oracle.**

*EGRES Technical Report, QP-2006-03, 2006.*)

**To be continued...**

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*EGRES Technical Report*, QP-2006-03, 2006.)

To be continued... and there are also spin-off papers:

K. Bérczi, T. Király, Y. Yamaguchi, Y. Yokoi:  
**Approximation by Lexicographically Maximal Solutions in Matching and Matroid Intersection Problems.**  
*Theoretical Computer Science*, 910 (2022), pp. 48–53.

K. Bérczi, T. Király, T. Schwarcz, Y. Yamaguchi, Y. Yokoi: **Hypergraph Characterization of Split Matroids.**  
*Journal of Combinatorial Theory, Series A*, 194 (2023), No. 105697.

# Outline

- Overview: Question and Results
- Matroid Intersection (Basics)
  - Matroid and Matroid Intersection
  - Augmenting-Path Algorithms and Exchangeability Graph
- Matroid Intersection under Restricted Oracles
  - First Step: What can be done in general by Common Independence Oracle
  - Results on Each Restricted Oracle
- Matroid Intersection under Minimum Rank Oracle
  - How to Solve Unweighted Problem
  - Results on Weighted Problem

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# What is Matroid Intersection?

## The intersection of two matroids

- Efficient Algorithms and Max-Min Theorems
  - A maximum-cardinality common independent set
  - A maximum-weight common independent set (of each cardinality)
- LP Formulation
  - Intersection of matroid polytopes = matroid intersection polytope
  - Total dual integrality (TDI) and well-structured dual solution
- Many Applications (= Unified Framework)  
Bipartite matching, Arborescence (packing), Dijoin, etc.

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Most of them require  
separate information

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## What is known

- Efficient Algorithms rely on **separate** oracles for the two matroids
- Max-Min Theorems and the Polyhedral Description are given by using the two rank functions **separately**

## What may be asked

- The resulting combinatorial structure is just  $\mathcal{I}_1 \cap \mathcal{I}_2$
- The polytope is completely determined by  $r_{\min} = \min\{r_1, r_2\}$
- When it is seen as a special case of Matroid Matching, the input should be  $r_{\text{sum}} = r_1 + r_2$  (oracle)

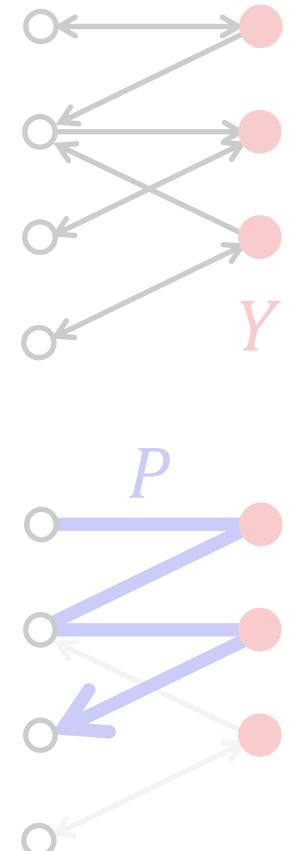
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Most of them require separate information of two matroids

## Basic Strategy of Efficient Algorithms

- Starting with  $Y = \emptyset$ , repeatedly update the current solution  $Y$ .
- For each update,
  - construct the exchangeability graph w.r.t.  $Y$ ,
  - find an augmenting path  $P$  in the graph, and
  - flip the current solution along the path, i.e.,  $Y \leftarrow Y \Delta P$ .
- The edges in the graph are oriented according to **in which matroid** the two elements are exchangeable.

Assumption: **Independence in each matroid** can be tested.



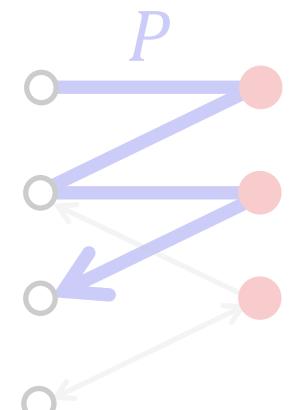
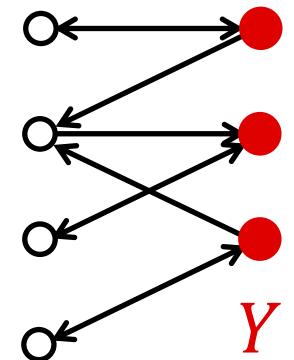
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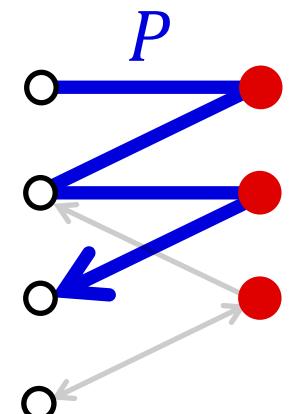
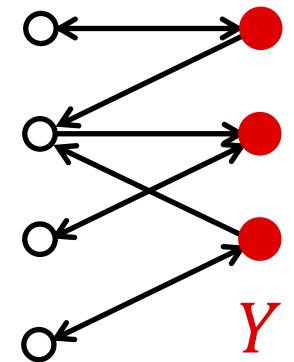
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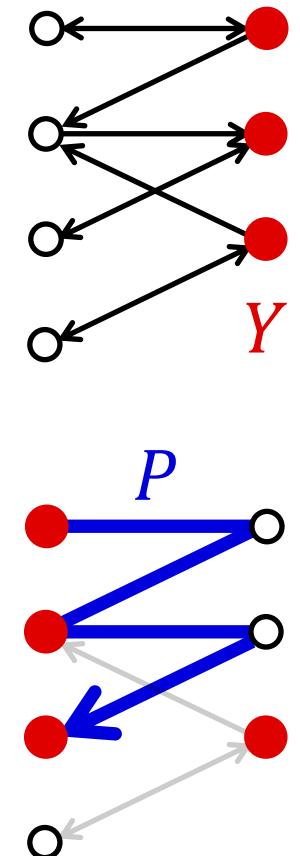
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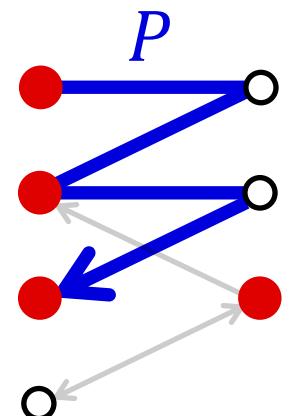
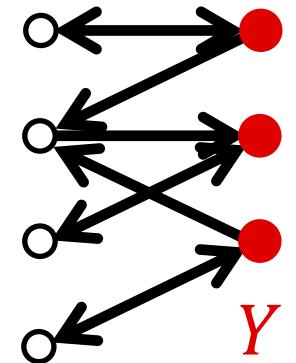
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# “Matroid Intersection is **Tractable**”

Most of them require separate information of two matroids

**Thm.** [Edmonds 1970]

$\mathbf{M}_1, \mathbf{M}_2$ : Matroids on a common ground set  $S$

$$\max\{|Y| \mid Y \in \mathcal{I}_1 \cap \mathcal{I}_2\} = \min\{r_1(Z) + r_2(S \setminus Z) \mid Z \subseteq S\}$$

**Thm.** [Frank 1981]

$\mathbf{M}_1, \mathbf{M}_2$ : Matroids on a common ground set  $S$ ,  $w: S \rightarrow \mathbb{R}$

$$\max \left\{ w(Y) \mid Y \in \mathcal{I}_1^{(k)} \cap \mathcal{I}_2^{(k)} \right\} \quad \left( \mathcal{I}_j^{(k)} := \{Y \in \mathcal{I}_j \mid |Y| = k\} \right)$$

$$= \min \left\{ \max_{Y_1 \in \mathcal{I}_1^{(k)}} w_1(Y_1) + \max_{Y_2 \in \mathcal{I}_2^{(k)}} w_2(Y_2) \mid w_1 + w_2 = w \right\}$$

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# “Matroid Intersection is **Tractable**”

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## LP-relaxation (Primal)

$$\text{maximize} \sum_{e \in S} w(e)x(e)$$

$$\begin{aligned} \text{subject to } & \sum_{e \in Z} x(e) \leq r_1(Z) \quad (Z \subseteq S) \\ & \sum_{e \in Z} x(e) \leq r_2(Z) \quad (Z \subseteq S) \\ & x(e) \geq 0 \quad (e \in S) \end{aligned}$$

Determine the convex hull of the common independent sets  
[Edmonds 1970]

## Dual LP

$$\text{minimize} \sum_{Z \subseteq S} r_1(Z)y_1(Z) + \sum_{Z \subseteq S} r_2(Z)y_2(Z)$$

$$\begin{aligned} \text{subject to } & \sum_{Z \ni e} (y_1(Z) + y_2(Z)) \geq w(e) \quad (e \in S) \\ & y_1(Z) \geq 0 \quad (Z \subseteq S) \\ & y_2(Z) \geq 0 \quad (Z \subseteq S) \end{aligned}$$

- $w$  is integer  $\Rightarrow \exists y_i^*$ : integer, optimal
- $\exists y_i^*$ : optimal s.t.  $\text{supp}(y_i^*)$  is a chain  
 $Z_{i,1} \subsetneq Z_{i,2} \subsetneq \dots \subsetneq Z_{i,k}$

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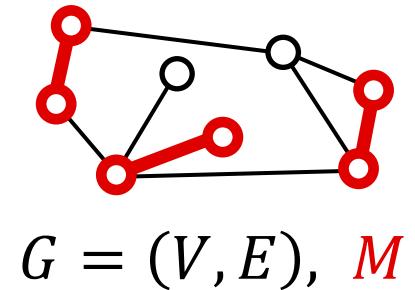
# “Matroid Intersection is Tractable”

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**Matroid Matching** generalizes Matroid Intersection and Matching

**Input:**  $f: 2^S \rightarrow \mathbf{Z}_{\geq 0}$ , 2-polymatroid function (oracle)

**Goal:** maximize  $|Y|$  subject to  $f(Y) = 2|Y|$  and  $Y \subseteq S$



[Matroid Intersection]  $f := r_{\text{sum}} := r_1 + r_2$

[Matching]  $f(F) := |V(F)|$  ( $F \subseteq E =: S$ )

- Matroid matching is hard in general
  - Including NP-hard problems (e.g., Maximum Clique)
  - Instances for which exponentially many oracle calls are necessary  
[Lovász 1981; Jensen–Korte 1982]
- Tractable for linearly represented matroids [Lovász 1980, 1981; ...]  
[Iwata–Kobayashi 2021; Pap 2013]

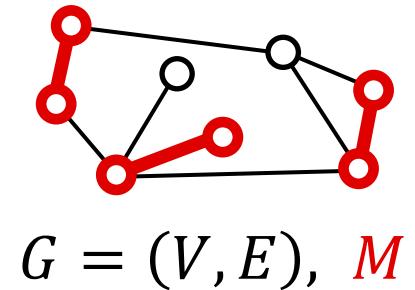
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## What is known

- Efficient Algorithms rely on **separate** oracles for the two matroids
- Max-Min Theorems and the Polyhedral Description are given by using the two rank functions **separately**

## What may be asked

- The resulting combinatorial structure is just  $\mathcal{I}_1 \cap \mathcal{I}_2$
- The polytope is completely determined by  $r_{\min} = \min\{r_1, r_2\}$
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# Matroid Intersection under Restricted Oracles

## Question

Is matroid intersection tractable if we only get the following information?

For each subset  $X \subseteq S$ ,

[CI] whether  $X \in \mathcal{I}_1 \cap \mathcal{I}_2$  or not,

[MIN]  $r_{\min}(X) = \min\{r_1(X), r_2(X)\}$ ,

[SUM]  $r_{\sum}(X) = r_1(X) + r_2(X)$ , or

[MAX]  $r_{\max}(X) = \max\{r_1(X), r_2(X)\}$ .

**Obs.** MAX is too weak as it gives no information on the second matroid if the first matroid is free, i.e.,  $r_1(X) = |X|$  ( $\forall X \subseteq S$ ).

# Results and Open Problems

## What we know (Results)

- Relation between Restricted Oracles
- SUM and CI+MAX can solve Weighted in general
- MIN can solve Unweighted in general, and Weighted in some cases
- CI can solve Unweighted/Weighted in some cases

## What we want to know (Open)

- Can MIN solve Weighted in general? Or, is it hard?
- Can CI solve Unweighted/Weighted in general? Or, is it hard?

[CI] whether  $X \in \mathcal{I}_1 \cap \mathcal{I}_2$  or not

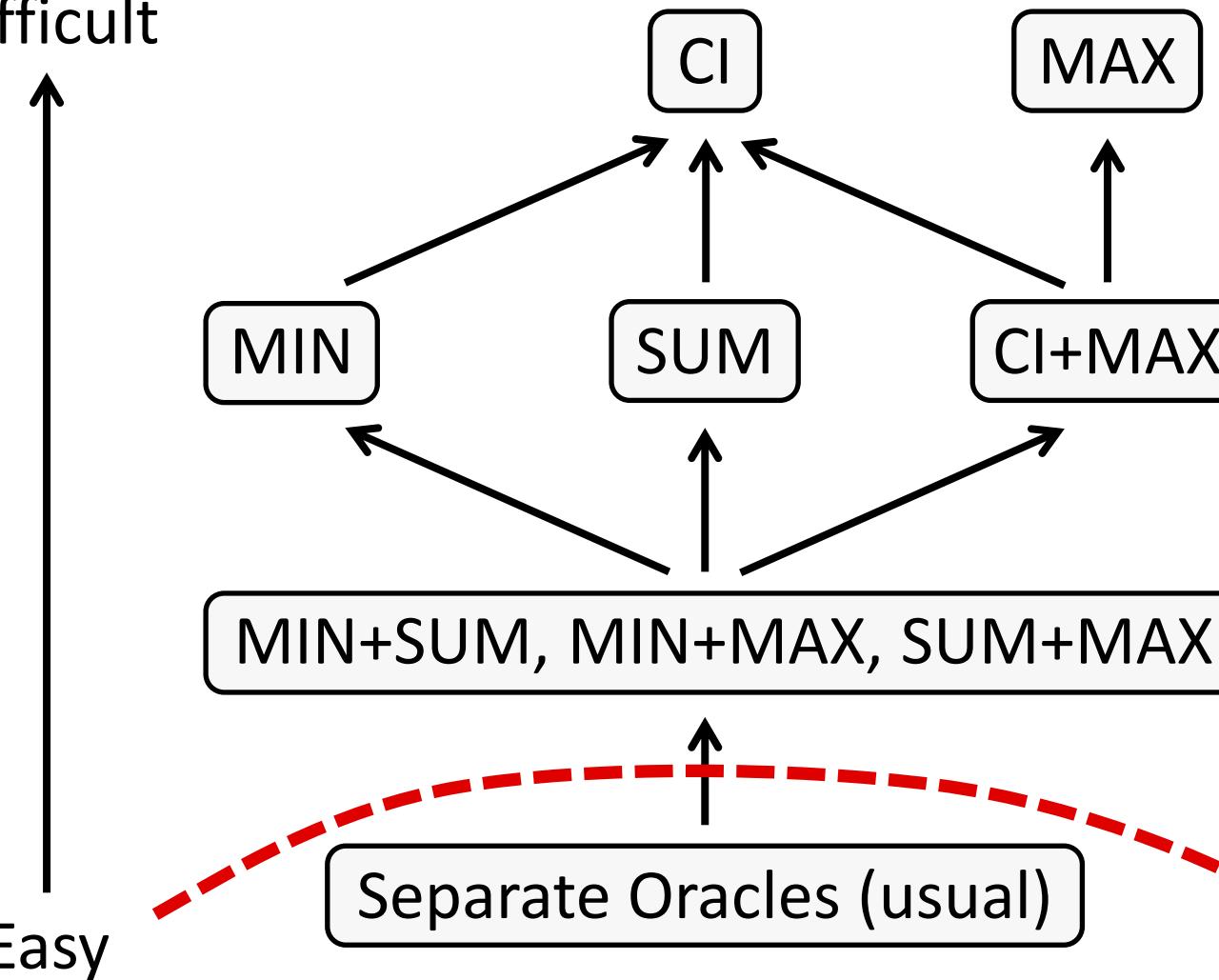
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# [Result 1] Relation between Restricted Oracles

Difficult



[CI] whether  $X \in \mathcal{I}_1 \cap \mathcal{I}_2$  or not

[MIN]  $r_{\min}(X) = \min\{r_1(X), r_2(X)\}$

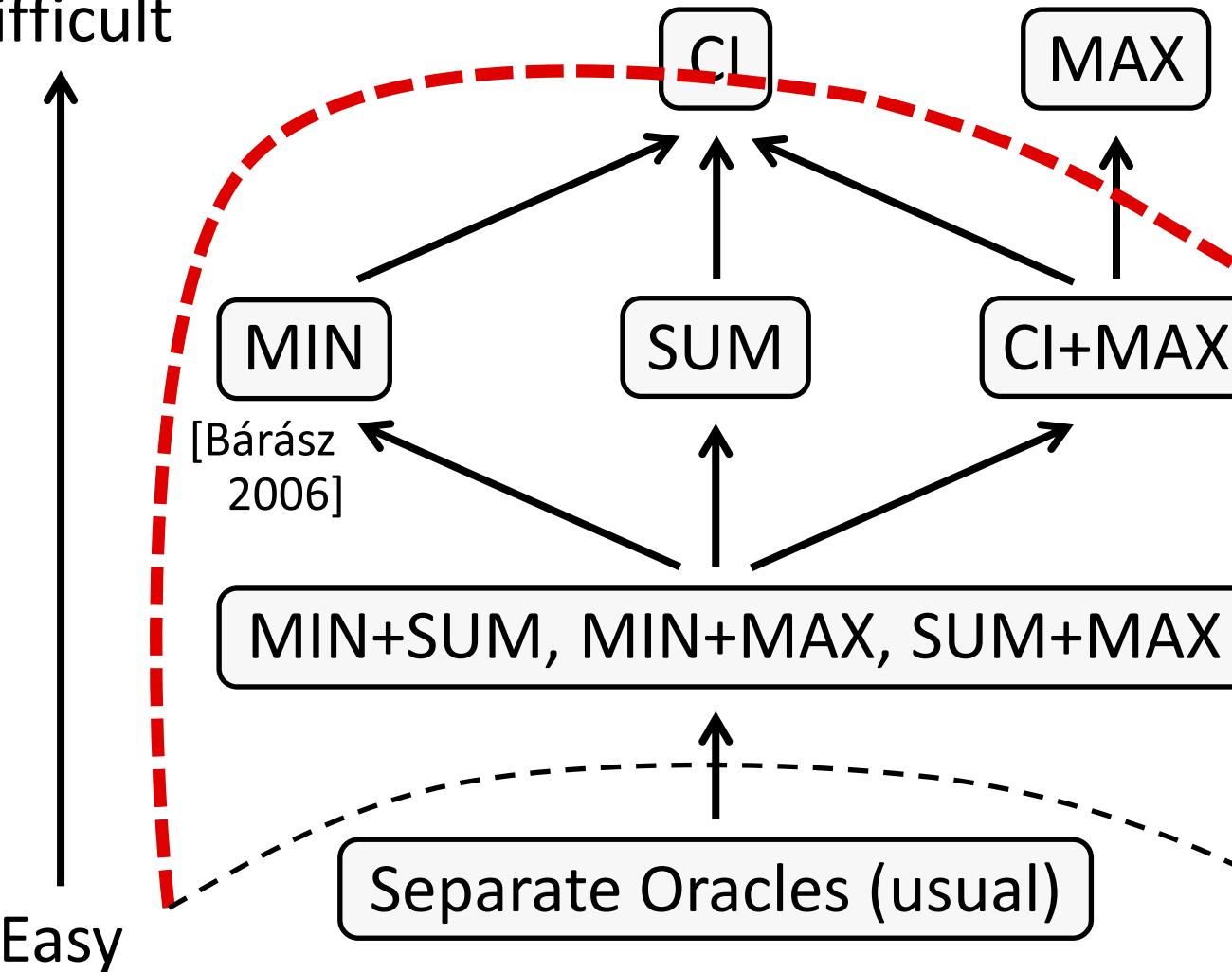
[SUM]  $r_{\sum}(X) = r_1(X) + r_2(X)$

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- A  $\rightsquigarrow$  B (**reachable**) means the oracle B is always emulated by using the oracle A
- A  $\not\rightsquigarrow$  B (**unreachable**) means  $\exists$  matroid intersection instances s.t. B can distinguish them but A cannot

# [Result 2] Unweighted Matroid Intersection

Difficult



[CI] whether  $X \in \mathcal{I}_1 \cap \mathcal{I}_2$  or not

[MIN]  $r_{\min}(X) = \min\{r_1(X), r_2(X)\}$

[SUM]  $r_{\sum}(X) = r_1(X) + r_2(X)$

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CI is tractable if one matroid is

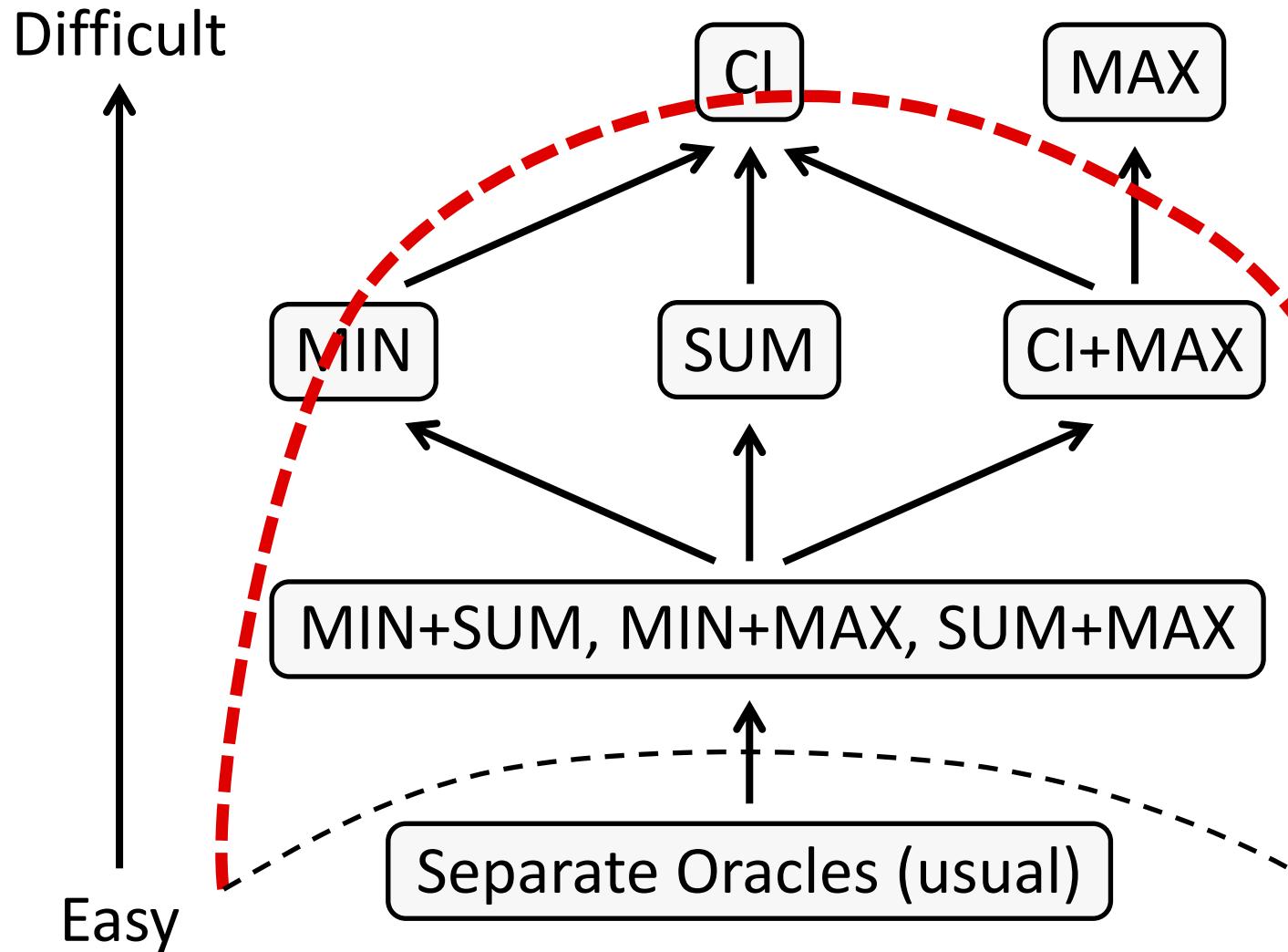
- a partition matroid, or
- an elementary split matroid.

Generalization of  
paving matroid

[Joswig–Schröter 2017; BKSYY 2023]

Tractable

# [Result 3] Weighted Matroid Intersection



[CI] whether  $X \in \mathcal{I}_1 \cap \mathcal{I}_2$  or not

$$[\text{MIN}] \quad r_{\min}(X) = \min\{r_1(X), r_2(X)\}$$

$$[\text{SUM}] \quad r_{\text{sum}}(X) = r_1(X) + r_2(X)$$

$$[\text{MAX}] \quad r_{\max}(X) = \max\{r_1(X), r_2(X)\}$$

MIN is tractable if

- every circuit in one matroid is small (FPT), or
  - no pair of circuits s.t. one includes the other.

CI is tractable if one matroid is an elementary split matroid.

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# Matroid (Notation)

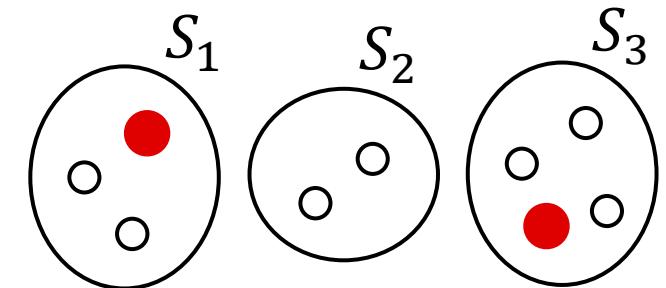
**M**: Matroid on a ground set  $S$

- $\mathcal{I} \subseteq 2^S$ : **Independent set** family
- $\mathcal{B} \subseteq 2^S$ : **Base (Basis)** family
- $\mathcal{C} \subseteq 2^S$ : **Circuit** family
- $r: 2^S \rightarrow \mathbf{Z}_{\geq 0}$ , **Rank** function;  $r(X) := \max \{ |Y| \mid Y \subseteq X, Y \in \mathcal{I} \}$
- $\text{cl}: 2^S \rightarrow 2^S$ , **Closure** operator;  $\text{cl}(X) := \{ e \in S \mid r(X \cup \{e\}) = r(X) \}$

# Matroid (Examples)

- Partition Matroid

$$S = S_1 \cup S_2 \cup \dots \cup S_k, \mathcal{I} = \{ Y \subseteq S \mid |Y \cap S_i| \leq 1 (\forall i \in [k]) \}$$



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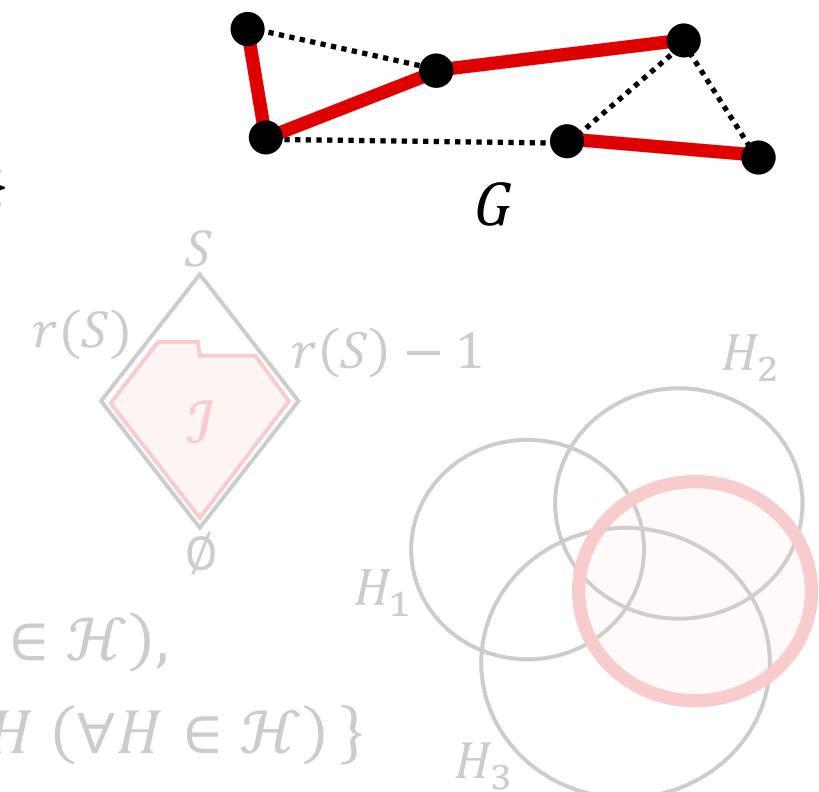
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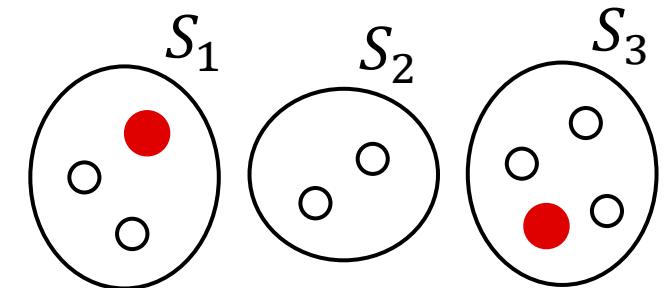


↓ Generalized  
Elementary Split Matroid  
[BKSYY 2023]

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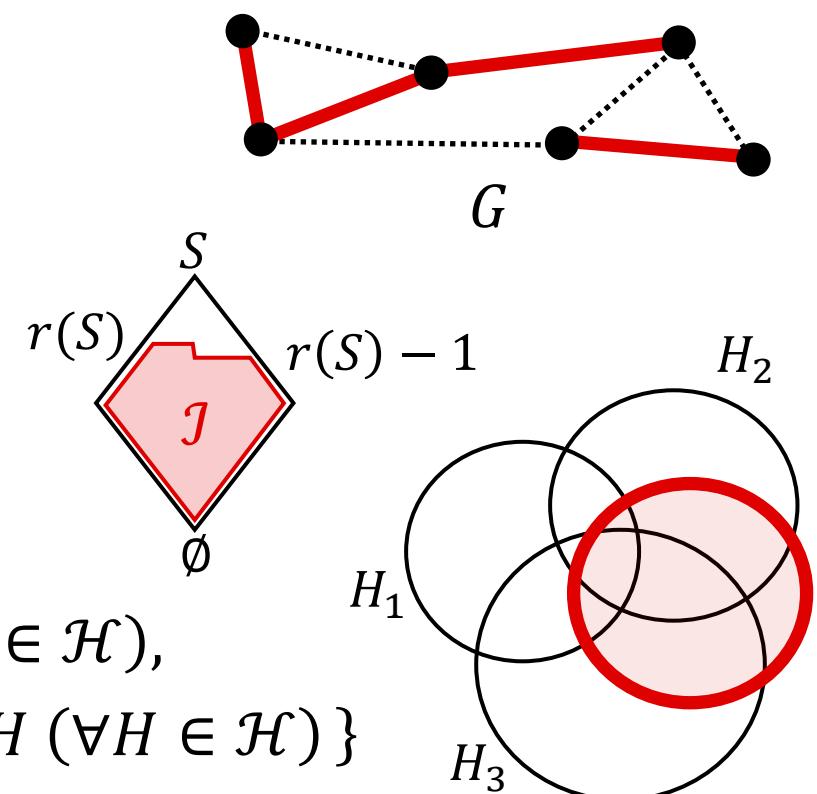
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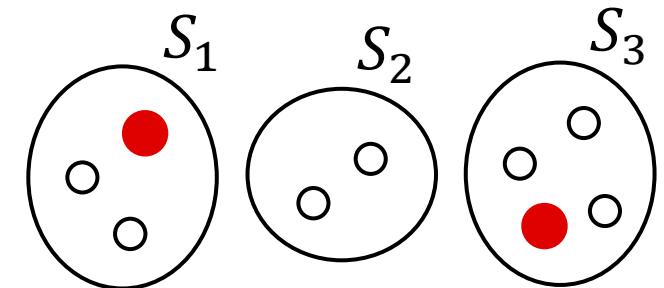
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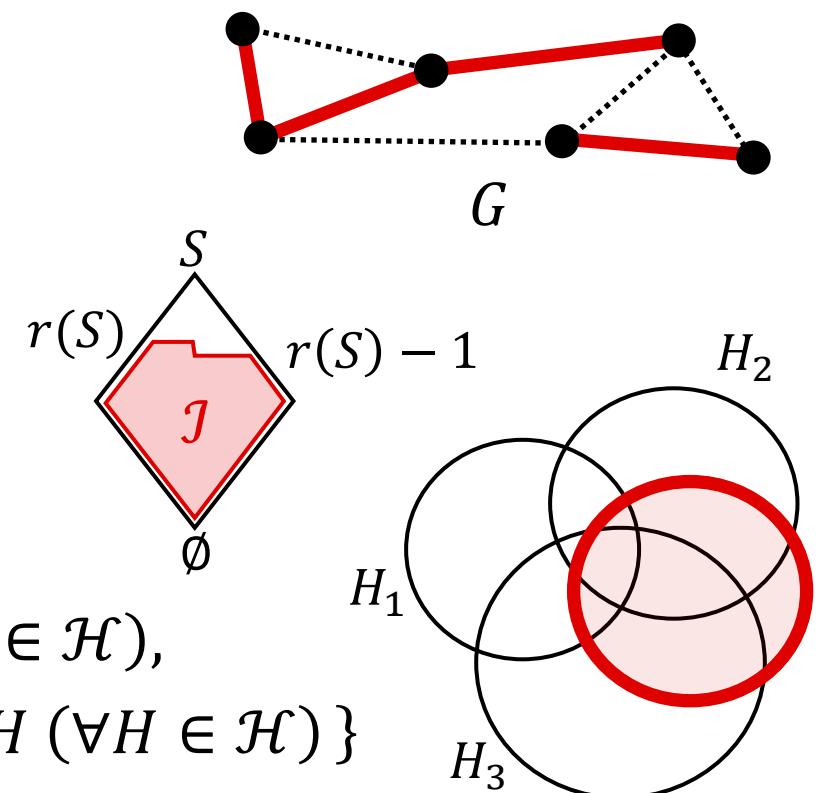
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# Matroid Intersection Problem (Unweighted)

**Input:**  $S$ : Finite set,  $\mathbf{M}_1, \mathbf{M}_2$ : Matroids on  $S$  (**oracle**)

**Goal:** maximize  $|Y|$  subject to  $Y \in \mathcal{I}_1 \cap \mathcal{I}_2$

- Usually, **separate oracles** are given, i.e., we can ask for each subset  $X \subseteq S$  and each  $i = 1, 2$ , whether  $X \in \mathcal{I}_i$  or not, the rank  $r_i(X)$ , etc.
- Many Applications (Special Cases)
  - Bipartite matching: Partition + Partition
  - Arborescence (packing): Partition + Graphic (unions)
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# Matroid Intersection Problem (Weighted)

**Input:**  $S$ : Finite set,  $\mathbf{M}_1, \mathbf{M}_2$ : Matroids on  $S$  (**oracle**),  $w: S \rightarrow \mathbf{R}$

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- The goal of this study is to clarify what happens if **the oracle is restricted**:
  - [CI] whether  $X \in \mathcal{I}_1 \cap \mathcal{I}_2$  or not,
  - [MIN]  $r_{\min}(X) = \min\{r_1(X), r_2(X)\}$ , or
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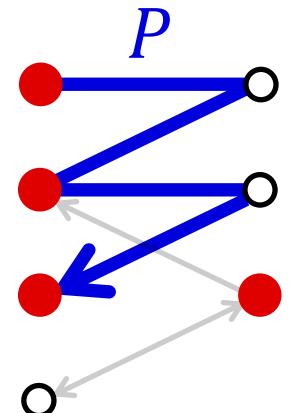
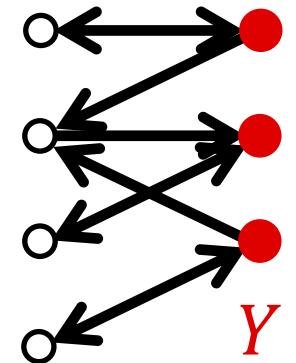
# “Matroid Intersection is **Tractable**”

Most of them require separate information of two matroids

## Basic Strategy of Efficient Algorithms

- Starting with  $Y = \emptyset$ , repeatedly update the current solution  $Y$ .
- For each update,
  - construct the exchangeability graph w.r.t.  $Y$ ,
  - find an augmenting path  $P$  in the graph, and
  - flip the current solution along the path, i.e.,  $Y \leftarrow Y \Delta P$ .
- The edges in the graph are oriented according to **in which matroid** the two elements are exchangeable.

**Assumption: Independence in each matroid** can be tested.

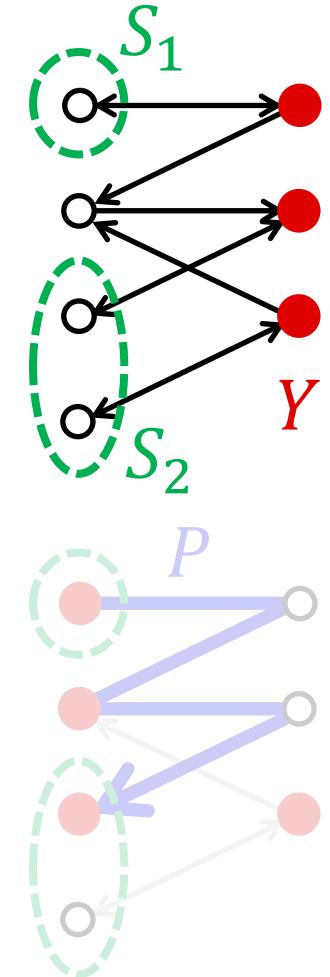
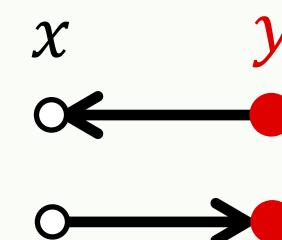


# Exchangeability Graph

**Def.**  $\mathbf{M}_1, \mathbf{M}_2$ : Matroids on a common ground set  $S$ ,  $Y \in \mathcal{I}_1 \cap \mathcal{I}_2$   
 $D[Y] = (S \setminus Y, Y; A[Y])$ : **Exchangeability Graph** w.r.t.  $Y$



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- $S_1 := \{x \mid Y + x \in \mathcal{I}_1\}$  (Sources)
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# Augmentability (Unweighted)

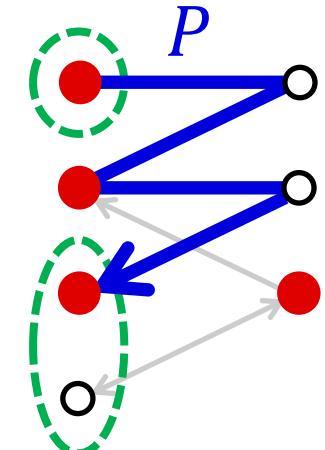
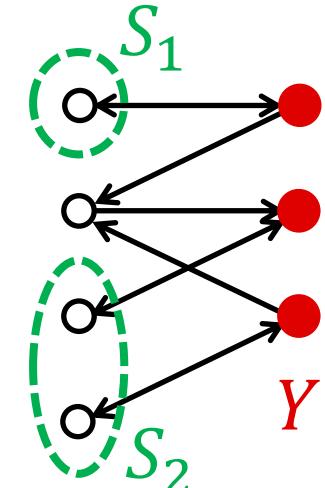
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- If  $D[Y]$  has no  $S_1$ – $S_2$  path, then  $|Y|$  is maximum.
- If  $P$  is a **shortest**  $S_1$ – $S_2$  path in  $D[Y]$ , then  $Y \triangle P \in \mathcal{I}_1 \cap \mathcal{I}_2$ .

$O(nr^2)$  time in total, where  $r := \text{opt. value} \leq n$

- $D[Y]$  is constructed by  $O(nr)$  oracle calls
- $P$  is found by BFS in linear time ( $n$  vertices,  $O(nr)$  edges)
- #(iteration) is  $r + 1$



# Augmentability (Weighted)

**Thm.**  $\mathbf{M}_1, \mathbf{M}_2$ : Matroids on a common ground set  $S$

$$Y \in \operatorname{argmax} \left\{ w(X) \mid X \in \mathcal{I}_1^{(k)} \cap \mathcal{I}_2^{(k)} \right\} \quad (k = |Y|)$$

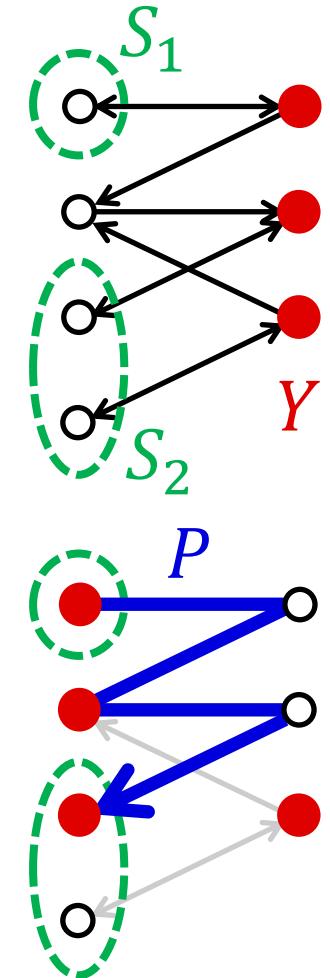
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$$\text{cost}(P) := w(P \cap Y) - w(P \setminus Y) \quad (P: \text{path/cycle})$$

- $D[Y]$  has **no negative-cost cycle**.
- If  $P$  is a **shortest cheapest**  $S_1$ – $S_2$  path in  $D[Y]$ ,

$$\text{then } Y \Delta P \in \operatorname{argmax} \left\{ w(X) \mid X \in \mathcal{I}_1^{(k+1)} \cap \mathcal{I}_2^{(k+1)} \right\}.$$

$O(n^2r^2)$  time in total (Bellman–Ford, Weight Splitting, etc.)



# Outline

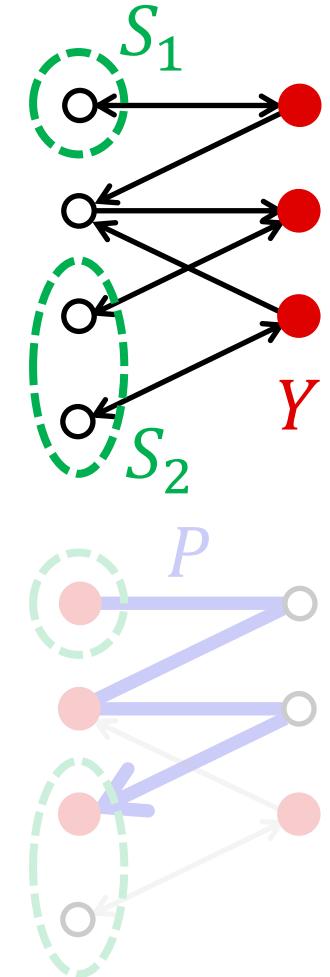
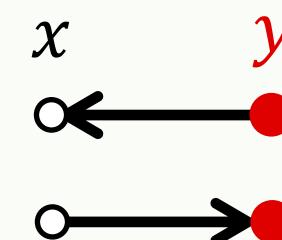
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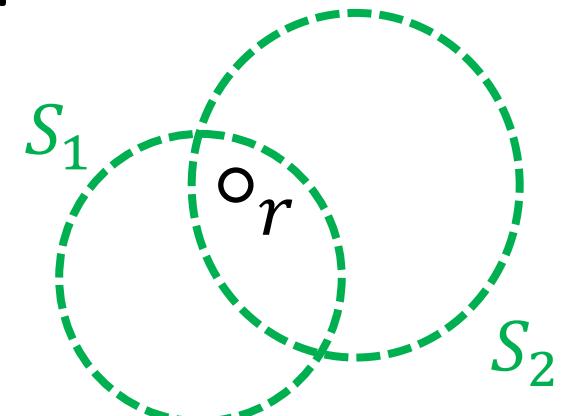
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This can be recognized by CI, and hence by MIN or SUM.

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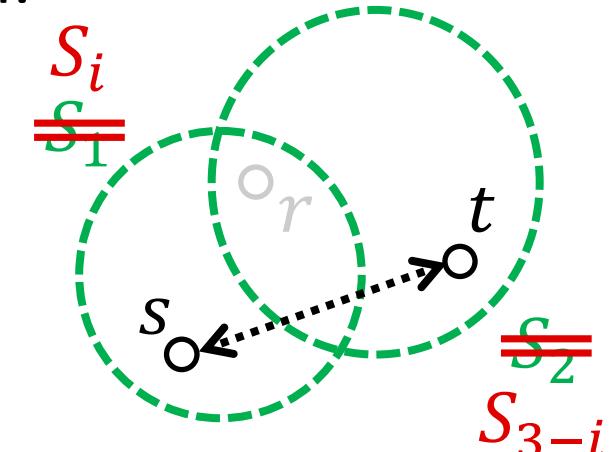
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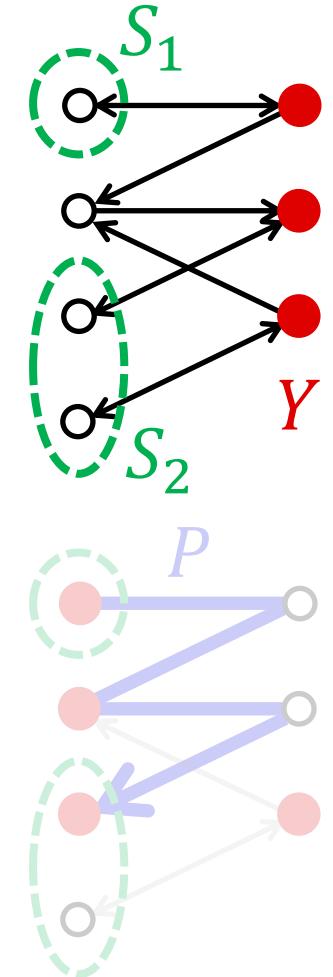
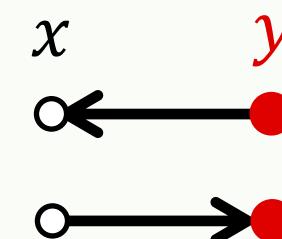
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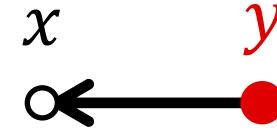


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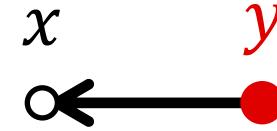


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Emulate a usual algorithm on the **Overestimated** Exchangeability Graph

- Sources, Sinks, and Edges around them are correctly recognized (up to sym.).
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if an edge  $e$  is wrongly estimated, there is a correct **shortcut** skipping  $e$ .  
→ **None of such fake edges is used in a shortest path!**

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Assume  $S_1 \cap S_2 = \emptyset$ ,  $s \in S_1$ ,  $t \in S_2$

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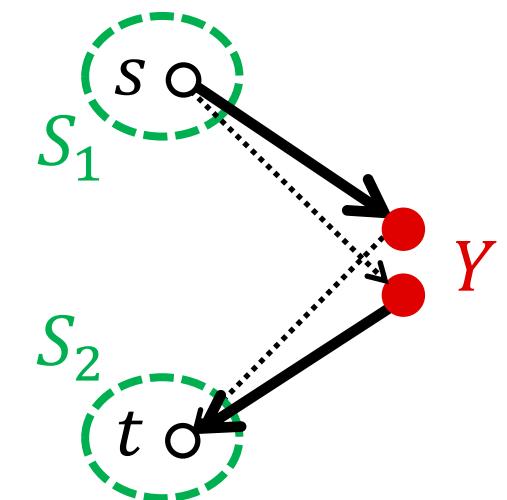
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**Enough to find a shortest path! (Wrong  $\Rightarrow \exists$  Shortcut)**

$$\begin{aligned} S_i &= \{x \mid Y + x \in \mathcal{J}_i\} \quad (i = 1, 2) \\ A_1[Y] &= \{(y, x) \mid Y - y + x \in \mathcal{J}_1\} \\ A_2[Y] &= \{(x, y) \mid Y - y + x \in \mathcal{J}_2\} \end{aligned}$$



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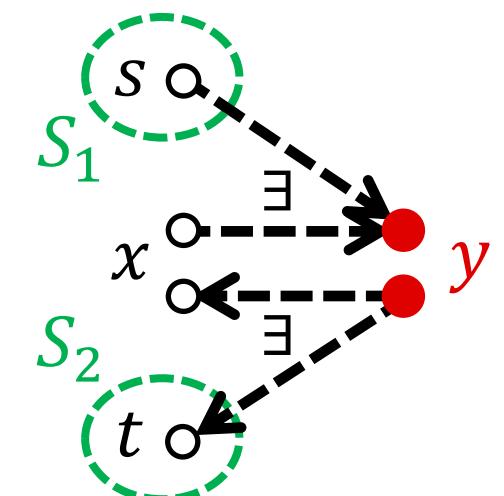
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# Overestimation of $A[Y]$ via MIN (Unweighted)

Assume  $S_1 \cap S_2 = \emptyset$ ,  $s \in S_1$ ,  $t \in S_2$

- $Y + s \in \mathcal{J}_1 \setminus \mathcal{J}_2$ ,  $Y + t \in \mathcal{J}_2 \setminus \mathcal{J}_1$

- $\forall y \in Y$ ,

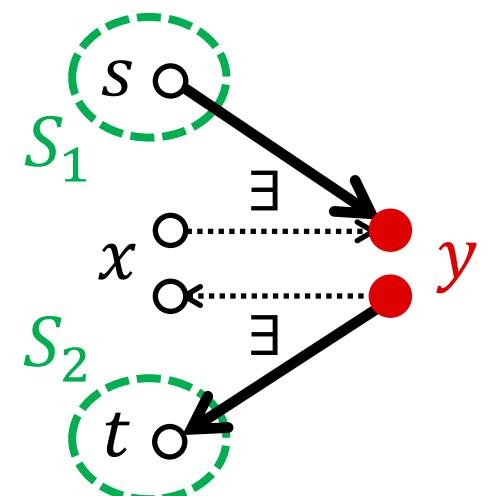
- $(y, t) \in A_1[Y] \Leftrightarrow r_1(Y - y + t) = |Y| \Leftrightarrow r_{\min}(Y - y + t) = |Y|$
- $(s, y) \in A_2[Y] \Leftrightarrow r_2(Y - y + s) = |Y| \Leftrightarrow r_{\min}(Y - y + s) = |Y|$

- Suppose that  $\forall x \in S \setminus (Y \cup S_1 \cup S_2)$ ,  $\forall y \in Y$ ,

- Estimate  $\exists(y, x) \Leftrightarrow (y, x) \in A_1[Y]$  or  $(y, t) \in A_1[Y]$
- Estimate  $\exists(x, y) \Leftrightarrow (x, y) \in A_2[Y]$  or  $(s, y) \in A_2[Y]$

**Enough to find a shortest path! (Wrong  $\Rightarrow \exists$  Shortcut)**

$$\begin{aligned} S_i &= \{x \mid Y + x \in \mathcal{J}_i\} \quad (i = 1, 2) \\ A_1[Y] &= \{(y, x) \mid Y - y + x \in \mathcal{J}_1\} \\ A_2[Y] &= \{(x, y) \mid Y - y + x \in \mathcal{J}_2\} \end{aligned}$$



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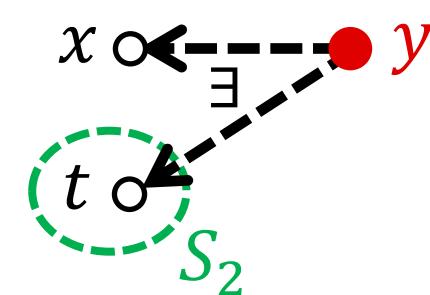
Assume  $S_1 \cap S_2 = \emptyset$ ,  $s \in S_1$ ,  $t \in S_2$

- $Y + s \in \mathcal{J}_1 \setminus \mathcal{J}_2$ ,  $Y + t \in \mathcal{J}_2 \setminus \mathcal{J}_1$
- Estimate  $\exists(y, x) (x \in S \setminus (Y \cup S_1 \cup S_2), y \in Y)$   
 $\Leftrightarrow (y, x) \in A_1[Y] \text{ or } (y, t) \in A_1[Y]$   
 $\Leftrightarrow r_1(Y - y + x) = |Y| \text{ or } r_1(Y - y + t) = |Y|$

$$S_i = \{x \mid Y + x \in \mathcal{J}_i\} \quad (i = 1, 2)$$

$$A_1[Y] = \{(y, x) \mid Y - y + x \in \mathcal{J}_1\}$$

$$A_2[Y] = \{(x, y) \mid Y - y + x \in \mathcal{J}_2\}$$



# Overestimation of $A[Y]$ via MIN (Unweighted)

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- Estimate  $\exists(y, x) (x \in S \setminus (Y \cup S_1 \cup S_2), y \in Y)$

$$\Leftrightarrow (y, x) \in A_1[Y] \text{ or } (y, t) \in A_1[Y]$$

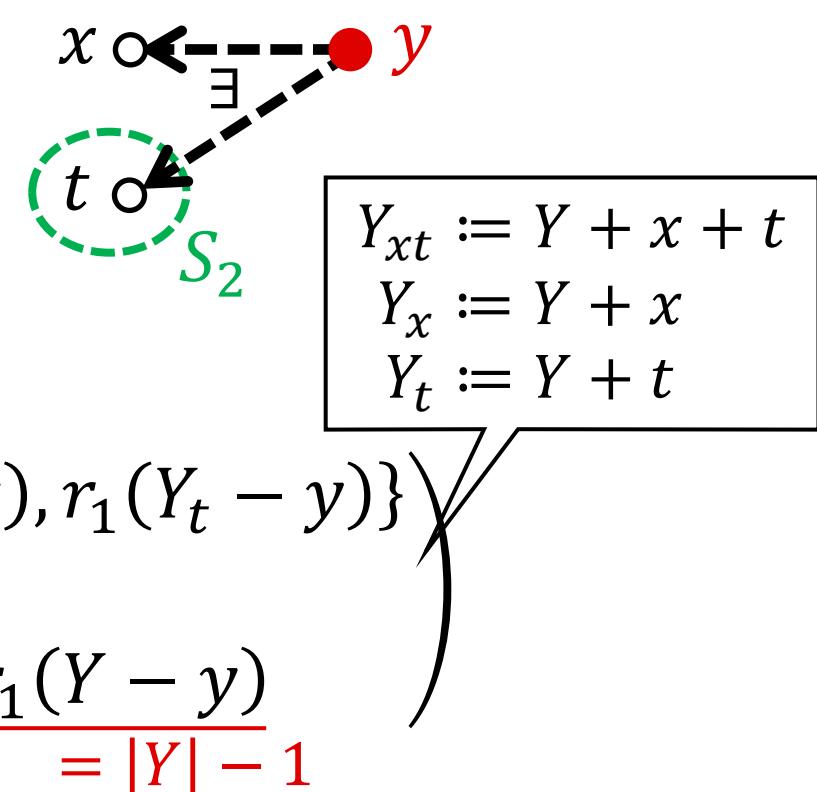
$$\Leftrightarrow r_1(Y - y + x) = |Y| \text{ or } r_1(Y - y + t) = |Y|$$

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$$\left( \frac{|Y| = r_1(Y_{xt}) \geq r_1(Y_{xt} - y) \geq \max\{r_1(Y_x - y), r_1(Y_t - y)\}}{x, t \in S \setminus S_1 = \text{cl}_1(Y)} \right)$$

$$r_1(Y_x - y) + r_1(Y_t - y) \geq r_1(Y_{xt} - y) + \underline{r_1(Y - y)} \\ = |Y| - 1$$

$$\begin{aligned} S_i &= \{x \mid Y + x \in \mathcal{J}_i\} \quad (i = 1, 2) \\ A_1[Y] &= \{(y, x) \mid Y - y + x \in \mathcal{J}_1\} \\ A_2[Y] &= \{(x, y) \mid Y - y + x \in \mathcal{J}_2\} \end{aligned}$$

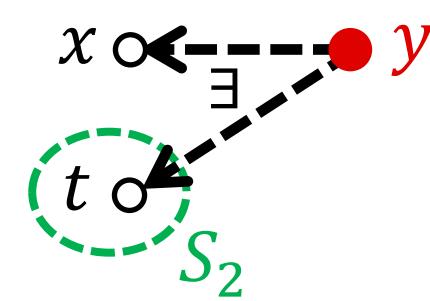


# Overestimation of $A[Y]$ via MIN (Unweighted)

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  - $\Leftrightarrow (y, x) \in A_1[Y]$  or  $(y, t) \in A_1[Y]$
  - $\Leftrightarrow r_1(Y - y + x) = |Y|$  or  $r_1(Y - y + t) = |Y|$
  - $\Leftrightarrow r_1(Y - y + x + t) = |Y|$
  - $\Leftrightarrow r_{\min}(Y - y + x + t) = |Y|$
  - $(r_2(Y - y + x + t) \geq r_2(Y + t) - 1 = |Y|)$

$$\begin{aligned} S_i &= \{x \mid Y + x \in \mathcal{J}_i\} \quad (i = 1, 2) \\ A_1[Y] &= \{(y, x) \mid Y - y + x \in \mathcal{J}_1\} \\ A_2[Y] &= \{(x, y) \mid Y - y + x \in \mathcal{J}_2\} \end{aligned}$$

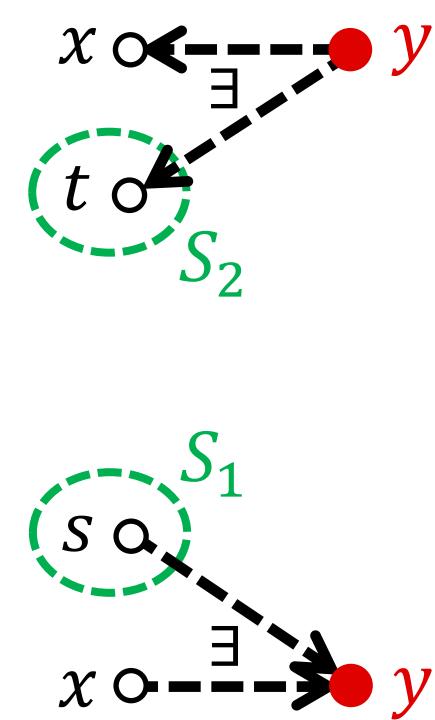


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- Estimate  $\exists(x, y) \Leftrightarrow r_{\min}(Y - y + x + s) = |Y|$

$S_i = \{x \mid Y + x \in \mathcal{J}_i\} (i = 1, 2)$
$A_1[Y] = \{(y, x) \mid Y - y + x \in \mathcal{J}_1\}$
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# Unweighted Matroid Intersection via MIN [Bárász 2006]

Emulate a usual algorithm on the Overestimated Exchangeability Graph

- Sources, Sinks, and Edges around them are correctly recognized (up to sym.).
- Other edges are overestimated so that  
if an edge  $e$  is wrongly estimated, there is a correct **shortcut** skipping  $e$ .  
→ **None of such fake edges is used in a shortest path!**

Thm.  $Y \in \mathcal{I}_1 \cap \mathcal{I}_2$ ,  $D[Y]$ : Exchangeability Graph w.r.t.  $Y$

- If  $D[Y]$  has no  $S_1$ – $S_2$  path, then  $|Y|$  is maximum.
- If  $P$  is a **shortest**  $S_1$ – $S_2$  path in  $D[Y]$ , then  $Y \Delta P \in \mathcal{I}_1 \cap \mathcal{I}_2$ .

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if an edge  $e$  is wrongly estimated, there is a correct **shortcut** skipping  $e$ .  
→ **None of such fake edges is used in a shortest path!**
- When the algorithm halts, a dual optimal solution is found by reachability.

**Thm.**  $\max\{ |Y| \mid Y \in \mathcal{I}_1 \cap \mathcal{I}_2 \} = \min\{ r_{\min}(Z) + r_{\min}(S \setminus Z) \mid Z \subseteq S \}$

$$|Y| = \frac{r_{\min}(Y \cap Z)}{|Y \cap Z|} + \frac{r_{\min}(Y \setminus Z)}{|Y \setminus Z|} \leq r_{\min}(Z) + r_{\min}(S \setminus Z) \leq r_1(Z) + r_2(S \setminus Z)$$

# Weighted Matroid Intersection via MIN [BBKYY 2023+]

Try to emulate usual algorithms on the Overestimated Exchangeability Graph

- Sources, Sinks, and Edges around them are correctly recognized (up to sym.).
- Other edges are overestimated so that
  - if an edge  $e$  is wrongly estimated, there is a correct shortcut skipping  $e$ .
    - Such **fake edges may be used** in a (shortest) cheapest path!
    - They **may cause negative-cost cycles!** (NP-hard!?)

Thm.  $Y \in \mathcal{I}_1^{(k)} \cap \mathcal{I}_2^{(k)}$ : Max-Weight,  $\text{cost}(P) := w(P \cap Y) - w(P \setminus Y)$  ( $P$ : path/cycle)

- $D[Y]$  has no negative-cost cycle.
- If  $P$  is a shortest cheapest  $S_1 - S_2$  path in  $D[Y]$ , then  $Y \Delta P \in \mathcal{I}_1^{(k+1)} \cap \mathcal{I}_2^{(k+1)}$  is max-weight.

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- Finding a consistent graph is **NP-hard**... (4-coloring of 3-colorable graphs)

# Extra Info. from Two-by-Two Exchange

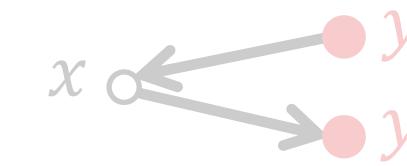
[BBKYY 2023+]

**Extra information** from **at most two-by-two exchanges** may refine the graph!

- $r_{\min}(Y - y + x) = |Y| \iff x \circ \longleftrightarrow y$

- Otherwise,

- $r_{\min}(Y - y_1 - y_2 + x) = |Y| - 1 \iff$

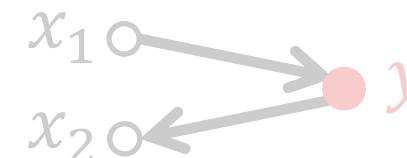


or

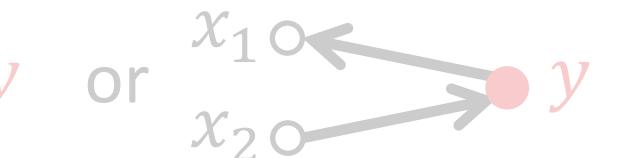


- $r_{\min}(Y - y + x_1 + x_2) = |Y| \iff$

$\iff$

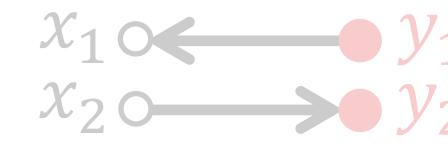


or



- Otherwise (none of the above holds),

$$r_{\min}(Y - y_1 - y_2 + x_1 + x_2) = |Y| - 1 \iff$$



or



or



or



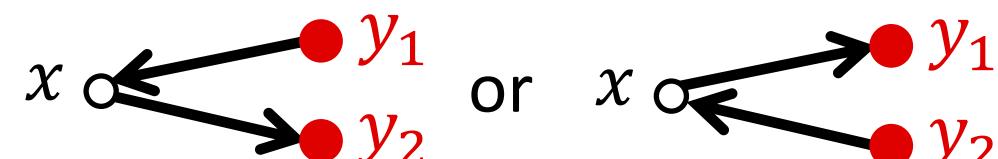
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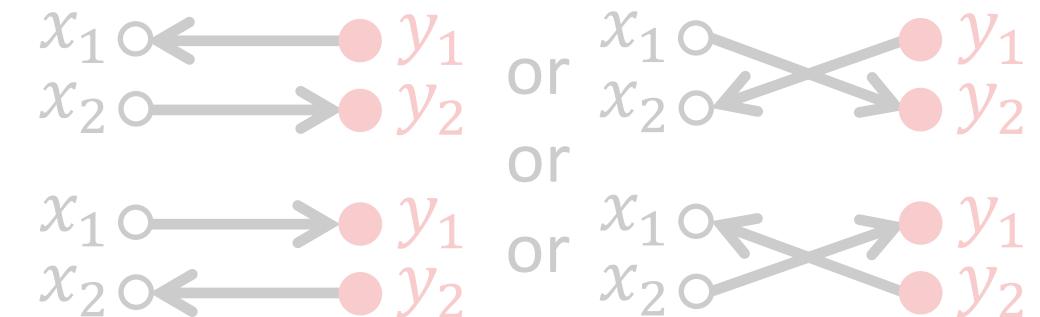
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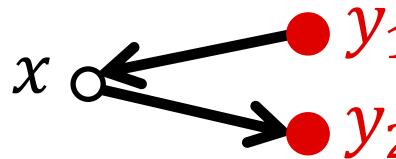
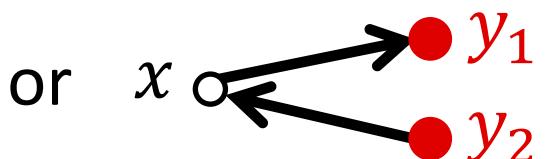
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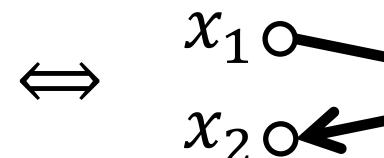
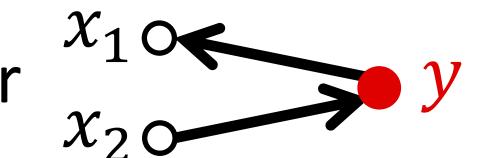
[BBKYY 2023+]

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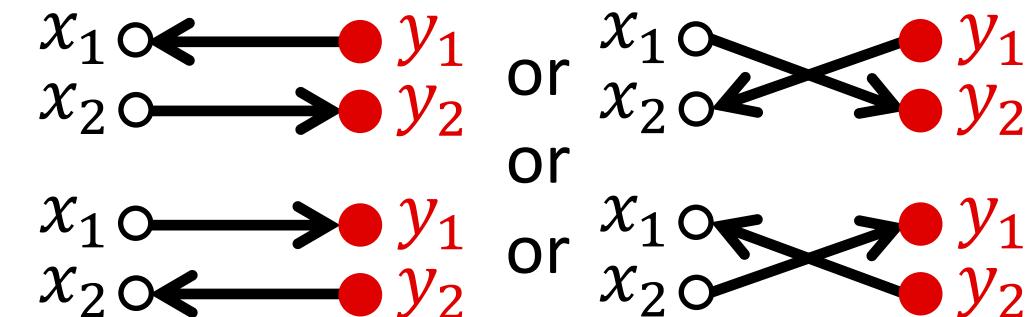
- Otherwise,

- $r_{\min}(Y - y_1 - y_2 + x) = |Y| - 1 \iff$   or 

- $r_{\min}(Y - y + x_1 + x_2) = |Y| \iff$   or 

- Otherwise (none of the above holds),

$$r_{\min}(Y - y_1 - y_2 + x_1 + x_2) = |Y| - 1 \iff$$



# Weighted Matroid Intersection via MIN [BBKYY 2023+]

Try to emulate usual algorithms on the Overestimated Exchangeability Graph

- Sources, Sinks, and Edges around them are correctly recognized (up to sym.).
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- **Any graph consistent** with all the extra information **is suitable** for emulation!!
- Finding a consistent graph is **NP-hard**... (4-coloring of 3-colorable graphs)

# Consistency with Extra Info. is Enough

[BBKYY 2023+]

**Any graph consistent** with all the extra information **is suitable** for emulation!!

**Thm.**  $Y \in \operatorname{argmax} \left\{ w(X) \mid X \in \mathcal{I}_1^{(k)} \cap \mathcal{I}_2^{(k)} \right\}$  ( $k = |Y|$ )

$D[Y]$ : Exchangeability Graph w.r.t.  $Y$

$\tilde{D}[Y]$ : Subgraph of the overestimation **consistent with all the extra info.**

$\text{cost}(P) := w(P \cap Y) - w(P \setminus Y)$  ( $P$ : path/cycle)

- $\tilde{D}[Y]$  has **no negative-cost cycle**.
- $\forall \tilde{P}$ : **shortest cheapest**  $S_1 - S_2$  path in  $\tilde{D}[Y]$ ,
- $\exists P$ : **shortest cheapest**  $S_1 - S_2$  path in  $D[Y]$  with the same vertex set, and vice versa.

# Weighted Matroid Intersection via MIN [BBKYY 2023+]

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# Tractable cases of WMI via MIN [BBKYY 2023+]

- When  $\forall C_1 \in \mathcal{C}_1, \forall C_2 \in \mathcal{C}_2, C_1 \not\subseteq C_2$  and  $C_2 \not\subseteq C_1$   
→ Finding a consistent graph is reduced to **2-SAT**
- When  $\exists i \in \{1, 2\}, \forall C \in \mathcal{C}_i, |C| \leq k$   
→  $O(2^k \cdot \text{poly}(n))$  time by 2-SAT + Brute-Force Guess
- Lexicographical Maximization
  - Max. #(heaviest); Sub. to this, Max. #(second heaviest); and so on
  - Update **with preserving the numbers of heavier elements** can be done via **Underestimation** of the Exchangeability Graph (by 2-SAT)
  - **Approximation with factor 2 or better** for the original problem [BKYY 2022]

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**Extra information from at most two-by-two exchanges may refine the graph!**

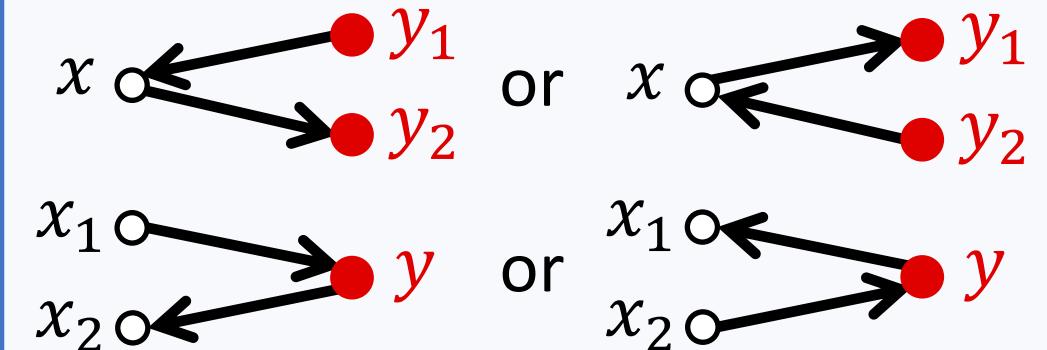
- $r_{\min}(Y - y + x) = |Y| \iff x \circ \longleftrightarrow y$

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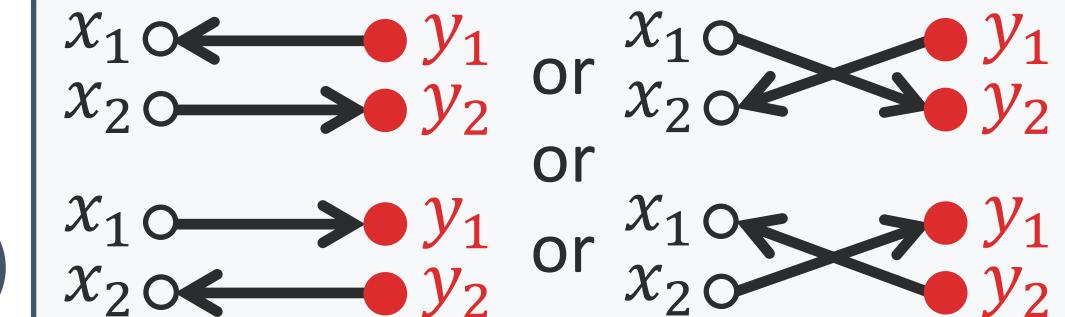
**2-SAT works! (Easy)**



- Otherwise (none of the above holds),

$$r_{\min}(Y - y_1 - y_2 + x_1 + x_2) = |Y| - 1 \iff$$

**Can represent 4-coloring (Hard)**



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# Summary

## Question

Is matroid intersection tractable if we only get the following information?

For each subset  $X \subseteq S$ ,

[CI] whether  $X \in \mathcal{I}_1 \cap \mathcal{I}_2$  or not,

[MIN]  $r_{\min}(X) = \min\{r_1(X), r_2(X)\}$ ,

[SUM]  $r_{\sum}(X) = r_1(X) + r_2(X)$ , or

[MAX]  $r_{\max}(X) = \max\{r_1(X), r_2(X)\}$ .

**Obs.** MAX is too weak as it gives no information on the second matroid if the first matroid is free, i.e.,  $r_1(X) = |X|$  ( $\forall X \subseteq S$ ).

# Summary

## What we know (Results)

- Relation between Restricted Oracles
- SUM and CI+MAX can solve Weighted in general (Emulate Bellman–Ford)
- MIN can solve Unweighted in general, and Weighted in some cases
  - No circuit inclusion (2-SAT)
  - All circuits are small in one matroid (2-SAT + Brute-Force Guess, FPT)
  - Lexicographical Maximization (2-SAT, 2- or better Approximation in general)
- CI can solve Unweighted/Weighted in some cases
  - One is a partition matroid, Unweighted (Emulate BFS)
  - One is an elementary split matroid, Weighted (Brute-Force)

[CI] whether  $X \in \mathcal{I}_1 \cap \mathcal{I}_2$  or not

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[MIN]  $r_{\min}(X) = \min\{r_1(X), r_2(X)\}$

[SUM]  $r_{\text{sum}}(X) = r_1(X) + r_2(X)$

[MAX]  $r_{\max}(X) = \max\{r_1(X), r_2(X)\}$

## What we want to know (Open)

- Can MIN solve Weighted in general? Or, is it hard?
- Can CI solve Unweighted/Weighted in general? Or, is it hard?

