

# A Fast Approximation for Maximum Weight Matroid Intersection

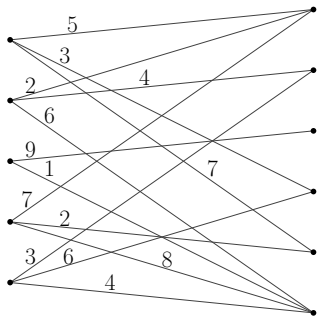
Chandra Chekuri    *Kent Quanrud*

University of Illinois at Urbana-Champaign

January 10, 2016

# Max. weight bipartite matching

**Input:**



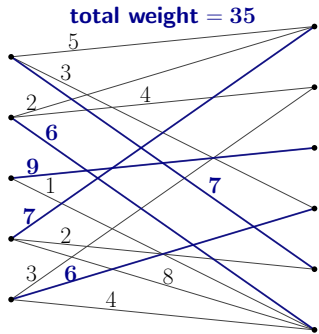
bipartite graph  $G = (U \sqcup V, E)$   
weights  $w : E \rightarrow \mathbb{R}_{\geq 0}$

**Output:**

matching  $M \subseteq E$  maximizing

$$w(M) \stackrel{\text{def}}{=} \sum_{e \in M} w(e)$$

# Max. weight bipartite matching



**Input:**

bipartite graph  $G = (U \sqcup V, E)$

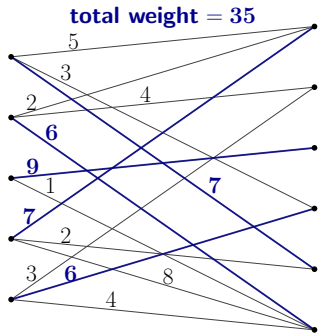
weights  $w : E \rightarrow \mathbb{R}_{\geq 0}$

**Output:**

matching  $M \subseteq E$  maximizing

$$w(M) \stackrel{\text{def}}{=} \sum_{e \in M} w(e)$$

# Max. weight bipartite matching



**Input:**

bipartite graph  $G = (U \sqcup V, E)$

weights  $w : E \rightarrow \mathbb{R}_{\geq 0}$

**Output:**

matching  $M \subseteq E$  maximizing

$$w(M) \stackrel{\text{def}}{=} \sum_{e \in M} w(e)$$

**Approximate output:**

matching  $M \subseteq E$  s.t.  $w(M) \geq (1 - \epsilon)\text{OPT}$

# Fast approximations: *bipartite matching*

	exact	approximate
cardinality	$O(m\sqrt{n})$ $\tilde{O}(m^{10/7})$ Hopcroft and Karp [1973] Mądry [2013]	$O(m/\epsilon)$
weighted	$O(mn + n^2 \log n)$ $O(m\sqrt{n} \log W)$ Fredman and Tarjan [1987] Duan and Su [2012]	$O(m \log(1/\epsilon)/\epsilon)$ Duan and Pettie [2014]

( $m = \text{edges}$ ,  $n = \text{vertices}$ ,  $W = \text{max weight}$ )

# Fast approximations: *bipartite matching*

	exact	approximate
cardinality	$O(m\sqrt{n})$ $\tilde{O}(m^{10/7})$ Hopcroft and Karp [1973] Mądry [2013]	$O(m/\epsilon)$
weighted	$O(mn + n^2 \log n)$ $O(m\sqrt{n} \log W)$ Fredman and Tarjan [1987] Duan and Su [2012]	$O(m \log(1/\epsilon)/\epsilon)$ Duan and Pettie [2014]

$(m = \text{edges}, n = \text{vertices}, W = \text{max weight})$

(!) For fixed  $\epsilon$ , weighted approximation is faster than unweighted exact

# Duan and Pettie [2014]

(extends to general matching)

1. Primal-dual
  - Only *approximates* dual optimal conditions
2. Scaling reduces weighted problem to unweighted
3. Runs an *approximate* subroutine at each scale
4. Updates dual variables with *small loss from approximation*

# Matroids

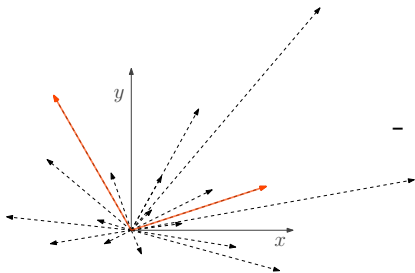
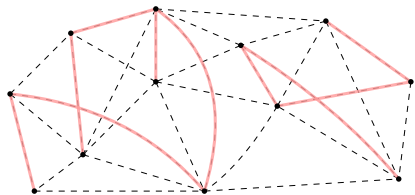
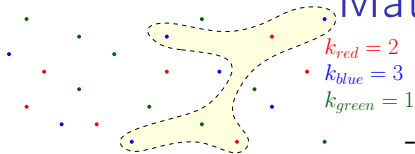
$$\mathcal{M} = (\mathcal{N}, \mathcal{I})$$

$\mathcal{N}$ : **ground set** of elements

$\mathcal{I} \subseteq 2^{\mathcal{N}}$ : **independent** (feasible) sets



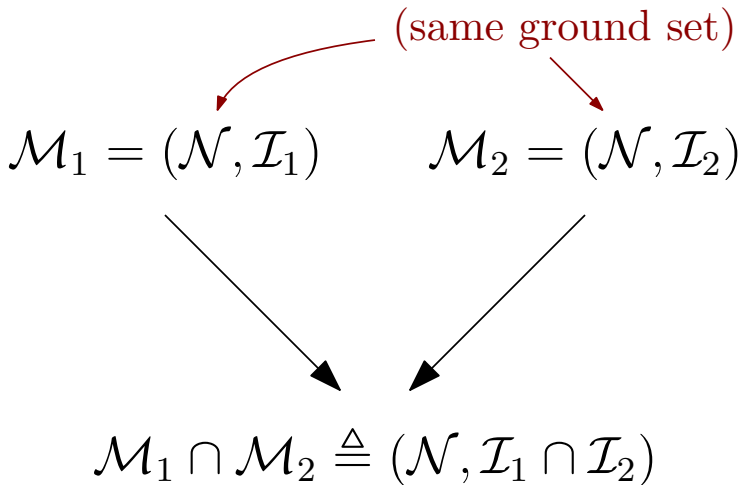
# Matroids



- Empty set is independent
  - Subsets of independent sets are independent
  - Maximal independent sets have same cardinality
- max independent set is a **base***  
*max cardinality is the **rank***

- If  $A, B \in \mathcal{I}$  and  $|A| < |B|$ , then there is  $b \in B \setminus A$  s.t.  $A + b \in \mathcal{I}$

# Matroid Intersection



e.g. bipartite matchings, arborescences

# Matroid intersection problems

## Maximum cardinality matroid intersection

*Input:* matroids  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$ ,  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$

*Output:*  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$  maximizing  $|S|$

# Matroid intersection problems

## Maximum cardinality matroid intersection

*Input:* matroids  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$ ,  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$

*Output:*  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$  maximizing  $|S|$

## Maximum weight matroid intersection

*Input:* matroids  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$ ,  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  
weights  $w : \mathcal{N} \rightarrow \mathbb{R}_{\geq 0}$

*Output:*  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$  maximizing  $w(S) \stackrel{\text{def}}{=} \sum_{e \in S} w(e)$

# Matroid intersection problems

## Maximum cardinality matroid intersection

*Input:* matroids  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$ ,  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$

*Output:*  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$  maximizing  $|S|$

## Maximum weight matroid intersection

*Input:* matroids  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$ ,  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  
weights  $w : \mathcal{N} \rightarrow \mathbb{R}_{\geq 0}$

*Output:*  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$  maximizing  $w(S) \stackrel{\text{def}}{=} \sum_{e \in S} w(e)$

## Oracle Model

Independence queries of the form “Is  $S \in \mathcal{I}$ ?”

# Running times for matroid intersection

# Running times for matroid intersection

## Cardinality

$$O(nk^{1.5}Q)$$

Cunningham [1986]

$(n = |\mathcal{N}|, k = \text{rank}(\mathcal{M}_1 \cap \mathcal{M}_2), Q = \text{cost of indep. query})$

# Running times for matroid intersection

## Cardinality

$$O(nk^{1.5}Q)$$

Cunningham [1986]

( $n = |\mathcal{N}|$ ,  $k = \text{rank}(\mathcal{M}_1 \cap \mathcal{M}_2)$ ,  $Q = \text{cost of indep. query}$ )

## Weighted

( $W = \max\{w(e) : e \in \mathcal{N}\}$ )

$$O(nk^2Q)$$

Frank [1981], Brezovec et al. [1986], Schrijver [2003]

$$O(n^2\sqrt{k} \log(kW)Q)$$

Fujishige and Zhang [1995]

$$O(nk^{1.5}WQ)$$

Huang et al. [2014]

$$O((n^2 \log(n)Q + n^3 \text{polylog}(n)) \log(nW))$$

Lee et al. [2015]



# Approximate matroid intersection

**Input:**  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$ ,  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  
 $w : \mathcal{N} \rightarrow \mathbb{R}_{\geq 0}$ ,  $\epsilon > 0$

**Output:**  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$  s.t.  $w(S) \geq (1 - \epsilon)\text{OPT}$   
( $\text{OPT} = \max\{w(T) : T \in \mathcal{I}_1 \cap \mathcal{I}_2\}$ )

**Previous bound:**

$O(nk^{1.5} \log(k)Q/\epsilon)$

Huang et al. [2014]

# Approximate matroid intersection

**Input:**  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$ ,  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  
 $w : \mathcal{N} \rightarrow \mathbb{R}_{\geq 0}$ ,  $\epsilon > 0$

**Output:**  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$  s.t.  $w(S) \geq (1 - \epsilon)\text{OPT}$   
( $\text{OPT} = \max\{w(T) : T \in \mathcal{I}_1 \cap \mathcal{I}_2\}$ )

**Previous bound:**

$O(nk^{1.5} \log(k)Q/\epsilon)$

Huang et al. [2014]

**Main result:**  $(1 - \epsilon)$ -approximation in time

$O(nkQ \log^2(\epsilon)/\epsilon^2)$

# Fast approximations: *matroid intersection*

	exact	approximate
cardinality	$O(nk^{1.5}Q)$ <small>Cunningham [1986]</small>	$O(nkQ/\epsilon)$
weighted	$(nk^2Q)$ <small>Frank [1981] and others</small> $O(n^2\sqrt{k}\log(kW)Q)$ <small>Fujishige and Zhang [1995]</small> $O(nk^{1.5}WQ)$ <small>Huang et al. [2014]</small> $O((n^2\log(n)Q + n^3\text{polylog}(n))\log(nW))$ <small>Lee et al. [2015]</small>	$O(nkQ\log^2(1/\epsilon)/\epsilon^2)$

$(n = \text{elements}, k = \text{rank}, Q = \text{indep. query}, W = \text{max weight})$

# Fast approximations: *matroid intersection*

	exact	approximate
cardinality	$O(nk^{1.5}Q)$ <span style="color: green;">Cunningham [1986]</span>	$O(nkQ/\epsilon)$
weighted	$(nk^2Q)$ <span style="color: green;">Frank [1981] and others</span> $O(n^2\sqrt{k}\log(kW)Q)$ <span style="color: green;">Fujishige and Zhang [1995]</span> $O(nk^{1.5}WQ)$ <span style="color: green;">Huang et al. [2014]</span> $O((n^2\log(n)Q + n^3\text{polylog}(n))\log(nW))$ <span style="color: green;">Lee et al. [2015]</span>	$O(nkQ\log^2(1/\epsilon)/\epsilon^2)$

$(n = \text{elements}, k = \text{rank}, Q = \text{indep. query}, W = \text{max weight})$

(!) For fixed  $\epsilon$ , weighted approximation is faster than unweighted exact

approximate  
cardinality

 + 

approximate  
scaling

⇒ approximate  
solution


# Some terminology

Matroid  $\mathcal{M} = (\mathcal{N}, \mathcal{I})$ ,  $S \in \mathcal{I}$

- **Span**:  $e \in \text{span}(S) \Rightarrow S + e \notin \mathcal{I}$
- **Free**:  $e \in \text{free}(S) \Rightarrow S + e \in \mathcal{I}$

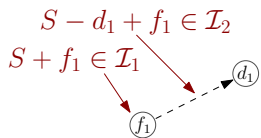
# Exchanges and exchange graphs

matroid  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$  and  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$

$$S + f_1 \in \mathcal{I}_1$$
A red arrow points from the text  $S + f_1 \in \mathcal{I}_1$  to a circled element  $f_1$ .

# Exchanges and exchange graphs

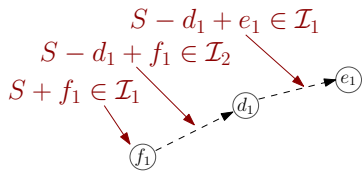
matroid  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$  and  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$





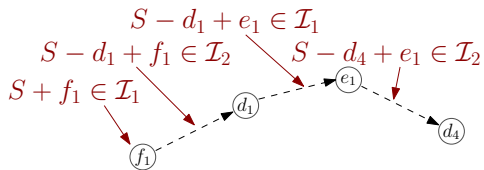
# Exchanges and exchange graphs

matroid  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$  and  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$



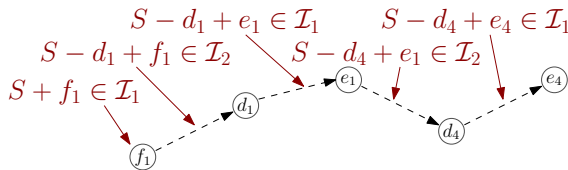
# Exchanges and exchange graphs

matroid  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$  and  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$



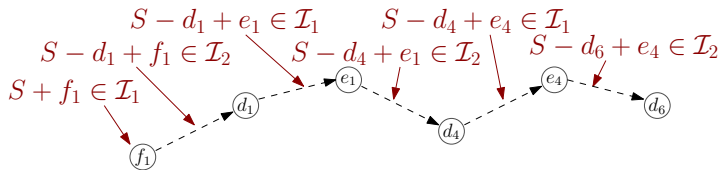
# Exchanges and exchange graphs

matroid  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$  and  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$



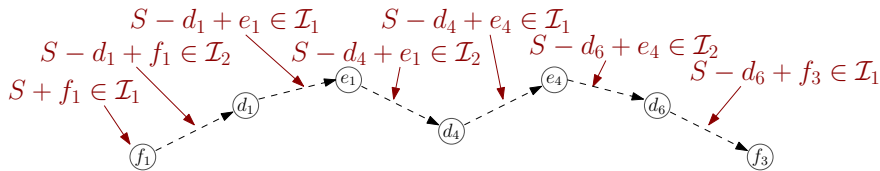
# Exchanges and exchange graphs

matroid  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$  and  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$



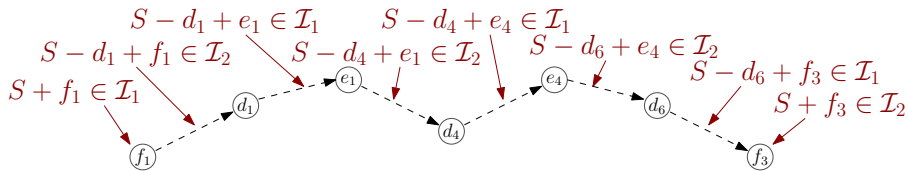
# Exchanges and exchange graphs

matroid  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$  and  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$



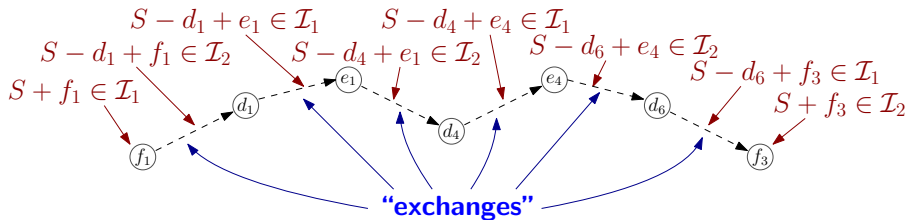
# Exchanges and exchange graphs

matroid  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$  and  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$



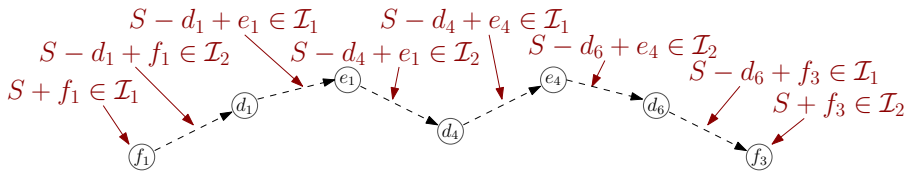
# Exchanges and exchange graphs

matroid  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$  and  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$



# Exchanges and exchange graphs

matroid  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$  and  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$

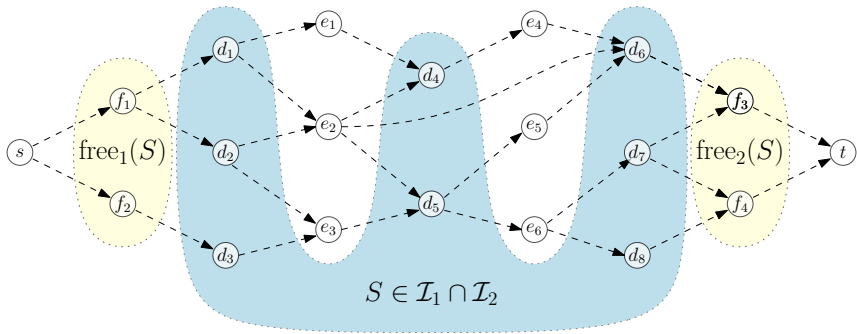


**“augmenting path”** if  $S + f_1 - d_1 + e_1 - d_4 + e_4 - d_6 + f_3 \in \mathcal{I}_1 \cap \mathcal{I}_2$



# Exchanges and exchange graphs

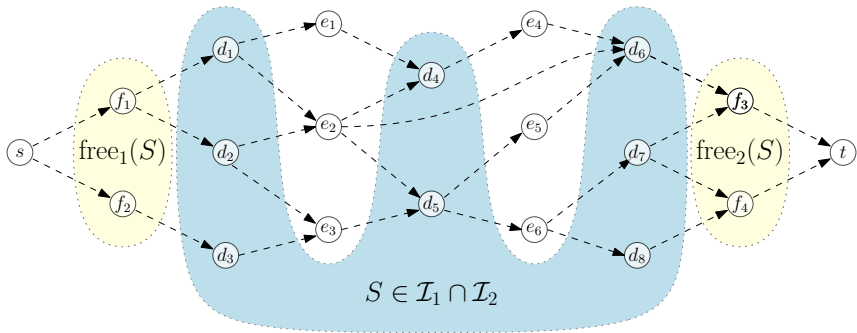
matroid  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$  and  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$



**exchange graph**

# Exchanges and exchange graphs

matroid  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$  and  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$

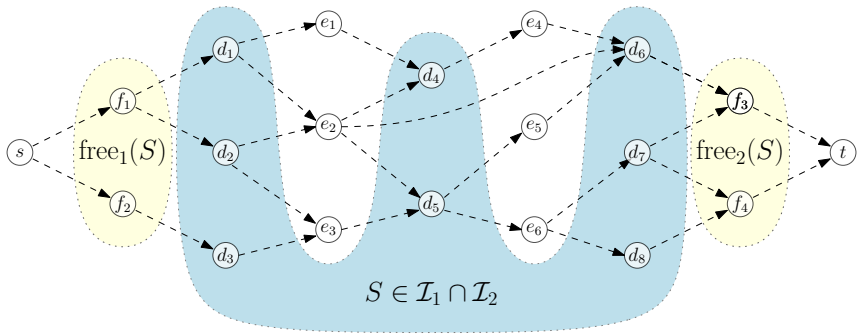


*Bad news:*

*Good news:*

# Exchanges and exchange graphs

matroid  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$  and  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$

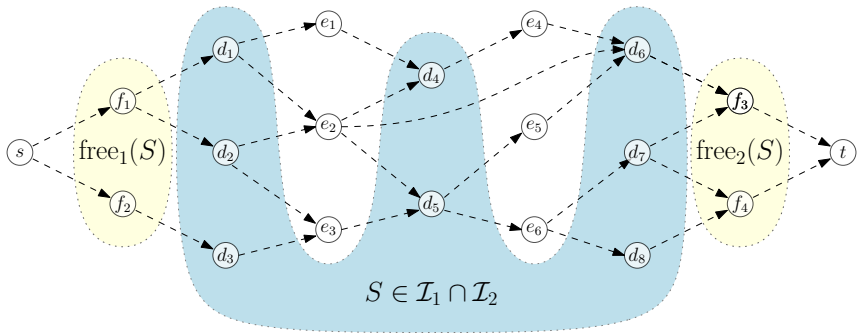


*Bad news:* not all paths are augmenting

*Good news:*

# Exchanges and exchange graphs

matroid  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$  and  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$

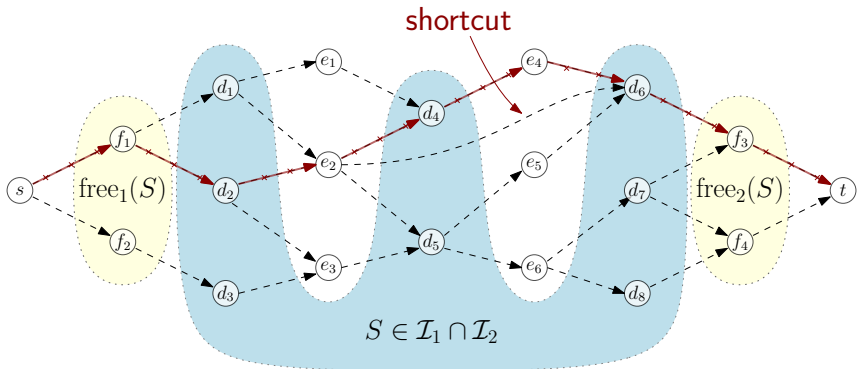


*Bad news:* not all paths are augmenting

*Good news:* shortest paths are augmenting

# Exchanges and exchange graphs

matroid  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$  and  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$

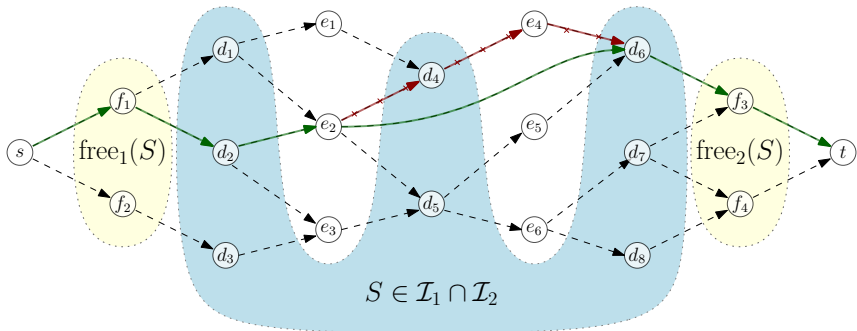


*Bad news:* not all paths are augmenting

*Good news:* shortest paths are augmenting

# Exchanges and exchange graphs

matroid  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$  and  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$



*Bad news:* not all paths are augmenting

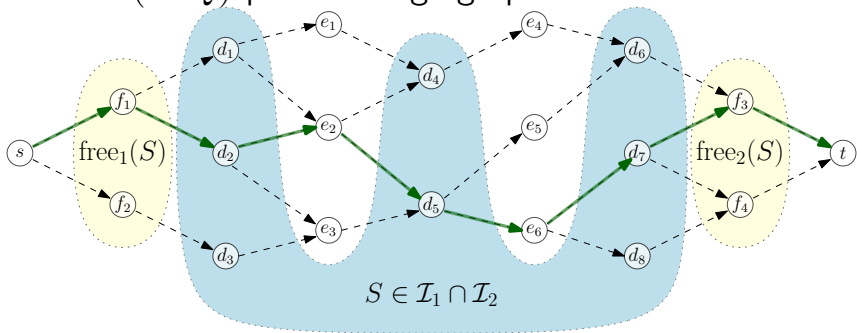
*Good news:* shortest paths are augmenting

# Unweighted matroid intersection

**Naive alg.:** augmenting paths

$O(nk^2Q)$

$O(nkQ)$  per exchange graph  $\times k$  iterations

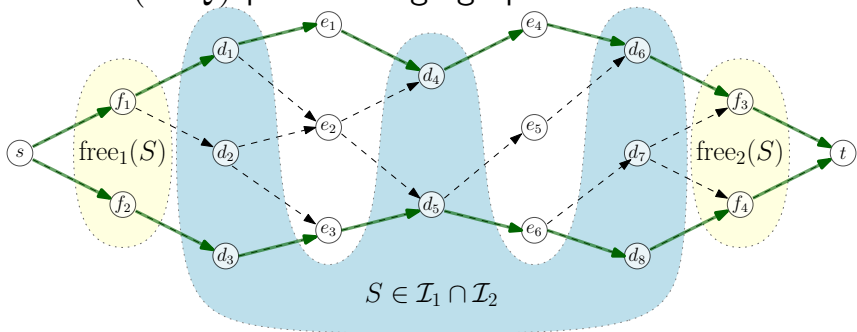


# Unweighted matroid intersection

**Naive alg.:** augmenting paths

$O(nk^2Q)$

$O(nkQ)$  per exchange graph  $\times k$  iterations



**Cunningham [1986]:**

$O(nk^{1.5}Q)$

Augment along several paths in each iteration



# Cunningham's algorithm

Similar to Hopcroft-Karp for bipartite matching

# Cunningham's algorithm

Similar to Hopcroft-Karp for bipartite matching

1. Increase length of short. aug. path in each iteration

**subroutine:** batch-augment( $S, \mathcal{M}_1, \mathcal{M}_2$ )

**input:**  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$  with length of short. aug. path =  $\ell$

**output:**  $S' \in \mathcal{I}_1 \cap \mathcal{I}_2$  with length of short. aug. path  $\geq \ell + 2$

**running time:**  $O(nkQ)$

# Cunningham's algorithm

Similar to Hopcroft-Karp for bipartite matching

1. Increase length of short. aug. path in each iteration

**subroutine:** batch-augment( $S, \mathcal{M}_1, \mathcal{M}_2$ )

**input:**  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$  with length of short. aug. path =  $\ell$

**output:**  $S' \in \mathcal{I}_1 \cap \mathcal{I}_2$  with length of short. aug. path  $\geq \ell + 2$

**running time:**  $O(nkQ)$

2. Most augmenting paths are short

$\Rightarrow O(\sqrt{k})$  iterations

```
Cunningham( $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1), \mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ )  
   $S \leftarrow \emptyset$   
  repeat until  $S$  is fixed  
     $S \leftarrow \text{batch-augment}(S, \mathcal{M}_1, \mathcal{M}_2)$  //  $O(nkQ)$   
  // After  $\sqrt{k}$  iterations,  $|S| \geq k - \sqrt{k}$   
  //  $O(\sqrt{k})$  iterations total  
  return  $S$ 
```

```
Cunningham( $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1), \mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ )
   $S \leftarrow \emptyset$ 
  repeat until  $S$  is fixed
     $S \leftarrow \text{batch-augment}(S, \mathcal{M}_1, \mathcal{M}_2)$  //  $O(nkQ)$ 
  // After  $\sqrt{k}$  iterations,  $|S| \geq k - \sqrt{k}$ 
  //  $O(\sqrt{k})$  iterations total
  return  $S$ 
```

```
Cunningham-APX( $\mathcal{M}_1, \mathcal{M}_2$ )
   $S \leftarrow \emptyset$ 
  repeat  $O(1/\epsilon)$  times
     $S \leftarrow \text{batch-augment}(S, \mathcal{M}_1, \mathcal{M}_2)$  //  $O(nkQ)$ 
  //  $|S| \geq (1 - \epsilon)k$  (!)
  return  $S$  //  $O(nkQ/\epsilon)$ 
```

# Fast approximations: *matroid intersection*

	exact	approximate
cardinality	$O(nk^{1.5}Q)$ Cunningham [1986]	$O(nkQ/\epsilon)$
weighted	$(nk^2Q)$ Frank [1981] and others $O(n^2\sqrt{k}\log(kW)Q)$ Fujishige and Zhang [1995] $O(nk^{1.5}WQ)$ Huang et al. [2014] $O((n^2\log(n)Q + n^3\text{polylog}(n))\log(nW))$ Lee et al. [2015]	$O(nkQ\log^2(1/\epsilon)/\epsilon^2)$

( $n$  = elements,  $k$  = rank,  $Q$  = indep. query,  $W$  = max weight)

# Fast approximations: *matroid intersection*

	exact	approximate
cardinality	$O(nk^{1.5}Q)$ <a href="#">Cunningham [1986]</a>	$O(nkQ/\epsilon)$
weighted	$(nk^2Q)$ <a href="#">Frank [1981]</a> and others $O(n^2\sqrt{k}\log(kW)Q)$ <a href="#">Fujishige and Zhang [1995]</a> $O(nk^{1.5}WQ)$ <a href="#">Huang et al. [2014]</a> $O((n^2\log(n)Q + n^3\text{polylog}(n))\log(nW))$ <a href="#">Lee et al. [2015]</a>	$O(nkQ\log^2(1/\epsilon)/\epsilon^2)$

$(n = \text{elements}, k = \text{rank}, Q = \text{indep. query}, W = \text{max weight})$

# Frank [1981]

**Input:** matroids  $\mathcal{M}_1 = (\mathcal{N}, \mathcal{I}_1)$  and  $\mathcal{M}_2 = (\mathcal{N}, \mathcal{I}_2)$ ,  
weights  $w : \mathcal{N} \rightarrow \mathbb{R}_{\geq 0}$

**Output:**  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$  maximizing  $w(S)$

**Running time:**  $O(nk^2Q)$



# 1. Weight Splittings

Frank [1981]

$$w_1, w_2 : \mathcal{N} \rightarrow \mathbb{R}_{\geq 0} \text{ s.t. } w = w_1 + w_2$$

**Fact:** For  $S \in \mathcal{I}_1 \cap \mathcal{I}_2$ , if

(a)  $w_1(S) = \max_{\mathcal{I}_1} w_1(T)$ , and

(b)  $w_2(S) = \max_{\mathcal{I}_2} w_2(T)$

then  $w(S) = \max\{w(T) : T \in \mathcal{I}_1 \cap \mathcal{I}_2\}$

## 2. Weight-induced matroid Frank [2008]

$$\mathcal{M}^w = (\mathcal{N}, \mathcal{I}^w)$$

$B \subseteq \mathcal{N}$  is a base in  $\mathcal{M}^w$



$B$  is a max. weight base in  $\mathcal{M}$  w/r/t  $w$

Reduces weighted matroid problems to unweighted problems

## 2. Weight-induced matroid Frank [2008]

$$\mathcal{M}^w = (\mathcal{N}, \mathcal{I}^w)$$

$B \subseteq \mathcal{N}$  is a base in  $\mathcal{M}^w$



$B$  is a max. weight base in  $\mathcal{M}$  w/r/t  $w$

Reduces weighted matroid problems to unweighted problems

Frank's algorithm maintains weight splitting  $w_1 + w_2 = w$  and  $S \in \mathcal{I}_1^{w_1} \cap \mathcal{I}_2^{w_2}$ .

```
Frank( $\mathcal{M}_1, \mathcal{M}_2, w : \mathcal{N} \rightarrow \mathbb{R}_{\geq 0}$ )  
   $S \leftarrow \emptyset, w_1 \leftarrow w, w_2 \leftarrow 0$   
  //  $w_2(f) = 0$  for all  $f \in \text{free}_2(S)$ 
```

```
  //  $w_1(f) = 0$  for all  $f \in \text{free}_1(S)$   
  //  $w_2(f) = 0$  for all  $f \in \text{free}_2(S)$   
  return  $S$ 
```

```

Frank( $\mathcal{M}_1, \mathcal{M}_2, w : \mathcal{N} \rightarrow \mathbb{R}_{\geq 0}$ )
   $S \leftarrow \emptyset, w_1 \leftarrow w, w_2 \leftarrow 0$ 
  //  $w_2(f) = 0$  for all  $f \in \text{free}_2(S)$ 
  let  $c$  denote  $\max\{w_1(f) : f \in \text{free}_1(S)\}$ 
  while  $c > 0$ 
    //  $\text{free}_1(f) \leq c$  for all  $f \in \text{free}_1(S)$ 

end while
//  $w_1(f) = 0$  for all  $f \in \text{free}_1(S)$ 
//  $w_2(f) = 0$  for all  $f \in \text{free}_2(S)$ 
return  $S$ 

```

```

Frank( $\mathcal{M}_1, \mathcal{M}_2, w : \mathcal{N} \rightarrow \mathbb{R}_{\geq 0}$ )
   $S \leftarrow \emptyset, w_1 \leftarrow w, w_2 \leftarrow 0$ 
  //  $w_2(f) = 0$  for all  $f \in \text{free}_2(S)$ 
  let  $c$  denote  $\max\{w_1(f) : f \in \text{free}_1(S)\}$ 
  while  $c > 0$ 
    //  $\text{free}_1(f) \leq c$  for all  $f \in \text{free}_1(S)$ 
     $\mathcal{N}_c = \{e : w(e) \geq c\}$ 
    augment  $S$  in  $(\mathcal{M}_1^{w_1} \cap \mathcal{M}_2^{w_2})|_{\mathcal{N}_c}$  until optimal
    shift weight from  $w_1$  to  $w_2$ 

  end while
  //  $w_1(f) = 0$  for all  $f \in \text{free}_1(S)$ 
  //  $w_2(f) = 0$  for all  $f \in \text{free}_2(S)$ 
  return  $S$ 

```

$$O(nk^2Q)$$

```

HKK( $\mathcal{M}_1, \mathcal{M}_2, w : \mathcal{N} \rightarrow \mathbb{R}_{\geq 0}$ )
   $S \leftarrow \emptyset, w_1 \leftarrow w, w_2 \leftarrow 0$ 
  //  $w_2(f) = 0$  for all  $f \in \text{free}_2(S)$ 
  let  $c$  denote  $\max\{w_1(f) : f \in \text{free}_1(S)\}$ 
  for  $c = W, W - 1, \dots, 3, 2, 1$ 
    //  $\text{free}_1(f) \leq c$  for all  $f \in \text{free}_1(S)$ 
     $\mathcal{N}_c = \{e : w(e) \geq c\}$ 
    run Cunningham on  $S$  in  $(\mathcal{M}_1^{w_1} \cap \mathcal{M}_2^{w_2})|_{\mathcal{N}_c}$ 
    shift one unit of weight from  $w_1$  to  $w_2$ 

  end for
  //  $w_1(f) = 0$  for all  $f \in \text{free}_1(S)$ 
  //  $w_2(f) = 0$  for all  $f \in \text{free}_2(S)$ 
  return  $S$ 

```

$$O(nk^{1.5}WQ)$$

approximate  
cardinality

 + 

approximate  
scaling

=

approximate  
weight splitting

⇒

approximate  
solution



```

HKK( $\mathcal{M}_1, \mathcal{M}_2, w : \mathcal{N} \rightarrow \mathbb{R}_{\geq 0}$ )
   $S \leftarrow \emptyset, w_1 \leftarrow w, w_2 \leftarrow 0$ 
  //  $w_2(f) = 0$  for all  $f \in \text{free}_2(S)$ 
  let  $c$  denote  $\max\{w_1(f) : f \in \text{free}_1(S)\}$ 
  for  $c = W, W - 1, \dots, 3, 2, 1$ 
    //  $\text{free}_1(f) \leq c$  for all  $f \in \text{free}_1(S)$ 
     $\mathcal{N}_c = \{e : w(e) \geq c\}$ 
    run Cunningham on  $S$  in  $(\mathcal{M}_1^{w_1} \cap \mathcal{M}_2^{w_2})|_{\mathcal{N}_c}$ 
    shift one unit of weight from  $w_1$  to  $w_2$ 

  end for
  //  $w_1(f) = 0$  for all  $f \in \text{free}_1(S)$ 
  //  $w_2(f) = 0$  for all  $f \in \text{free}_2(S)$ 
  return  $S$ 

```

$$O(nk^{1.5}WQ)$$

Frank-APX( $\mathcal{M}_1, \mathcal{M}_2, w : \mathcal{N} \rightarrow \mathbb{R}_{\geq 0}, \epsilon > 0$ )

$S \leftarrow \emptyset, w_1 \leftarrow w, w_2 \leftarrow 0$

//  $w_2(f) = 0$  for all  $f \in \text{free}_2(S)$

let  $c$  denote  $\max\{w_1(f) : f \in \text{free}_1(S)\}$

for  $c = W, W/2, \dots, 4\epsilon, 2\epsilon, \epsilon$

//  $\text{free}_1(f) \leq c$  for all  $f \in \text{free}_1(S)$

$\mathcal{N}_c = \{e : w(e) \geq c\}$

repeat  $1/\epsilon$  times:

run Cunningham-APX on  $S$  in  $(\mathcal{M}_1^{w_1} \cap \mathcal{M}_2^{w_2})|_{\mathcal{N}_c}$

shift  $\epsilon c$  units of weight from  $w_1$  to  $w_2$

end for

//  $w_1(f) = 0$  for all  $f \in \text{free}_1(S)$

//  $w_2(f) = 0$  for all  $f \in \text{free}_2(S)$

return  $S$

$$O(nkQ \log^2(1/\epsilon)/\epsilon^2)$$

approximate  
cardinality

 + 

approximate  
scaling

=

approximate  
weight splitting

⇒

approximate  
solution

# thank you

approximate  
cardinality

 + 

approximate  
scaling

=

approximate  
weight splitting

⇒

approximate  
solution

# Bibliography I

- C. Brezovec, G. Cornuéjols, and F. Glover. Two algorithms for weighted matroid intersection. *Math. Prog.*, 36(1):39–53, 1986.
- W. H. Cunningham. Improved bounds for matroid partition and intersection algorithms. *SIAM J. Comput.*, 15(4):948–957, 1986.
- R. Duan and S. Pettie. Linear-time approximation for maximum weight matching. *J. Assoc. Comput. Mach.*, 61(1), January 2014.
- R. Duan and H. Su. A scaling algorithm for maximum weight matching in bipartite graphs. In *Proc. 23rd ACM-SIAM Sympos. Discrete Algs. (SODA)*, pages 1413–1424, 2012.
- A. Frank. A weighted matroid intersection algorithm. *J. Algorithms*, 2: 328–336, 1981.
- A. Frank. A quick proof of the matroid intersection weight-splitting theorem. *EGRES Quick-Proofs*, 3, 2008.
- M. L. Fredman and R. E. Tarjan. Fibonacci heaps and their uses in improved network optimization algorithms. *J. Assoc. Comput. Mach.*, 34(3):596–615, 1987.

# Bibliography II

- S. Fujishige and X. Zhang. An efficient cost scaling algorithm for the independent assignment problem. *J. Oper. Res. Soc. Japan*, 38(1): 124–136, 1995.
- J. Hopcroft and R. Karp. An  $n^{5/2}$  algorithm for maximum matchings in bipartite graphs. *SIAM J. Comput.*, pages 225–231, 1973.
- C. Huang, N. Kakimura, and N. Kamiyama. Exact and approximation algorithms for weighted matroid intersection. *MI Preprint Series, Math. for Industry, Kyushu University*, November 2014. To appear in Proc. 27th ACM-SIAM Symp. Discrete Algs. (*SODA*), 2016.
- Y.T. Lee, A. Sidford, and S.C. Wong. A faster cutting plane method and its implications for combinatorial and convex optimization. In *Proc. 56th Annu. IEEE Symp. Found. Comput. Sci. (FOCS)*, 2015.
- A. Mądry. Navigating central path with electrical flows: from flows to matchings, and back. In *Proc. 54th Annu. IEEE Symp. Found. Comput. Sci. (FOCS)*, pages 253–262, 2013.
- A. Schrijver. *Combinatorial Optimization: Polyhedra and Efficiency*, volume 24 of *Algorithms and Combinatorics*. Springer, 2003.