



# Optimal assignments with supervisions

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# BASIC DEFINITIONS AND CONCEPTS

# Tropical linear algebra

- Consider real numbers  $\mathbb{R} \cup \{-\infty\}$  equipped with

$$a \odot b = a + b, \quad a \oplus b := \max(a, b).$$

- Semifield with  $\mathbf{0} = -\infty$ ,  $\mathbf{1} = 0$ .

**I.e.**  $a^{-1} = -a$  and  $\nexists \ominus a$ .

- Applies to matrices and vectors entry-wise:

$$(A \oplus B)_{i,j} := (A_{i,j} \oplus B_{i,j})$$

$$(A \odot B)_{i,j} := \bigoplus_k A_{i,k} \odot B_{kj}$$

# Jacobi identity

Correspondence :  $I, J$  minor of  $A^{-1}$  to  $J^c, I^c$  minor of  $A$ .

Theorem (the classical identity)

For  $A \in GL_n(\mathbb{F})$ ,  $I, J \subseteq [n]$  s.t.  $|I| = |J| = k$

$$(DA^{-1}D)_{I,J}^{\wedge k} = (\det(A))^{-1} A_{J^c, I^c}^{\wedge n-k},$$

where  $D_{i,i} = (-1)^i$  and  $D_{i,j} = 0$  for  $i \neq j$ .

(for instance) S. M. Fallat and C. R. Johnson, *Totally Nonnegative Matrices*.  
Princeton press, 2011.

# Jacobi identity

## Theorem (the tropical identity)

Let  $M \in \mathbb{R}_{\max}^{n \times n}$  and  $I, J \subseteq [n]$  s.t.  $|I| = |J| = k$ .

Either:

$$[D(\det(M)^{-1} \text{adj}(M))D]_{I,J}^{\wedge k} = \det(M)^{-1} M_{J^c, I^c}^{\wedge n-k}$$

Or:

There exist distinct bijections  $\pi, \sigma \in S_{I,J}$  such that

$$[\text{adj}(M)]_{I,J}^{\wedge k} = \bigodot_{i \in I} \text{adj}(M)_{i, \pi(i)} = \bigodot_{i \in I} \text{adj}(M)_{i, \sigma(i)}.$$

M. Akian, S. Gaubert and N. TROPICAL Compound Matrix Identities, LAA, 2018.

## How did it form?

The tropical *determinant* is actually the *permanent* with respect to  $\oplus, \odot$ . That is

$$\text{per}(A) = \bigoplus_{\pi \in S_n} \bigodot_{i \in [n]} A_{i, \pi(i)} = \max_{\pi \in S_n} \sum_{i \in [n]} A_{i, \pi(i)},$$

Graphically: the permutation of optimal weight in the graph of  $A$ ,  
Combinatorially: the 'optimal assignment problem'.

Let  $\pi, \tau$  be permutations of identical weight  $w$ .

- \* In supertropical  $w(\pi) \oplus w(\tau)$  is singular.
- \* In symmetrized  $w(\pi) \oplus w(\tau)$  is singular if  $\pi$  and  $\tau$  are permutations of opposite signs.

## How did it form?

- 2013 - PhD (with L.Rowen) - Conjecture: Let  $A^\nabla = \text{per}^{-1}(A)\text{adj}(A)$  (sort of inverse). Then (supertropically) coefficient-wise  $\text{per}(A)f_{A^\nabla}(x) = x^n f_A(x^{-1}) \oplus$  'singular polynomial'.

**That is,  $\oplus \mathbf{A}_{I,I}^\nabla$  corresponds to  $\oplus \mathbf{A}_{I^c,I^c}$ .**

[Y.Shitov 'On the Char. Polynomial of a Supertropical Adjoint Matrix', LAA.]

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- 2015 - Postdoc (with M.Akian and S.Gaubert) - (symmetrized) Tropical Jacobi:  
$$[D(\det(M)^{-1}\text{adj}(M))D]_{I,J}^{\wedge k} = \det(M)^{-1} M_{J^c,I^c}^{\wedge n-k} \oplus \text{'singular matrix'}$$
  
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 So, **entry-wise**, for **every**  $I, J$ , and **including signs**.

- 2016-2018 (with McCaig and Sergeev) - Graph theory version:  
 Every optimal  $(1, k)$ -regular multigraph of  $M$  w.r.t.  $I, J$   
**either:** corresponds to an optimal bijection w.r.t.  $I^c, J^c$ ,  
**or:** there exists another optimal  $(1, k)$ -regular w.r.t.  $I, J$ .  
 [That is, combinatorially, without signs, which led to the application.]

## Definitions: digraphs

- A **weighted digraph**  $G$  is a pair  $(V_G, E_G)$  where
  - $V_G$  is **set of nodes** and
  - $E_G \subseteq V_G \times V_G$  is **set of directed edges** on  $|V_G|$  nodes (allowing loops and multiple edges).
  - **Weight:**  $w(i, j)$  for each  $(i, j)$ .
- A **bipartite graph** is a triple  $(V_{H,1}, V_{H,2}, E_H)$  s.t.  
 $i \in V_{H,1} \Leftrightarrow j \in V_{H,2}$  for every  $(i, j) \in E_H$ , **weighted:**  $w(i, j)$  for each  $(i, j)$ .

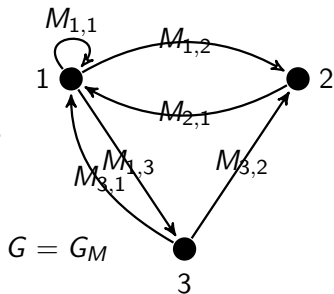
## Associated digraphs

- **Matrix**  $M \in \mathbb{R}_{\max}^{n \times n} \longrightarrow$  **weighted digraph**  $G_M = (V, E)$ ,  
where  $V = [n]$  and  $E = \{(i, j) : M_{i,j} \neq \mathbf{0}\}$ ,  
and weight  $w(i, j) = M_{i,j}$ .
- **Weighted digraph**  $G = ([n], E, w) \longrightarrow$  **matrix**  $M_G$ ,

$$\text{where } (M_G)_{i,j} = \begin{cases} w(i, j) & ; \text{ if } (i, j) \in E, \\ \mathbf{0} & ; \text{ otherwise.} \end{cases}$$

# Digraphs and matrices

$$M = M_G = \begin{pmatrix} M_{1,1} & M_{1,2} & M_{1,3} \\ M_{2,1} & \mathbf{0} & \mathbf{0} \\ M_{3,1} & M_{3,2} & \mathbf{0} \end{pmatrix}$$



## Associated bipartite graphs

- **Matrix**  $M \in \mathbb{R}_{\max}^{m \times n} \longrightarrow$  **bipartite graph**  $G_M = (V_{H_1}, V_{H_2}, E_H)$ ,  
 $|V_{H_1}| = m$ ,  $|V_{H_2}| = n$ , and  $E_H = \{(i, j) : M_{i,j} \neq -\infty\}$ ,  
 weight  $w(i, j) = M_{i,j}$ .

- **Bipartite graph**  $G = (V_{H_1}, V_{H_2}, E_H) \longrightarrow$  **matrix**  $M_G \in \mathbb{R}_{\max}^{m \times n}$   
 $|V_{H_1}| = m$ ,  $|V_{H_2}| = n$

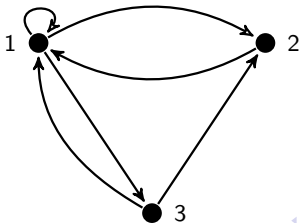
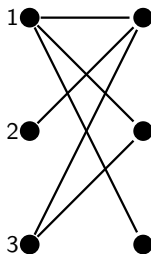
where

$$(M_G)_{i,j} = \begin{cases} w(i, j) & ; \text{ if } (i, j) \in E_H, \\ \mathbf{0} & ; \text{ otherwise.} \end{cases}$$

- **Digraph**  $DG = ([n], E_D) \longleftrightarrow$  **bipartite graph**  $BG = ([2n], E_B)$ ,  
 s.t.  $(i, j + n) \in E_B$  for every  $(i, j) \in E_D$ .

# Bipartite graphs and matrices

$$M = \begin{pmatrix} M_{1,1} & M_{1,2} & M_{1,3} \\ M_{2,1} & \mathbf{0} & \mathbf{0} \\ M_{3,1} & M_{3,2} & \mathbf{0} \end{pmatrix}$$



## Definitions: assignment problems

- Let  $S_n$  denote the set of permutations on  $[n]$ , and  $S_{I,J}$  denote the set of bijections from  $I \subseteq [n]$  to  $J \subseteq [n]$  (that is,  $|I| = |J|$ ).
- For  $M \in \mathbb{R}_{\max}^{n \times n}$  **tropical permanent** is defined by

$$\text{per}(M) = \max_{\pi \in S_n} \sum_{i \in [n]} M_{i, \pi(i)} = \bigoplus_{\pi \in S_n} \bigodot_{i \in [n]} M_{i, \pi(i)}.$$

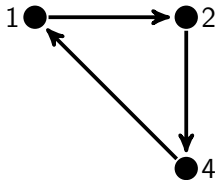
- A permutation  $\pi$  of maximal weight in  $\text{per}(M)$  is an **optimal permutation** in  $M$  or  $G_M$ . That is,

$$\text{per}(M) = \bigodot_{i \in [n]} M_{i, \pi(i)} = \sum_{i \in [n]} w(i, \pi(i)).$$

- This is identical to the set of **optimal assignments**, i.e., optimal solutions to the assignment problem in the bipartite graph associated with  $M$ .

# Permutation subgraphs

- A non- $\mathbf{0}$  tropical "summand"  $w(\pi) = \bigodot_{i \in [n]} M_{i, \pi(i)}$  in  $\text{per}M$ , or in  $M \leftrightarrow$  **permutation-subgraph** of  $G_M$  with  $V(E_\pi) = [n]$ ,  $E_\pi = \{(i, \pi(i)) \mid \forall i \in [n]\}$ .



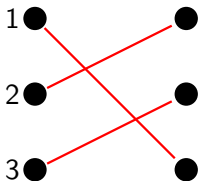
$$(1\ 2\ 4)(5\ 3)(6)$$

(and the same for path, cycle, bijection,...)



## Assignment subgraphs

- A non- $\mathbf{0}$  tropical "summand"  $w(\pi) = \odot_{i \in [n]} M_{i, \pi(i)}$  in  $\text{per} M$   
 $\leftrightarrow$  **assignment subgraph** with  
 $V(E_\pi) = [n] + [n]$ ,  $E_\pi = \{(i, \pi(i)) \mid i \in [n]\}$ .



(2 1 3)

(and the same for path, cycle, bijection,...)

## $k$ -regular graphs

- A graph or digraph  $G = (V, E)$  is  **$k$ -regular** if
$$\forall v \in V: \deg(v) = k \text{ (if } G \text{ is a graph)}$$
$$\forall v \in V: \deg^+(v) = \deg^-(v) = k \text{ (if } G \text{ is a digraph).}$$
- **Observation:** Let  $G = ([n], E)$  be a  $k$ -regular digraph, then

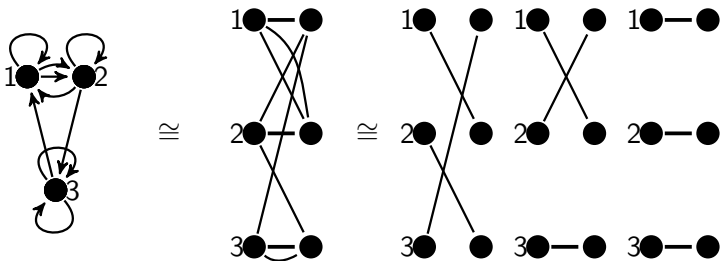
$$E = \bigsqcup_{i \in [k]} E_{\rho_i}, \quad \rho_i \in S_n$$

i.e., a disjoint union of edge sets of  $k$  permutation-subgraphs  $G_i = ([n], E_{\rho_i})$  for some  $\rho_i$ , for  $i \in [k]$ .

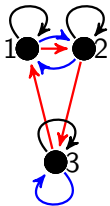
[Hall's Marriage Thm and Z.Izhakian and L.Rowen, Supertropical matrix algebra.]

- So  $G = ([n], \bigsqcup_{i \in [k]} E_{\rho_i})$ .

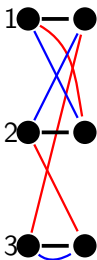
# Hall's Marriage Theorem



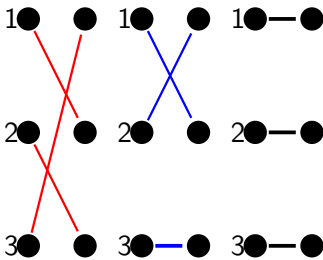
# Hall's Marriage Theorem



$\cong$



$\cong$



## $(1, k)$ -regular graphs

- Let  $G$  be  $k$ -regular (with  $\rho_1, \dots, \rho_k$ ). We say  $G$  is  **$(1, k)$ -regular** w.r.t.  $I, J$  with  $|I| = |J| = k$  if there exist  $e_i \in E_{\rho_i} \ \forall i \in [k]$  s.t.  $s(e_i) \in I, t(e_i) \in J$  and

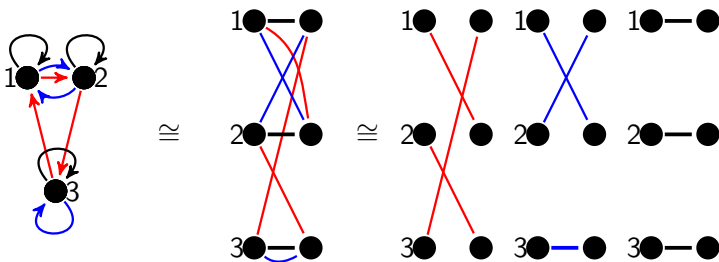
$$(V(E_\pi), E_\pi = \{e_1, \dots, e_k\})$$

is a bijection-subgraph.

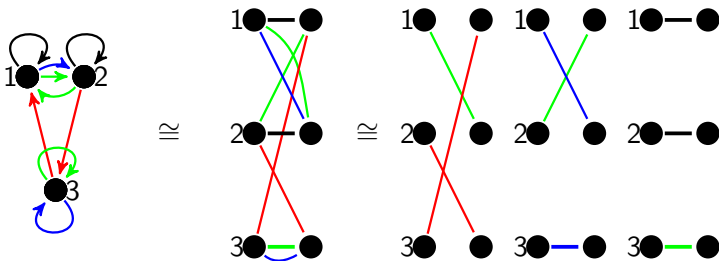
- We denote

$$G = ([n], \bigsqcup_{i \in [k]} E_{\rho_i}, \pi).$$

# Example: (1,3)-regular graph



# Example: (1,3)-regular graph



## Tropical adjugate

- Denote by  $M^{\wedge k} \in \mathbb{R}_{\max}^{\binom{n}{k} \times \binom{n}{k}}$  the **tropical  $k^{\text{th}}$  compound matrix** of  $M$  defined by

$$M_{I,J}^{\wedge k} = \bigoplus_{\sigma \in S_{I,J}} \bigodot_{i \in I} M_{i,\sigma(i)} = \max_{\sigma \in S_{I,J}} \sum_{i \in I} M_{i,\sigma(i)}$$

$\forall I, J \subseteq [n] : |I| = |J| = k, I, J$  ordered lexicographically.

- In particular,  $M^{\wedge 1} = M$ ,  $M^{\wedge 0} = \mathbf{1}$  and  $\text{per}(M) = M^{\wedge n}$  is the tropical permanent of  $M$ .
- $\text{adj}(M)_{i,j} = M_{\{j\}^c, \{i\}^c}^{\wedge n-1}$   
 is the  $(i j)$  entry of the **tropical adjugate** of  $M$ .



## Optimal $(1, k)$ -regular multigraphs

- We say that  $([n], \biguplus_{i \in [k]} E_{\rho_i}, \sigma)$  is an **optimal  $(1, k)$ -regular multigraph** of  $G$  w.r.t.  $I, J$  if

$$\left( \sum_{i \in [k]} w(\rho_i) \right) - w(\sigma) \geq \left( \sum_{i \in [k]} w(\rho'_i) \right) - w(\sigma'),$$

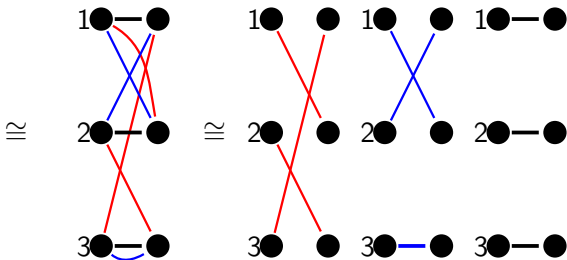
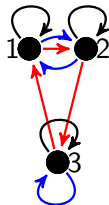
for every  $(1, k)$ -regular multigraph  $([n], \biguplus_{i \in [k]} E_{\rho'_i}, \sigma')$  of  $G$ .

- Equivalently**

$$(\text{adj}(M_G))_{J,I}^{\wedge k} = \bigodot_{i \in I} (\text{adj}(M_G))_{\sigma(i),i}, \text{ where}$$

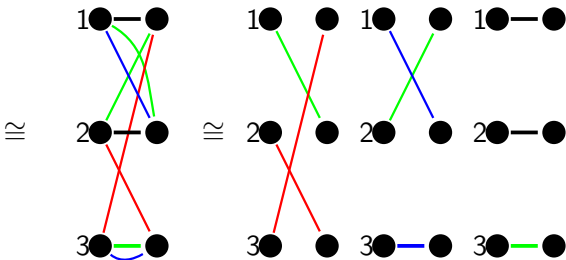
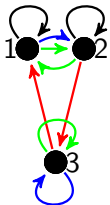
$$(\text{adj}(M_G))_{\sigma(i),i} = \bigodot_{j \in \{i\}^c} (M_G)_{j, \rho_i(j)}.$$

# Example: (1,3)-regular graph



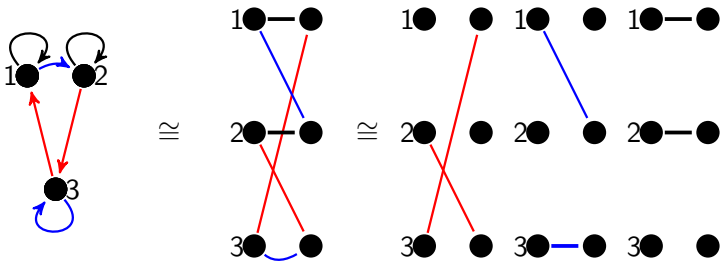
$$\left( \sum_{i \in [3]} w(\rho_i) \right)$$

# Example: (1,3)-regular graph



$$\left( \sum_{i \in [3]} w(\rho_i) \right), w(\sigma)$$

# Example: (1,3)-regular graph



$$\left( \sum_{i \in [3]} w(\rho_i) \right) - w(\sigma) = (\text{adj}(M_G))_{J,I}^{\wedge k}$$

# OPTIMAL ASSIGNMENTS WITH SUPERVISIONS

# Assignments with supervisions

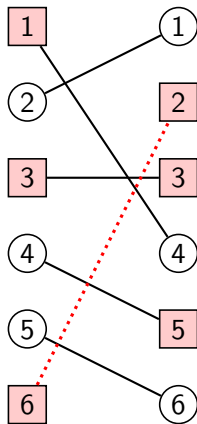
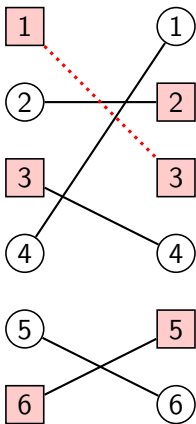
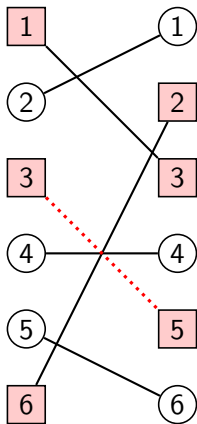
- Supervisions: Let
  - $M \in \mathbb{R}_{\max}^{n \times n}$
  - $\rho_t \in S_n$  for  $t \in [k]$  be  $k$  assignments,
  - $(i_t, j_t) \in I \times J$  be  $k$  edges s.t.  $\sigma(i_t) = j_t$  for  $\sigma \in S_{I,J}$ .

$\sigma$  defines **supervisions** on  $\{\rho_t : t \in [k]\}$  if  $\rho_t(i_t) = j_t \forall t$ .
- The **base value** of these assignments with supervisions is

$$\sum_{t \in [k]} (w(\rho_t, M) - M_{i_t, \sigma(i_t)}) = \sum_{t=1}^k \sum_{i \neq i_t} M_{i, \rho_t(i)}.$$

- This is also the **weight of  $(1, k)$ -regular multigraph**  
 $([n], \biguplus_{t \in [k]} E_{\rho_t}, \sigma)$ .

# Assignments with supervisions of people $\{1, 3, 6\}$ on tasks $\{2, 3, 5\}$



## Key observation

- The **optimal base value** of  $k$  assignments with supervisions  $I$  on  $J$  is

$$\bigoplus_{\sigma \in \mathcal{S}_{J,I}} w(\sigma, \text{adj}(M)_{J,I}) = [\text{adj}(M)]_{J,I}^{\wedge k}$$

- It is also the the weight of an **optimal  $(1, k)$ -regular multigraph** w.r.t.  $I$  and  $J$ .



## Example

- Let

$$M = \begin{pmatrix} 0 & 1 & -2 & -4 \\ -3 & 0 & 5 & 2 \\ -5 & 4 & 0 & 6 \\ -1 & -6 & 3 & 0 \end{pmatrix}, \text{ then } \text{adj}(M) = \begin{pmatrix} 9 & 10 & 6^\bullet & 12 \\ 10 & 9 & 5^\bullet & 11 \\ 5 & 6 & 2 & 6^\bullet \\ 8 & 9 & 5 & 9 \end{pmatrix}.$$

- Goal:** Find optimal assignments with supervisions of  $I = \{2, 4\}$  on  $J = \{1, 2\}$ .
- The **maximum base value** is given by

$$\text{adj}(M)_{J,I}^{\wedge 2} = \text{per} \begin{pmatrix} 10 & 12 \\ 9 & 11 \end{pmatrix} = 21^\bullet.$$

- The **optimal bijections (supervisions)** are  $\sigma_1 = (2 \rightarrow 1)(4 \rightarrow 2)$  and  $\sigma_2 = (2 \rightarrow 2)(4 \rightarrow 1)$ .

## Example: the end of solution

- We found that  $\sigma_1 : (2 \rightarrow 1)(4 \rightarrow 2)$  is optimal.
- Supervision  $2 \rightarrow 1$  corresponds to

$$M_{\{1,3,4\},\{2,3,4\}} = \begin{pmatrix} \mathbf{1} & -2 & -4 \\ 4 & 0 & \mathbf{6} \\ -6 & \mathbf{3} & 0 \end{pmatrix}.$$

$$\beta_1 = (1 \rightarrow 2)(3 \rightarrow 4)(4 \rightarrow 3) \in S_{\{1,3,4\},\{2,3,4\}},$$

$$\rho_1 = (1 \rightarrow 2)(\mathbf{2} \rightarrow \mathbf{1})(3 \rightarrow 4)(4 \rightarrow 3) \in S_4.$$

- For supervision  $4 \rightarrow 2$ , we similarly obtain:  
 $\beta_2 = (1 \rightarrow 1)(2 \rightarrow 3)(3 \rightarrow 4) \in S_{\{1,2,3\},\{1,3,4\}},$   
 $\rho_2 = (1 \rightarrow 1)(2 \rightarrow 3)(3 \rightarrow 4)(\mathbf{4} \rightarrow \mathbf{2}) \in S_4.$
- **Optimal  $(1, k)$ -regular multigraph:**  
 $F = (E_{\rho_1} \uplus E_{\rho_2}, \sigma_1).$

# TROPICAL JACOBI IDENTITY IN GRAPHS

## (non-symmetrized) Tropical Jacobi identity

### Theorem (Tropical Jacobi identity)

Let  $M \in \mathbb{R}_{\max}^{n \times n}$  and  $I, J \subseteq [n]$  such that  $|I| = |J| = k$ . Then:

- ①  $[\text{per}(M)^{-1} \text{adj}(M)]_{I,J}^{\wedge k} = \text{per}(M)^{-1} M_{J^c, I^c}^{\wedge n-k}$  **OR**
- ② *There exist distinct bijections  $\pi, \sigma \in S_{I,J}$  such that*

$$[\text{adj}(M)]_{I,J}^{\wedge k} = \sum_{i \in I} \text{adj}(M)_{i, \pi(i)} = \sum_{i \in I} \text{adj}(M)_{i, \sigma(i)}.$$

[M. Akian, S. Gaubert and N, Tropical compound matrix identities, LAA.]

## Tropical adjugate and optimal multigraphs

- $(\text{adj}M)_{J,I}^{\wedge k} =$   
**the weight of an optimal  $(1, k)$ -regular multigraph**  
 $F = ([n], \biguplus_{i \in [k]} E_{\rho_i}, \pi)$  w.r.t.  $I, J \subseteq [n]$ .

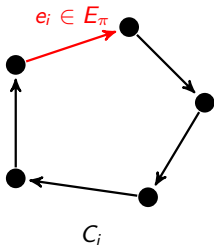
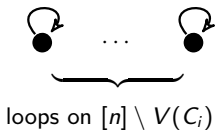
- We will assume that  $M_{i,j} = \mathbf{1}$  and  $\text{Id} \in S_n$  is an optimal assignment in  $M$ . That is,  $\text{per}(M) = \bigodot_{i \in [n]} M_{i,i} = \mathbf{1}$ .

Indeed, this normalization  $M \mapsto PM$  process is invertible, so by Binet-Cauchy and classical Jacobi, if tropical Jacobi holds for  $PM$ , it holds for  $M$ .

- This means  $\text{Id} \in S_k$  is an optimal assignment of weight  $\mathbf{1}$  in  $M$  for every  $k$ , and in particular, loops are 'equally or more optimal' than every cycle.

# Case of unicycle permutations $\rho_i$

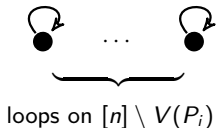
$\rho_i :$



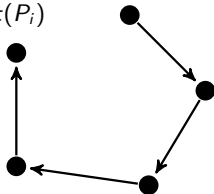
$C_i$

$$s(P_i) = t(e_i)$$

$\beta_i :$



$$s(e_i) = t(P_i)$$



$P_i$

# Tropical Jacobi identity in multigraphs

## Theorem

- Let  $\text{Id}$  be an optimal permutation in  $G = ([n], E)$ .
- Let  $F = ([n], \biguplus_{i \in [k]} E_{\rho_i}, \pi)$  be an optimal  $(1, k)$ -regular multigraph of  $G$  with respect to  $I, J \subseteq [n]$ .

EITHER:

$w(F) = w(\sigma)$  where  $\sigma \in S_{I^c, J^c}$  is an optimal bijection,

OR:

There exists  $\tilde{\pi} \in S_{I, J}$  and  $\tau_i \in S_n$  s.t.

$F' = ([n], \biguplus_{i \in [k]} E_{\tau_i}, \tilde{\pi}) \neq F$  is also an optimal  $(1, k)$ -regular multigraph with respect to  $I, J$ .

# Example

- Let

$$A = \begin{pmatrix} 0 & -1 & -5 & -4 \\ -6 & 0 & -2 & -1 \\ -3 & -4 & 0 & -3 \\ -2 & -7 & 0 & 0 \end{pmatrix}.$$

- Then

$$\text{adj}(A) = \begin{pmatrix} 0 & -1 & -2 & -2 \\ -3 & 0 & -1 & -1 \\ -3 & -4 & 0 & -3 \\ -2 & -3 & 0 & 0 \end{pmatrix} \leftrightarrow \left( \begin{array}{cccc} \begin{array}{c} \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \end{array} & \begin{array}{c} \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \end{array} & \begin{array}{c} \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \end{array} & \begin{array}{c} \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \end{array} \\ \begin{array}{c} \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \end{array} & \begin{array}{c} \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \end{array} & \begin{array}{c} \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \end{array} & \begin{array}{c} \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \end{array} \\ \begin{array}{c} \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \end{array} & \begin{array}{c} \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \end{array} \text{ or } \begin{array}{c} \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \end{array} & \begin{array}{c} \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \end{array} & \begin{array}{c} \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \end{array} \\ \begin{array}{c} \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \end{array} & \begin{array}{c} \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \end{array} & \begin{array}{c} \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \end{array} & \begin{array}{c} \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \\ \bullet \text{---} \bullet \end{array} \end{array} \right)$$



## The first case

- **Case 1: All paths  $P_i$  for  $i \in [k]$  are pairwise disjoint.**
- Under this condition, we take  $\sigma =$  composition  $P_1 \circ \dots \circ P_k$  with disjoint loops. That is:
  - (a) All sources and targets of  $P_i$  are disjoint,
  - (b) Sources and targets are disjoint to all intermediate nodes,
  - (c) All intermediate nodes of  $P_i$  are disjoint.

# The first case

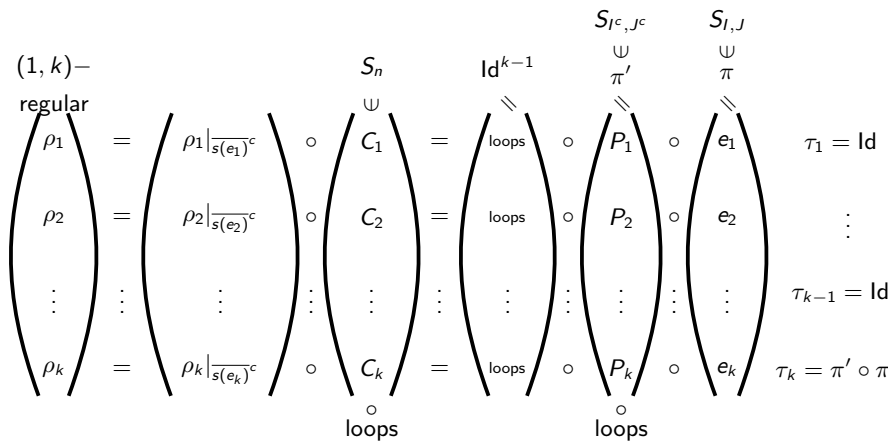


Figure: Case (1): Optimal  $(1, k)$ -regular multigraph  $F$  corresponds to an

## Example: Case 1

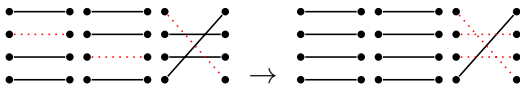
$$A = \begin{pmatrix} 0 & -1 & -5 & -4 \\ -6 & 0 & -2 & -1 \\ -3 & -4 & 0 & -3 \\ -2 & -7 & 0 & 0 \end{pmatrix}, \quad \text{adj}(A) = \begin{pmatrix} 0 & -1 & -2 & -2 \\ -3 & 0 & -1 & -1 \\ -3 & -4 & 0 & -3 \\ -2 & -3 & 0 & 0 \end{pmatrix}$$

Case 1 in the theorem:

$$\text{adj}(A)_{\{1\}^c, \{4\}^c}^3 = -2 = A_{4,1} = A_{\{4\}, \{1\}}^1,$$

# Example: Case 1

- **Left:**  $\text{adj}(A)_{\{2,3,4\},\{1,2,3\}}^3$  is the weight of  $F$  (an optimal  $(1, k)$ -regular multigraph).
- **Right:**  $\sigma$  is the (optimal) bijection:  $J^c = \{4\} \rightarrow I^c = \{1\}$ .  
 Joined with loops and the supervision edges, it makes a permutation.

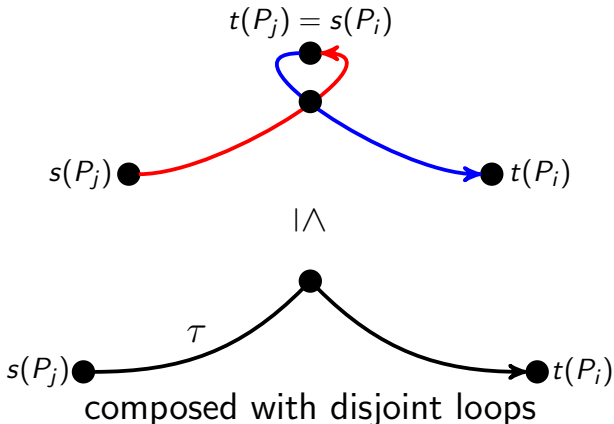


$$(2) (3) 4 \rightarrow 1$$

## Violation of a)

**Case 2a):** There exists a source which is also a target.

In this case  $\exists i, j \in [k] : t(P_j) = s(P_i)$  .



## Violation of a)

Construct  $F' = (\bigoplus_{i \in [k]} E_{\tau_i}, \pi')$  by:

- Replacing  $\rho_i, \rho_j \longrightarrow (\tau_i = \tau), (\tau_j = \text{Id})$ ,
- Keeping  $\tau_\ell = \rho_\ell$  for all  $\ell \neq i, j$ ,
- $\tilde{\pi}$  is formed from  $\pi$  by replacing  $(t(P_j), s(P_j)), (t(P_i), s(P_i)) \longrightarrow (t(P_i), s(P_j)), (t(P_j), s(P_i))$ .

## Example: Case 2a

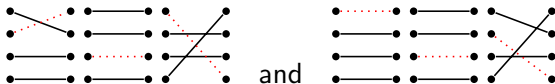
$$A = \begin{pmatrix} 0 & -1 & -5 & -4 \\ -6 & 0 & -2 & -1 \\ -3 & -4 & 0 & -3 \\ -2 & -7 & 0 & 0 \end{pmatrix}, \quad \text{adj}(A) = \begin{pmatrix} 0 & -1 & -2 & -2 \\ -3 & 0 & -1 & -1 \\ -3 & -4^\bullet & 0 & -3 \\ -2 & -3 & 0 & 0 \end{pmatrix}$$

Case 2a in the theorem For  $I = \{1, 2, 3\}$  and  $J = \{1, 3, 4\}$  we have

$$\text{adj}(A)_{J,I}^{\wedge 3} = -3^\bullet > A_{I^c, J^c}^{\wedge 1} = A_{4,2} = -7.$$

## Example: Case 2a

- **Left** is attained by two bijections in  $\text{adj}(A)$ :  
 $(3), 4 \rightarrow 1 \rightarrow 2$  and  $(1)(3), 4 \rightarrow 2$ .
- These bijections represent, in  $A$ , the following choices for 3 assignments with supervisions:



obtained by the same set of reorganized edges:

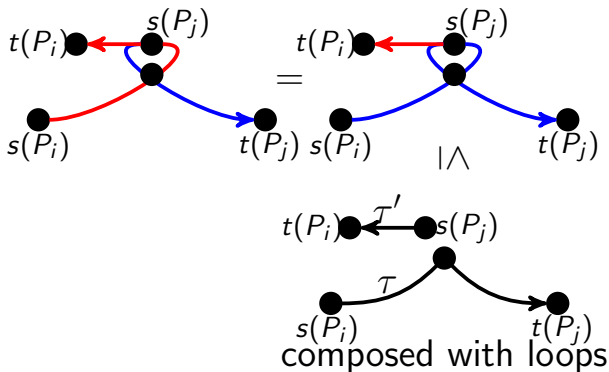
$$4 \rightarrow 1 \quad 1 \rightarrow 2 \quad \text{and} \quad 4 \rightarrow 1 \rightarrow 2$$



# Violation of b)

**Case 2b: There exists an intermediate node which is also a source or a target.**

Assume w.l.o.g. that Case 2a does not occur.



## Violation of b)

Construct  $F' = (\bigoplus_{i \in [k]} E_{\tau_i}, \pi')$  by:

- Replacing  $\rho_i, \rho_j \rightarrow (\tau_i = \tau), (\tau_j = \tau')$ ,
- Keeping  $\tau_\ell = \rho_\ell$  for all  $\ell \neq i, j$ ,
- $\tilde{\pi}$  is formed from  $\pi$  by replacing  $(t(P_j), s(P_j)), (t(P_i), s(P_i)) \rightarrow (t(P_i), s(P_j)), (t(P_j), s(P_i))$ .

## Example: Case 2b

$$A = \begin{pmatrix} 0 & -1 & -5 & -4 \\ -6 & 0 & -2 & -1 \\ -3 & -4 & 0 & -3 \\ -2 & -7 & 0 & 0 \end{pmatrix}, \quad \text{adj}(A) = \begin{pmatrix} 0 & -1 & -2 & -2 \\ -3 & 0 & -1 & -1 \\ -3 & -4^\bullet & 0 & -3 \\ -2 & -3 & 0 & 0 \end{pmatrix}$$

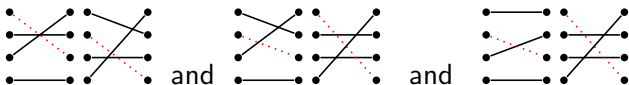
Case 2b in the theorem For  $I = \{1, 2\}$  and  $J = \{3, 4\}$  we have:

$$\text{adj}(A)_{J,I}^{\wedge 2} = -6^\bullet = (\text{adj}(A)_{3,1}\text{adj}(A)_{4,2}) \oplus (\text{adj}(A)_{3,2}\text{adj}(A)_{4,1}),$$

$$A_{I^c, J^c}^{\wedge 2} = -6 = A_{3,2}A_{4,1}.$$

## Example: Case 2b

- In this case  $\text{adj}(A)_{J,I}^{\wedge 2}$  is attained twice AND equality holds in the tropical Jacobi identity.
- There are three sets of 2 assignments obtaining the optimal base value:



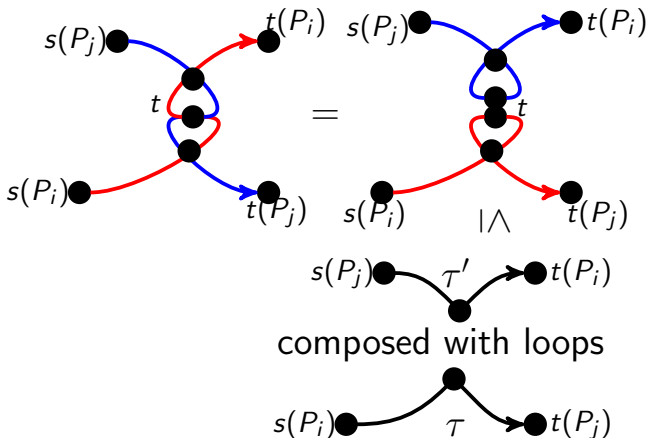
The first two are obtained by the same set of reorganized edges:

$$3 \rightarrow 1 \quad 4 \rightarrow 1 \rightarrow 2 \quad \text{and} \quad 4 \rightarrow 1 \quad 3 \rightarrow 1 \rightarrow 2$$

The third is case1 - disjoint paths:  $3 \rightarrow 2 \quad 4 \rightarrow 1$  obtaining  $A_{I^c, J^c}^{\wedge 2}$ .

# Violation of c)

**Case 2c):** There exists an intermediate node common to two paths. Assume w.l.o.g. that Cases 2a,2b do not occur.



## Violation of c)

Construct  $F' = (\bigoplus_{i \in [k]} E_{\tau_i}, \pi')$  by:

- Replacing  $\rho_i, \rho_j \rightarrow (\tau_i = \tau), (\tau_j = \tau')$ ,
- Keeping  $\tau_\ell = \rho_\ell$  for all  $\ell \neq i, j$ ,
- $\tilde{\pi}$  is formed from  $\pi$  by replacing  $(t(P_j), s(P_j)), (t(P_i), s(P_i)) \rightarrow (t(P_i), s(P_j)), (t(P_j), s(P_i))$ .

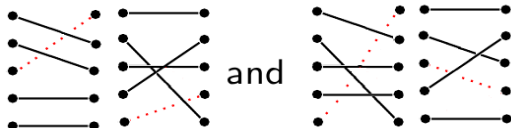
## Example: Case 2C

(4) (5)  $1 \rightarrow 2 \rightarrow 3$  , (1) (3)  $4 \rightarrow 2 \rightarrow 5$

with the bijection  $3 \rightarrow 1$  ,  $5 \rightarrow 4$ , becomes

(3) (4)  $1 \rightarrow 2 \rightarrow 5$  , (1) (5)  $4 \rightarrow 2 \rightarrow 3$

with the bijection  $5 \rightarrow 1$  ,  $3 \rightarrow 4$ :



# Supervised assignment optimization



Monday      Tuesday      Wednesday      Thursday

1. Work schedule
2. Lunch
3. Tips
4. Carpool
5. Inventory
6. Leftovers

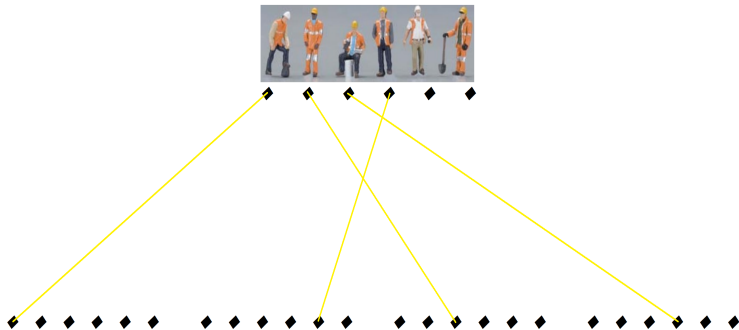


# Supervised assignment optimization



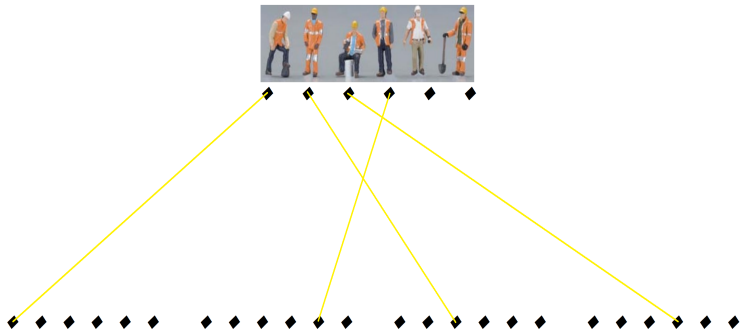
- Monday      Tuesday      Wednesday      Thursday
1. Work schedule (Monday)
  2. Lunch
  3. Tips (Wednesday)
  4. Carpool
  5. Inventory (Tuesday)
  6. Leftovers (Thursday)

# Supervised assignment optimization



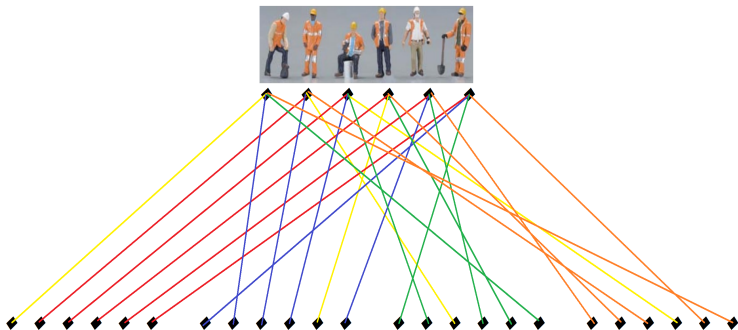
- Monday      Tuesday      Wednesday      Thursday
1. Work schedule (Monday)
  2. Lunch
  3. Tips (Wednesday)
  4. Carpool
  5. Inventory (Tuesday)
  6. Leftovers (Thursday)

# Supervised assignment optimization



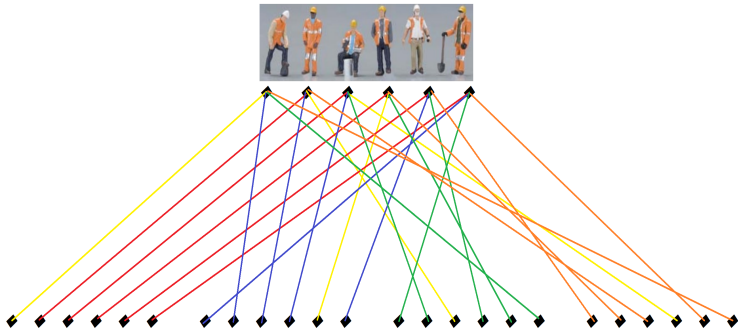
$$(1) 2 \rightarrow 3 \rightarrow 4 \rightarrow 5 \in S_{\{1,2,3,4\},\{1,3,4,5\}}$$

# Supervised assignment optimization



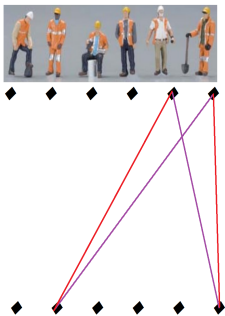
(1)(2)(3)(4)(5)(6) (1 2 3 4 5 6) (2 3)(4 5)(1 6) (2 3 4)(1 6 5)  
 $(1) 2 \rightarrow 3 \rightarrow 4 \rightarrow 5 \in \mathcal{S}_{\{1,2,3,4\},\{1,3,4,5\}}$

# Supervised assignment optimization



$(1)(2)(3)(4)(5)(6) \quad (1 \ 2 \ 3 \ 4 \ 5 \ 6) \quad (2 \ 3)(4 \ 5)(1 \ 6) \quad (2 \ 3 \ 4)(1 \ 6 \ 5)$   
 $(1) \ 2 \rightarrow 3 \rightarrow 4 \rightarrow 5 \in S_{\{1,2,3,4\},\{1,3,4,5\}}$   
 $(6) \ 5 \rightarrow 2 \text{ or } 5 \rightarrow 6 \rightarrow 2 \in S_{\{5,6\},\{2,6\}}$

# Supervised assignment optimization



$$(6) \ 5 \rightarrow 2 \text{ or } 5 \rightarrow 6 \rightarrow 2 \in S_{\{5,6\},\{2,6\}}$$

THANK YOU!