

MATH 380

Hemanshu Kaul

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## Vertex cover problem as an optimization problem

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For each  $v_i \in V(G)$ , we have to decide whether or not to include it in our set  $S$ , vertex cover.

$$\text{Let } x_i = \begin{cases} 1 & \text{if } v_i \in S \\ 0 & \text{if } v_i \notin S \end{cases}$$

Using the variables  $x_i$ , how will you model the requirement that each edge must be incident to at least one vertex in  $S$ ?



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$$x_i + x_j \geq 1 \quad \text{for all } v_i v_j \in E(G) \quad \begin{array}{c} \bullet \\ \text{---} \\ \bullet \\ v_i \quad v_j \end{array}$$

Objective function?



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$$\min \sum_{i=1}^n x_i$$

s.t.

$$x_i + x_j \geq 1 \quad \forall v_i, v_j \in E(G)$$

$$x_i \in \{0, 1\} \quad \forall i = 1, \dots, n$$

[minimize total # vertices picked]

[each edge is "covered" by at least one vertex]



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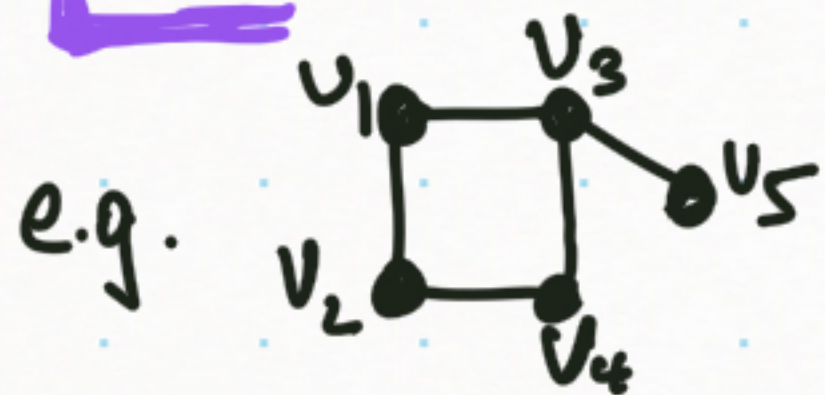
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$$\min x_1 + x_2 + x_3 + x_4 + x_5 \quad \text{s.t.}$$

$$\begin{array}{l|l|l} x_1 + x_2 \geq 1 & x_2 + x_4 \geq 1 & x_3 + x_5 \geq 1 \\ x_1 + x_3 \geq 1 & x_3 + x_4 \geq 1 & \\ \hline x_1, x_2, x_3, x_4, x_5 \in \{0, 1\} \end{array}$$



Most Binary Optimization Problems are very hard to solve.

We use different ideas to get approximate / non optimal solutions

① Heuristics based on greedy algorithms / local search.

② Randomized Rounding schemes.

Consider

$$\begin{aligned} \min \quad & \vec{c}^T \vec{x} \\ \text{s.t.} \quad & A\vec{x} \leq \vec{b} \end{aligned}$$

$$x_1, x_2, \dots, x_n \in \{0, 1\}$$

LP relaxation  $\rightarrow$

$$\begin{aligned} \min \quad & \vec{c}^T \vec{y} \\ \text{s.t.} \quad & A\vec{y} \leq \vec{b} \\ & y_1, \dots, y_n \in [0, 1] \end{aligned}$$



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"round"  
 $\leftarrow$  each  $y_i \in [0, 1]$   
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$$\text{e.g. } x_i = \begin{cases} 1 & \text{if } y_i \geq 0.5 \\ 0 & \text{if } y_i < 0.5 \end{cases}$$



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Issues

- may not be satisfied
- this solution might be far from optimal solution.

whether  
 Check if these variables  
 satisfy  $A\vec{x} \leq \vec{b}$



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Flip a coin with  $P[\text{heads}] = y_i$   
 $P[\text{tails}] = 1 - y_i$

$$x_i = \begin{cases} 1 & \text{with probab. } y_i \\ 0 & \text{with probab. } 1 - y_i \end{cases}$$



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Compute the probability of "failure" & Expected value of objective function

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Analyze this to guarantee a good / close to optimal solution with high probability

Compute the probability of "failure" & Expected value of objective function

(Monte Carlo) Randomized Algorithm:



## Assigning jobs to qualified applicants / processors / machines

Let  $a_1, a_2, \dots, a_n$  be jobs to be carried out.

Let  $b_1, b_2, \dots, b_m$  be a pool <sup>available</sup> of applicants / processors / machines.

Each  $b_j$  is capable of carrying out only certain jobs  $a_i$ .

We have to assign each job to one applicant qualified to do it.

How can we maximize the number of assigned jobs?



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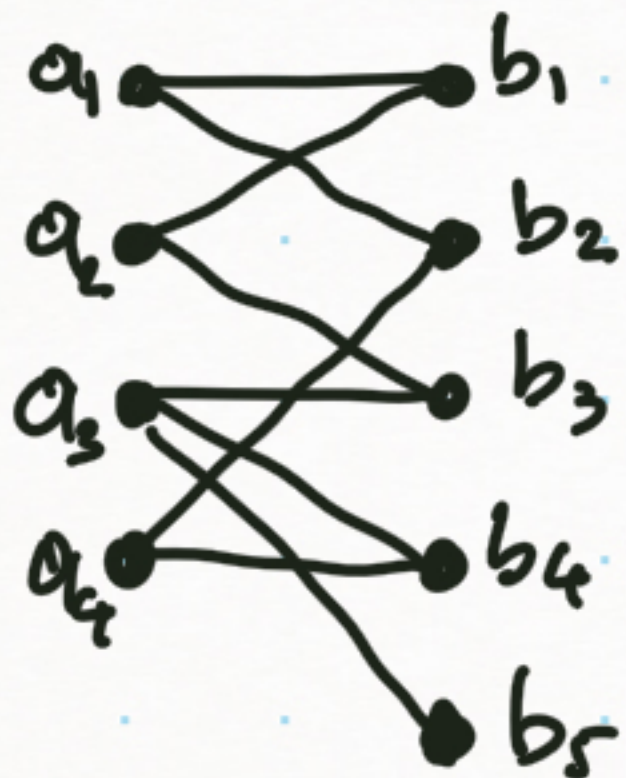
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A graph whose vertices can be partitioned into two disjoint subsets such that all edges only go in between the two parts, is called a bipartite graph.



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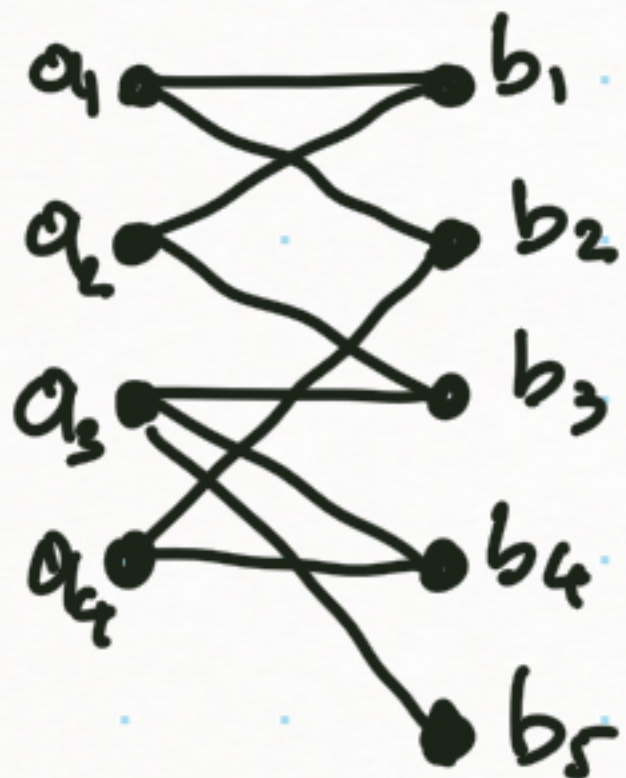
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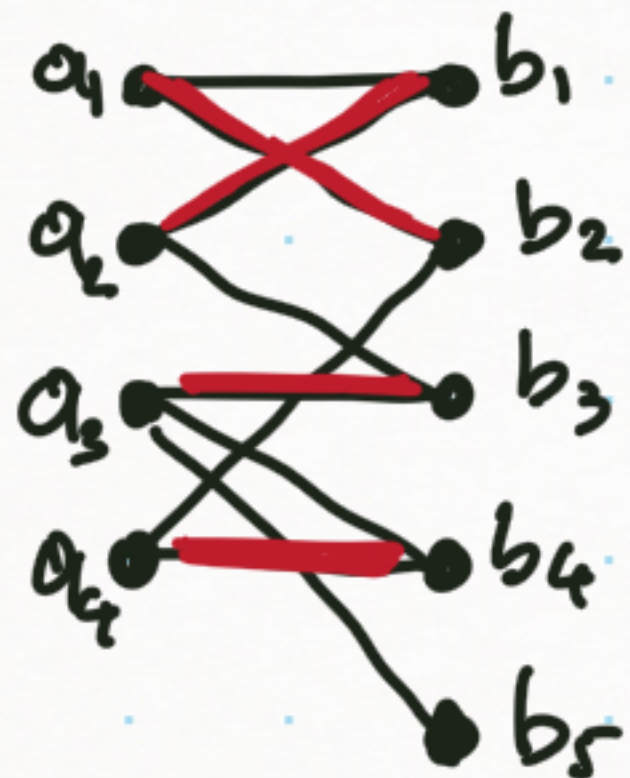
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$a_1$  assigned to  $b_2$

$a_2$  — " —  $b_1$

$a_3$  — " —  $b_3$

$a_4$  — " —  $b_4$

A matching is subset of  $E(G)$

such that all edges in it are vertex disjoint.

Maximum matching:  $\max |M|$   
M matching



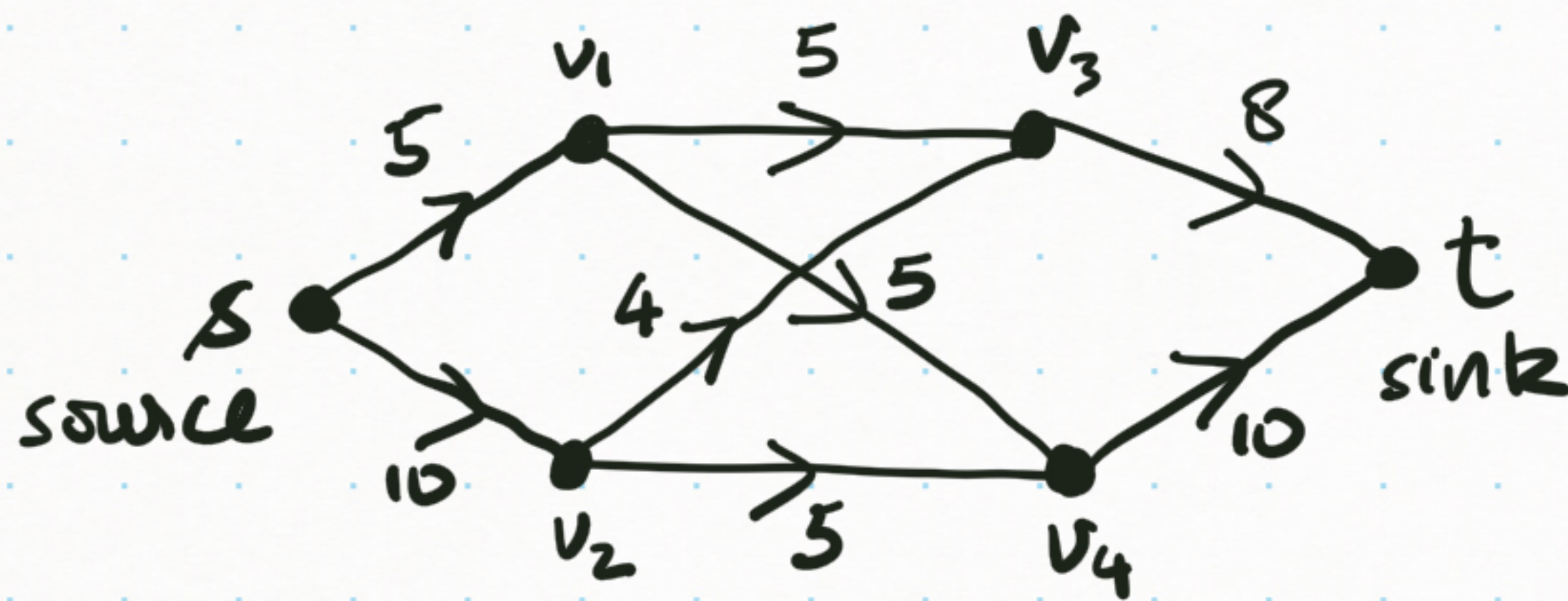
Maximum Matching problem in a bipartite graph can be modeled as a more general network optimization problem called the

## Maximum Flow Problem

Given a network  $G = (N, A)$

two special nodes  $s = \text{source}$  &  $t = \text{sink}$

capacity on each arc,  $u_{ij} \geq 0$  for each arc  $(v_i, v_j) \in A$



source  $\equiv$  factory

sink  $\equiv$  retail store

network  $\equiv$  transportation network

capacity  $\equiv$  how much can be transported using that link

$v_1, v_2, v_3, v_4 \equiv$  warehouses /

transshipment nodes.

Aim maximize the amount of flow from  $s$  to  $t$  while respecting the capacities

Flow  $\equiv$  quantity of goods transported across each arc.



# Max Flow Problem

Decision variables

$x_{ij}$  = amount of flow from  $v_i$  to  $v_j$   
for each arc  $(v_i, v_j)$

$$\max \sum_{i \in N^+(s)} x_{si}$$

such that

$$\sum_{v_k \in N^-(v_i)} x_{ki} - \sum_{v_j \in N^+(v_i)} x_{ij} = 0 \quad \forall v_i \in N$$

$$x_{ij} \leq u_{ij} \quad \forall (v_i, v_j) \in A$$

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$N^+(s)$   
out-neighborhood of  $s$

[total flow out of node  $s$ ]

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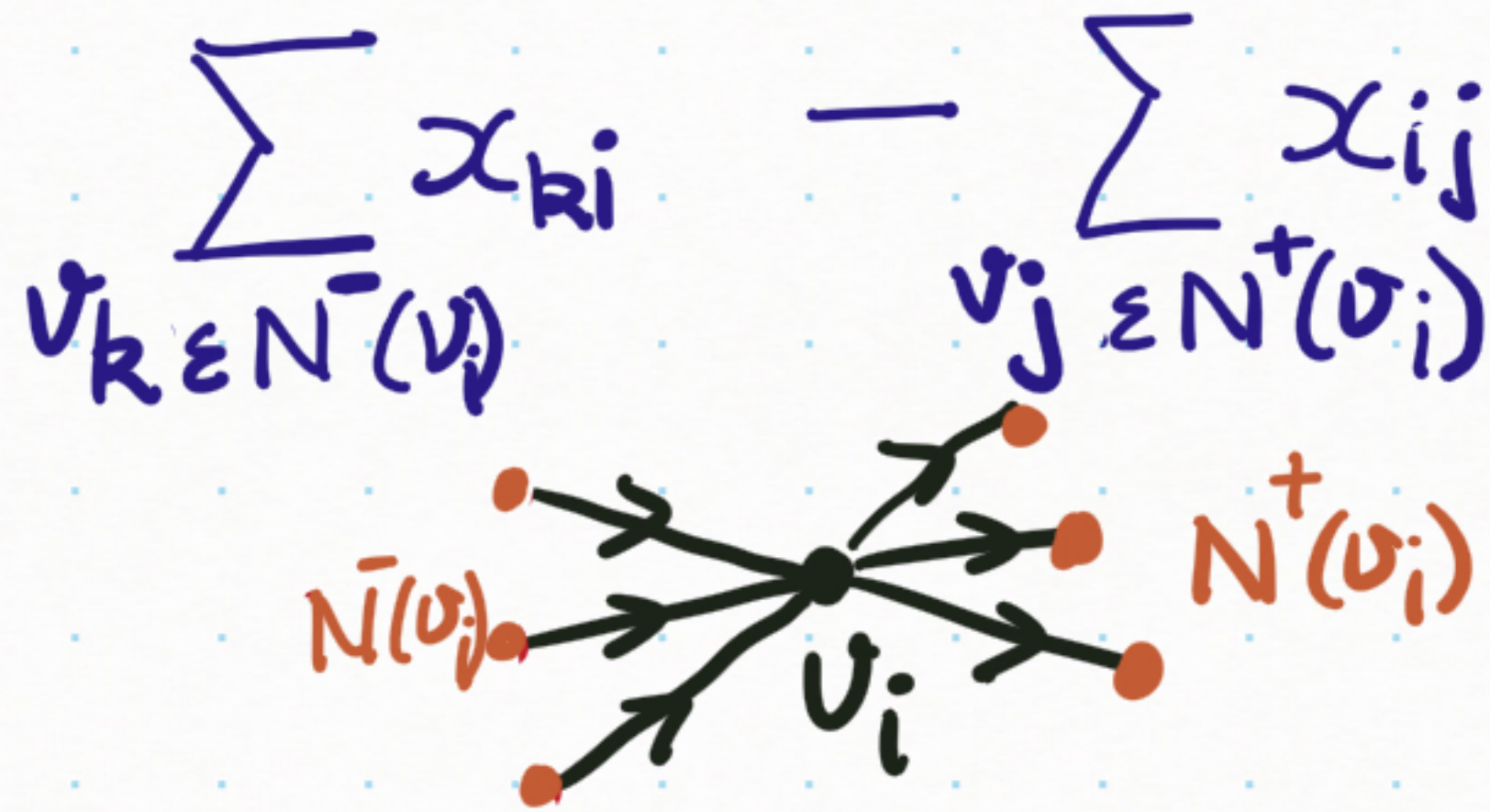
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[in-flow = out-flow]  
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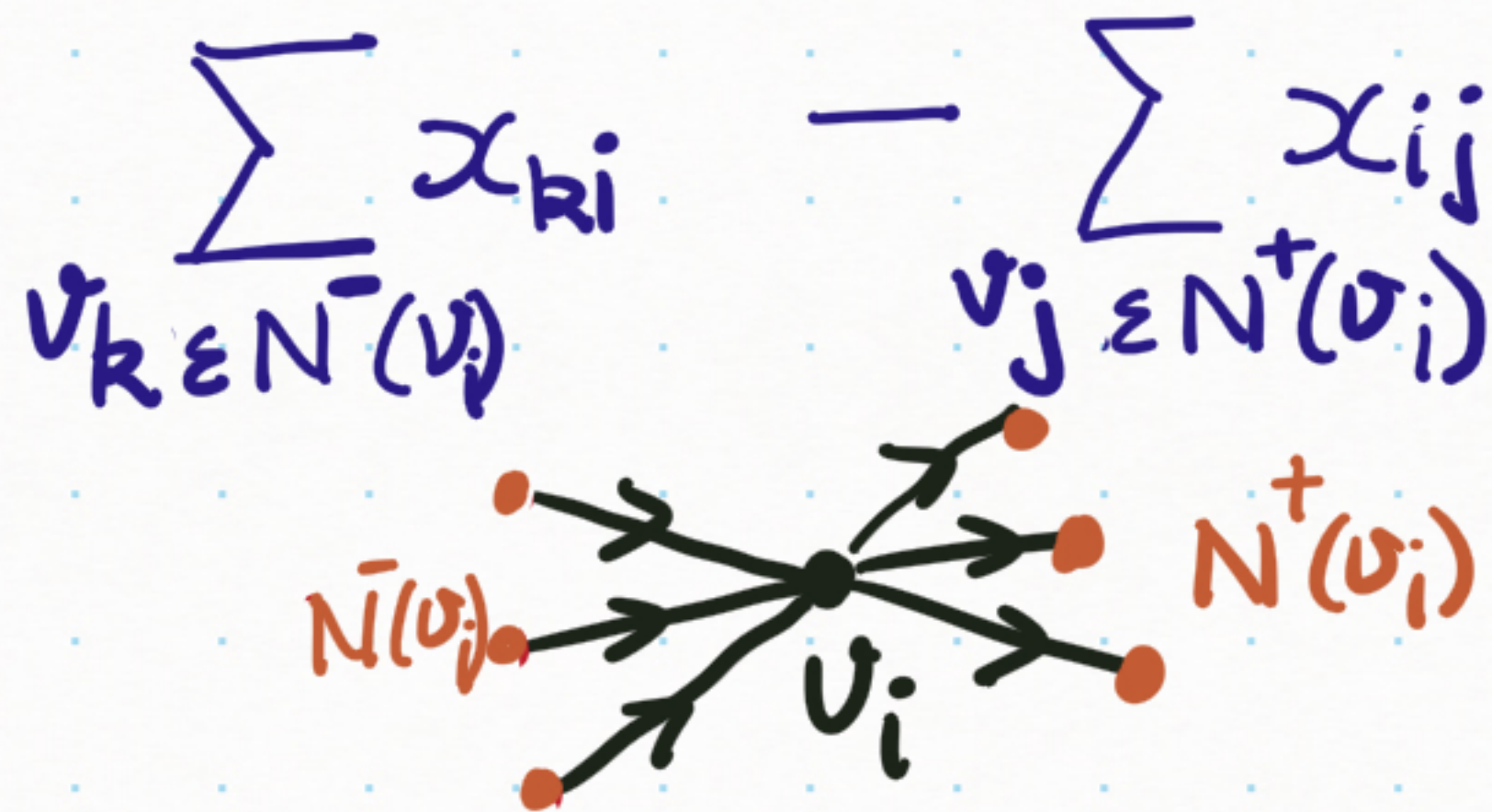
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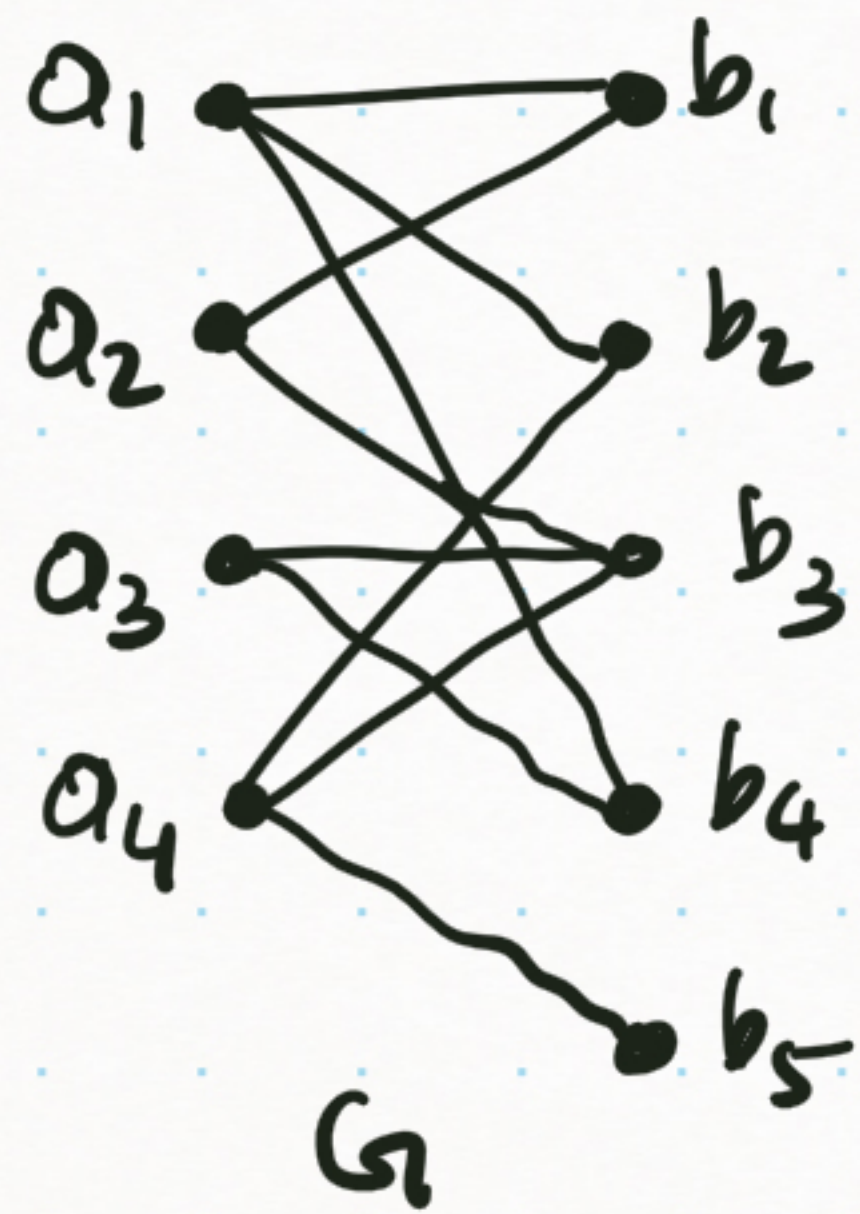
[in-flow = out-flow]  
[at each node  $v_i$ ]

$$x_{ij} \leq u_{ij} \quad \forall (v_i, v_j) \in A \quad \text{[capacity on each arc } v_i v_j]$$

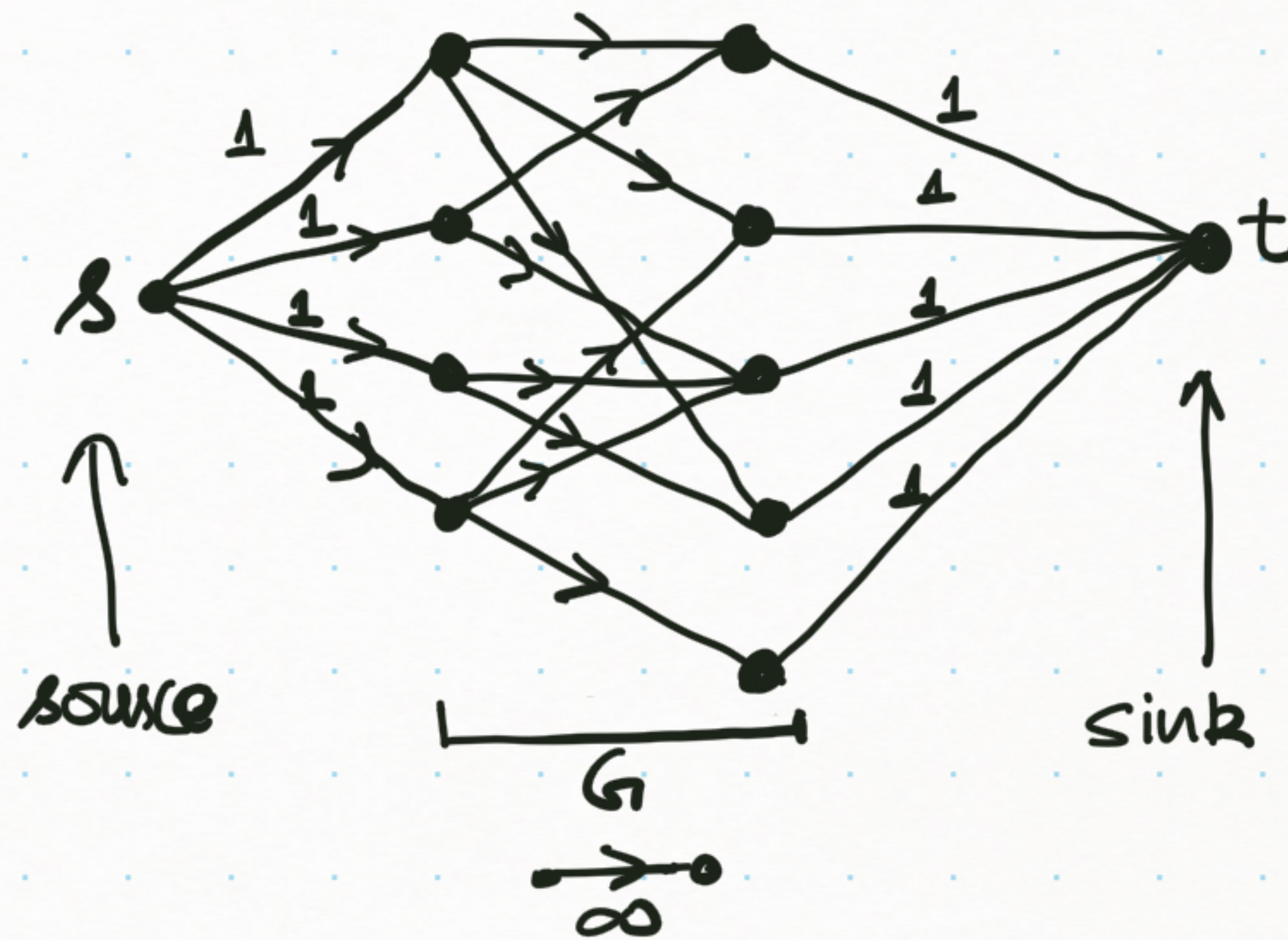
$$x_{ij} \geq 0 \quad \forall (v_i, v_j) \in A \quad \text{[non-negativity]}$$



Finding max matching on



is the same as  
finding the  
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## Modeling Logistics problem on a transportation network

We have a transportation network connecting factories, warehouses and retail stores. Each factory can supply certain amount of goods and Each store has a demand for a certain amount of goods. How can we satisfy demand while minimizing transportation costs?



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Given a network  $G = (N, A)$

for each arc  $(v_i, v_j)$ , we are given capacity bounds  
$$l_{ij} \leq x_{ij} \leq u_{ij}$$

and there is cost per unit amount of transportation on that arc  $c_{ij}$

for each node  $v_i$ , there is supply/demand  $b(i)$  associated with it  
[  $b(i) \geq 0$  means supply from factory  
 $b(i) < 0$  means demand from store  
 $b(i) = 0$  means transshipment point ]



Decision variables

$x_{ij}$  = flow over the arc  $(v_i, v_j)$   
for all arcs  $(v_i, v_j) \in A$ .



$$\min \sum_{(v_i, v_j) \in A} c_{ij} x_{ij}$$

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$$\sum_{v_j \in N^+(v_i)} x_{ij} - \sum_{v_k \in N^-(v_i)} x_{ki} = b(i) \quad \text{for all } v_i \in N$$

Transportation Problem



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[Total cost of flow]

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[capacity bound on  
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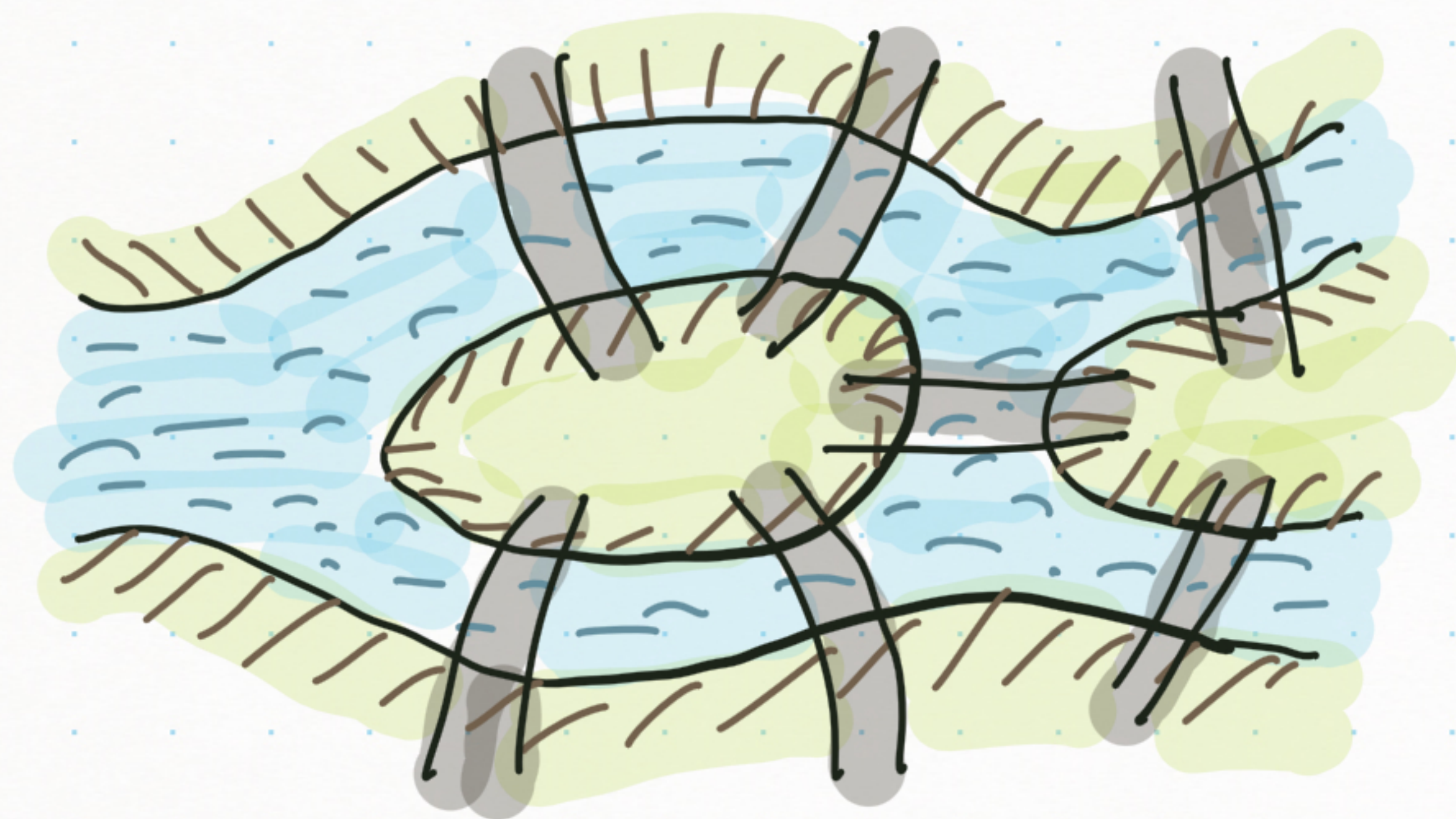
[Outflow - Inflow = "supply" at node  $v_i$ ]

Transportation Problem



Read

The Seven Bridges of Königsberg,  
Euler's problem, and  
how to solve this topological problem  
using graph theory in section 8.1.



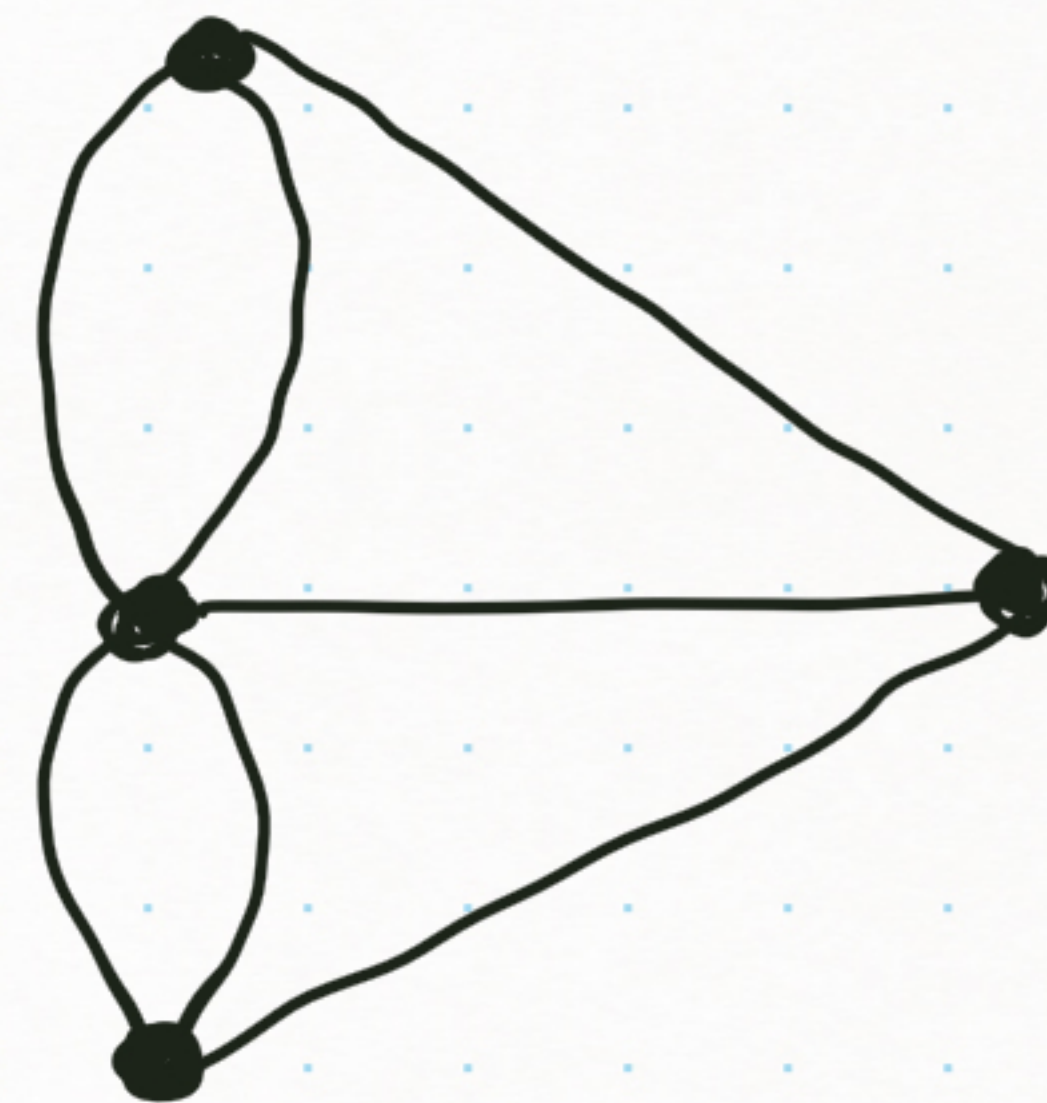
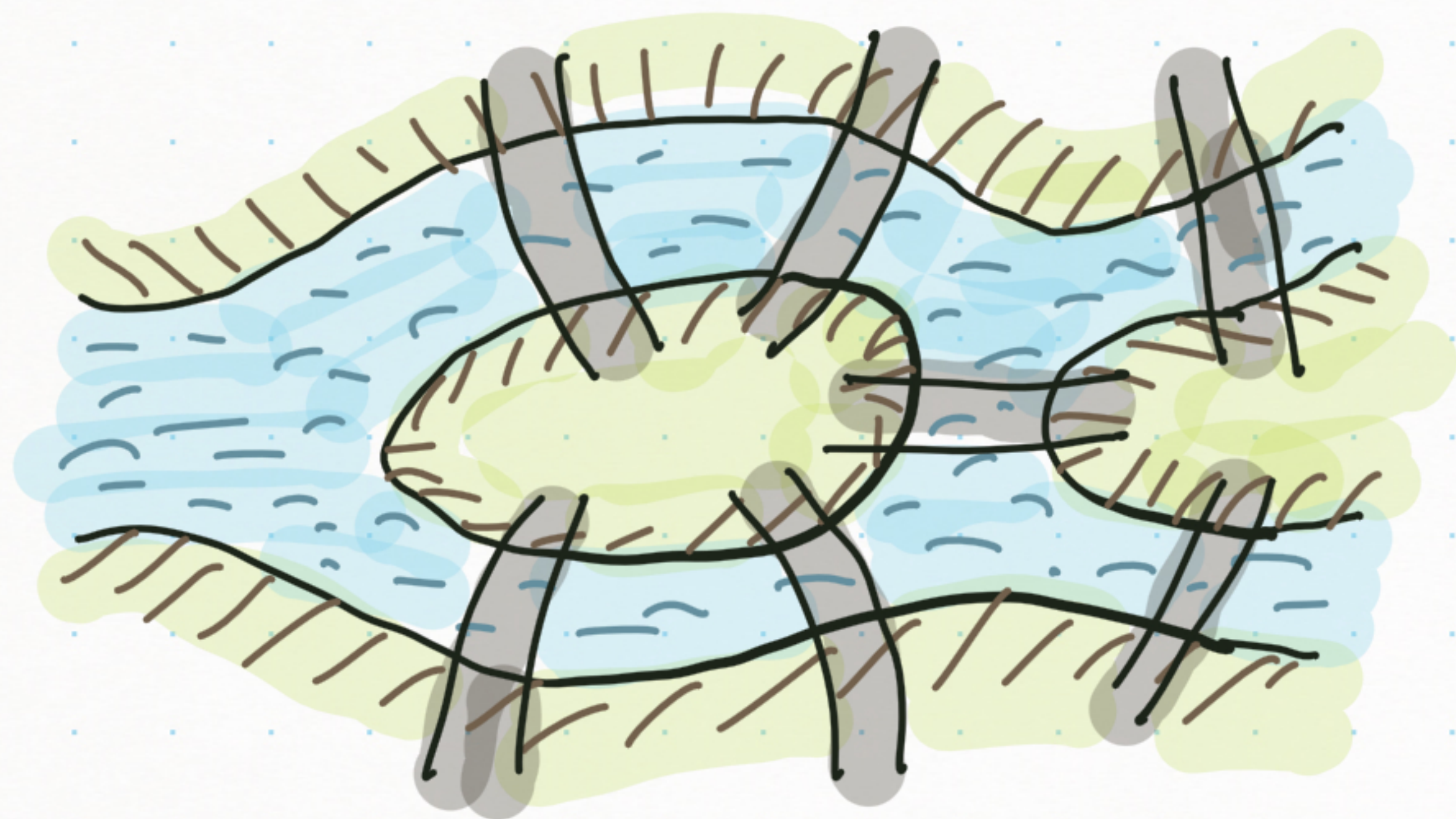
Start & end a walk at the same spot  
and cross each of the 7 bridges exactly  
once during the walk.

← Can this be done?



Read

The Seven Bridges of Königsberg,  
Euler's problem, and  
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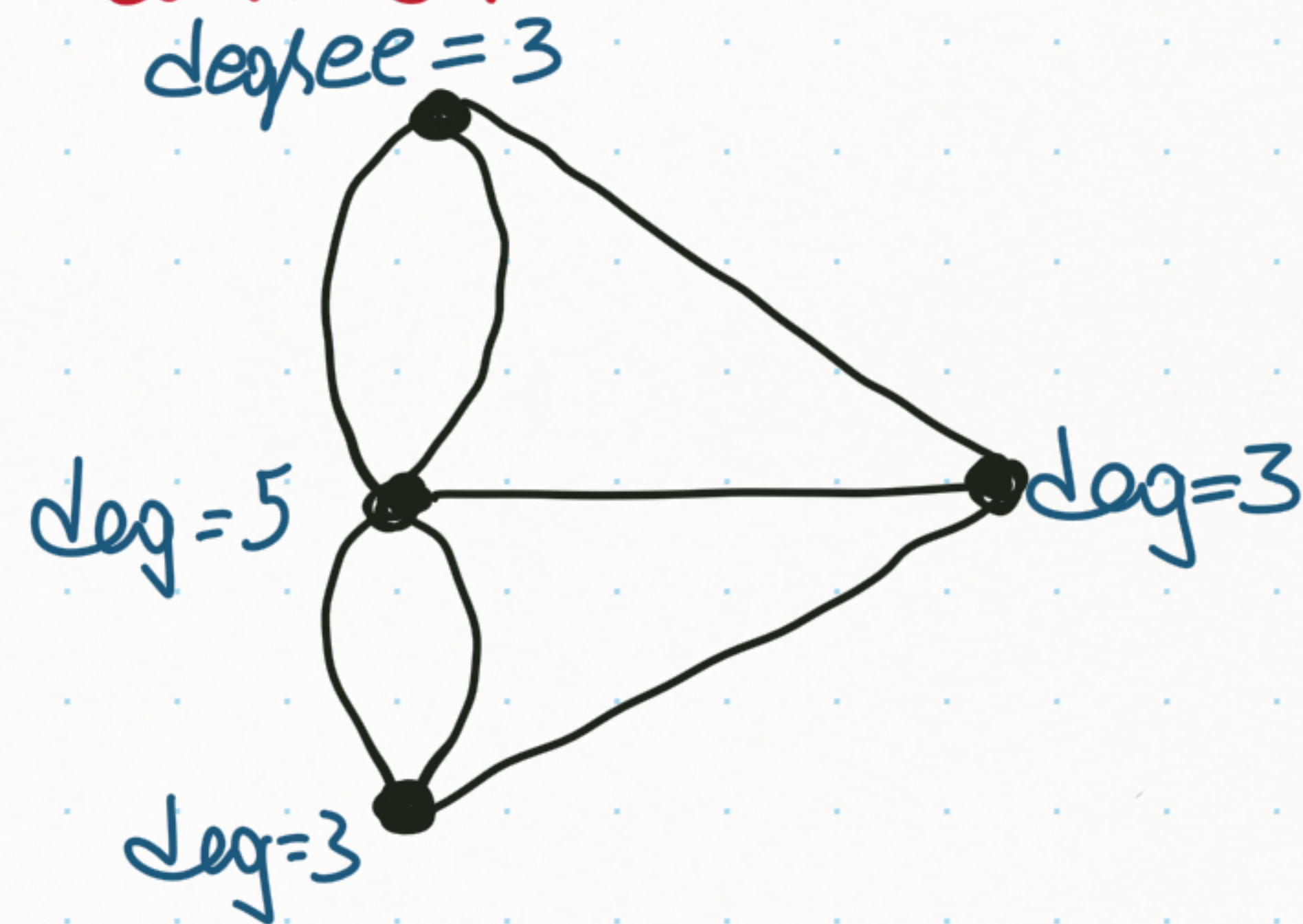
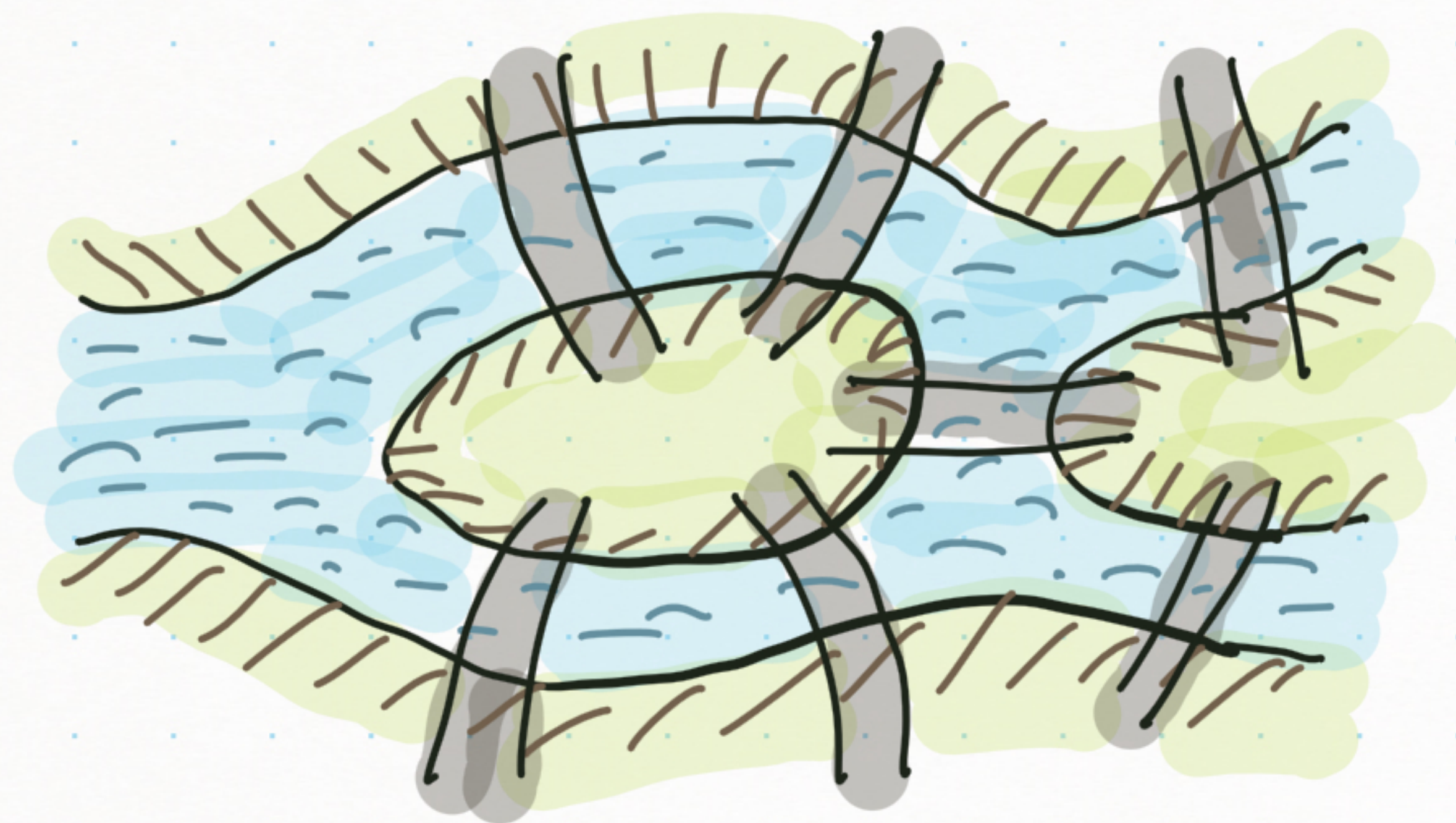
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