

# Data-driven Decision Making

Graph problems: partitioning, stable sets, cliques

João Pedro Pedroso

**Kseniia Klimetnova**

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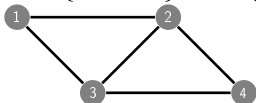
- Location problems:
  - k-Median
  - k-Center
  - k-Cover
- Graph problems
  - coloring
- Models:
  - how to formulate
  - how to improve the formulation

- **Undirected graphs:** lines connecting two vertices have no implied direction

- lines with no direction  $\rightarrow$  **edges**

- example:  $G = (V, E)$

$$V = \{1, 2, 3, 4\}, E = \{\{1, 2\}, \{2, 3\}, \{2, 4\}, \{3, 1\}, \{3, 4\}\}$$

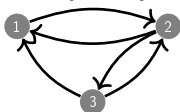


- **Directed graphs:** direction of lines connecting two vertices is important

- directed lines  $\rightarrow$  **arcs**

- example:  $G = (V, A)$

$$V = \{1, 2, 3\}, A = \{(1, 2), (2, 1), (2, 3), (3, 2), (3, 1)\}$$

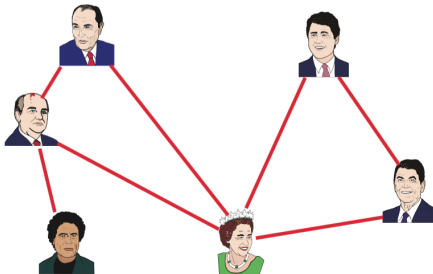


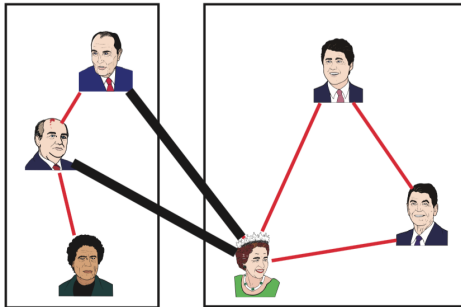
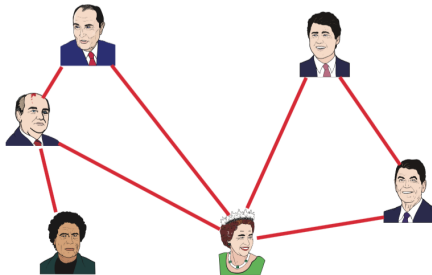
# Graph partitioning problem

# Graph partitioning problem

*Six friends are deciding how to split for forming two teams of mini-soccer. Of course, in order to be fair, each team must have the same number of persons — three, in this case. However, having good friends in separate teams should be avoided as much as possible. So, how should the group of persons be divided into two teams?*

- Graph:
  - nodes  $\rightarrow$  persons
  - edge between two persons  $\rightarrow$  good friends





- Pairs of good friends belonging to different teams  $\rightarrow$  thick line

# Graph partitioning problem

*Given an undirected graph  $G = (V, E)$  with an even number of vertices  $n = |V|$ , divide  $V$  into two subsets  $L$  and  $R$  with the same number of vertices, satisfying the conditions below, so as to minimize the number of edges across  $L$  and  $R$  (NP-hard)*

- $L \cap R = \emptyset \rightarrow$  no intersection
- $L \cup R = V \rightarrow$  union is the whole set of vertices
- $|L| = |R| = n/2$
- Applications:
  - VLSI (very-large-scale integration) design
  - network immunization: pick out a set of nodes to immunize so that network is infected as little as possible by an epidemic
  - ...
- Subsets with same cardinality
  - uniform partition or equipartition
  - (cardinality of set  $|V|$ : its number of elements)
- Edges across  $L$  and  $R$ :
  - $\{i, j\}$  such that either  $i \in L$  and  $j \in R$ , or  $i \in R$  and  $j \in L$

# Formulation as an integer optimization problem

- (Binary) Variables: for each vertex  $i \in V$ :
  - $x_i = 1$  if  $i$  is included in subset  $L$
  - $x_i = 0$  if  $i$  is included in subset  $R$
- Vertices equally divided  $\rightarrow$  sum  $x_i$  must be equal to  $n/2$
- When an edge  $\{i, j\}$  is across  $L$  and  $R$ : either
  - $x_i(1 - x_j) = 1$ , or
  - $(1 - x_i)x_j = 1$

$$\text{minimize} \quad \sum_{\{i,j\} \in E} (x_i(1 - x_j) + (1 - x_i)x_j)$$

$$\text{subject to} \quad \sum_{i \in V} x_i = n/2$$

$$x_i \in \{0, 1\}$$

$$\forall i \in V$$

- previous model: **quadratic expressions** in objective
- many solvers do not support minimization problems whose objective function is not convex
- quadratic terms are **not convex**
- even if some solvers provide support for these cases, they usually are much more efficient for solving linear problems
- if we can find an **equivalent linear formulation**, is it advisable to use it

# Reformulation

- (Binary) variables  $y_{ij}$  model the case where **edges are incident to different subsets**:
  - $y_{ij} = 1$  if the endpoints of edge  $\{i, j\}$  are across  $L$  and  $R$
  - $y_{ij} = 0$  otherwise
- $x_i, i \in V \rightarrow$  as in previous formulation

$$\begin{aligned} & \text{minimize} && \sum_{\{i,j\} \in E} y_{ij} \\ & \text{subject to} && \sum_{i \in V} x_i = n/2 \\ & && x_i - x_j \leq y_{ij} && \forall \{i, j\} \in E \\ & && x_j - x_i \leq y_{ij} && \forall \{i, j\} \in E \\ & && x_i \in \{0, 1\} && \forall i \in V \\ & && y_{ij} \in \{0, 1\} && \forall \{i, j\} \in E \end{aligned}$$

- Objective: minimize sum of variables  $y_{ij}$ 
  - their value will tend to be zero, but constraints force some of them to be one
- First constraint: equal division of the set of vertices
- Second constraint: if  $i \in L$  and  $j \notin L$  then  $y_{ij} = 1$ 
  - *i.e.*, edge  $\{i, j\}$  is across subsets  $L$  and  $R$
- Third constraint: if  $j \in L$  and  $i \notin L$ , then  $y_{ij} = 1$ .

$$x_i - x_j \leq y_{ij} \qquad \forall \{i, j\} \in E$$

$$x_j - x_i \leq y_{ij} \qquad \forall \{i, j\} \in E$$

# AMPL model

---

```
set V;  
set E within {V,V};  
param n := card(V);  
  
var x {V} binary;  
var y {E} binary;  
  
minimize z: sum {(i,j) in E} y[i,j];  
  
subject to  
Equipart: sum {i in V} x[i] = n/2;  
AcrossLR {(i,j) in E}: x[i] - x[j] <= y[i,j];  
AcrossRL {(i,j) in E}: x[j] - x[i] <= y[i,j];
```

---

# Toy instance

```
data;  
set V := 1 2 3 4 5 6;  
set E :=  
  (1,2)  
  (2,3)  
  (2,6)  
  (3,6)  
  (4,5)  
  (5,6)  
  (4,6)  
;
```

## Create a random graph

---

```
import random
def make_data(n,prob):
    """make_data: prepare data for a random graph
    Parameters:
        - n: number of vertices
        - prob: probability of existence of an edge, for each pair of vertices
    Returns a tuple with a list of vertices and a list edges.
    """
    V = list(range(1,n+1))
    E = [(i,j) for i in V for j in V if i < j and random.random() < prob]
    return V,E
```

---

# AMPL + Python usage

```
V, E = make_data(10, .5)

from amply import AMPL, Environment, DataFrame
ampl = AMPL()
ampl.option['solver'] = 'gurobi'
ampl.read("gpp.mod")
ampl.set['V'] = V
ampl.set['E'] = E

ampl.solve()
z = ampl.obj['z']
print("Optimum:", z.value())

x = ampl.var['x']
L = [i for i in V if x[i].value() > .5]
R = [i for i in V if x[i].value() < .5]
print("L =", L)
print("R =", R)

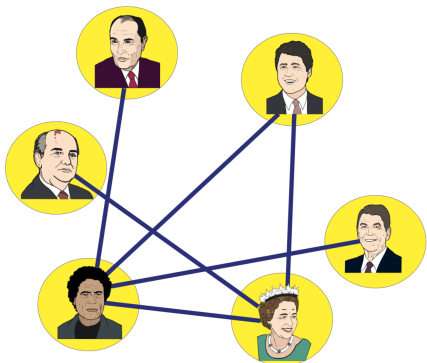
y = ampl.var['y']
across = [(i,j) for (i,j) in E if y[i,j].value() > .5]
print("Across edges:", across)
```

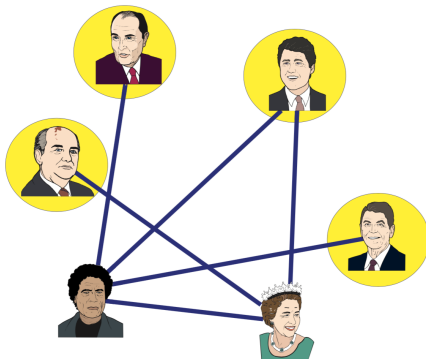
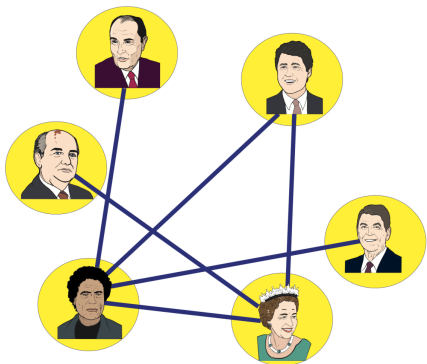
# Maximum stable set and maximum clique problems

# Maximum stable set problem

*You are choosing, from a group of six friends, with whom to go for a picnic. However, persons linked with an edge in the figure are on very unfriendly terms with each other, so if both of them go to the picnic, it will be spoiled. To have as many friends as possible in the picnic, who should be invited?*

- Graph:
  - nodes  $\rightarrow$  persons
  - edge between two persons  $\rightarrow$  **bad terms**





- nodes encircled → persons who can all be at the picnic without spoiling it

# Maximum stable set problem

*Given an undirected graph  $G = (V, E)$ , a subset  $S \subseteq V$  is called a **stable set** when there isn't any edge among vertices of  $S$ . The problem is to find a stable set  $S$  such that its cardinality is maximum.*

- Cardinality of a set  $|S|$ : the number of elements it contains

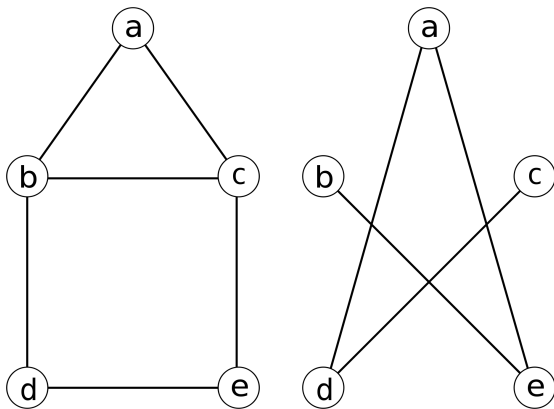
# Maximum clique problem

*Given an undirected graph  $G = (V, E)$ , a subset  $C \subseteq V$  is called a **clique** when the subgraph induced by  $C$  is complete. The problem is to find a clique  $C$  which maximizes cardinality  $|C|$ .*

- **Complete graph**: there is edge connecting all pairs of vertices
- Clique: subgraph **induced** by a subset of vertices containing all the edges of the original graph with both ends in that subset
- Applications: coding theory, reliability, genetics, archeology and VLSI design, ...

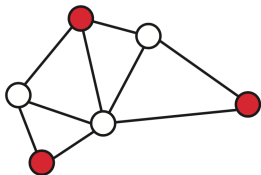
# Complementary graph

- Given a graph, its **complementary graph** inverts the edges:
  - contains edges only between pairs of vertices for which there is **no edge** in the original graph

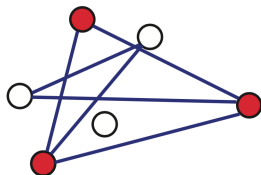


# Maximum stable set and Maximum clique

- **Maximum clique problem:** equivalent to maximum stable set:
  - can be converted through a simple transformation, and
  - solution is the same



(a)



(b)

- Red vertices are both:
  - Maximum stable set on the original graph
  - Maximum clique on the complementary graph

# Maximum stable set: formulation in integer optimization

- (Binary) Variables: for each vertex  $i \in V$ :
  - $x_i = 1$  if  $i$  is included in the stable set
  - $x_i = 0$  otherwise
- Model:

$$\begin{array}{ll} \text{maximize} & \sum_{i \in V} x_i \\ \text{subject to} & x_i + x_j \leq 1 \quad \forall \{i, j\} \in E \\ & x_i \in \{0, 1\} \quad \forall i \in V \end{array}$$

# AMPL model

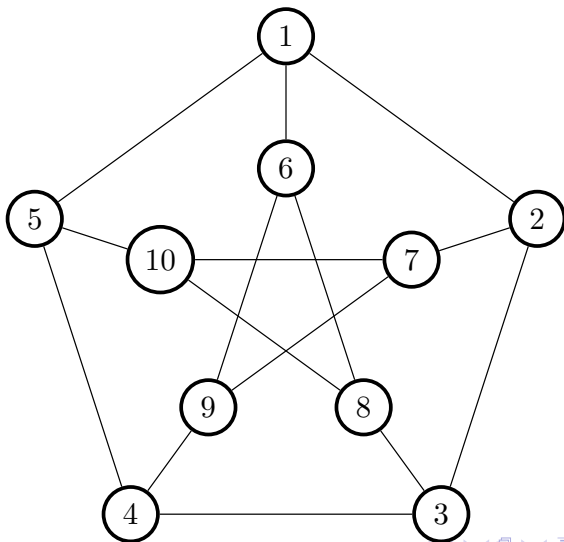
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```
set V;  
set E within {V,V};  
  
var x {V} binary;  
  
maximize z: sum {i in V} x[i];  
  
subject to  
Edge {(i,j) in E}: x[i] + x[j] <= 1;
```

---

# Challenge:

Find the maximum clique for the Petersen graph:



# Bonus: Routing problems: the (Asymmetric) TSP

# Miller-Tucker-Zemlin (potential) formulation

- Consider an **asymmetric traveling salesman problem**
- We will now see a formulation with a **number of constraints of polynomial order**

Given:

- **directed graph**  $G = (V, A)$ , where
  - $V \rightarrow$  set of vertices
  - $A \rightarrow$  a set of (directed) arcs
- **function**  $c : A \rightarrow \mathbb{R}$  associating a distance to each arc

**Find a tour which passes exactly once in each city and minimizes the total distance**

- Variables:
  - (binary)  $x_{ij} = 1$  if visiting vertex  $j$  next to  $i$ , 0 otherwise
  - (real)  $u_i \rightarrow$  represents the visiting order of vertex  $i$
- Interpretation:  $u_1$  is the **potential** at each vertex
  - starting point:  $u_1 = 0$
- When visiting  $j$  next to vertex  $i$ :
  - force the potential of  $j$  to be  $u_j = u_i + 1$
  - do this for any vertex except 1
  - possible values for  $u_i$  are  $1, 2, \dots, n - 1$ 
    - $n \rightarrow$  number of vertices

# MTZ formulation for the asymmetric traveling salesman problem

$$\begin{aligned} & \text{minimize} && \sum_{i \neq j} c_{ij} x_{ij} \\ & \text{subject to} && \sum_{j:j \neq i} x_{ij} = 1 && i = 1, \dots, n \\ & && \sum_{j:j \neq i} x_{ji} = 1 && i = 1, \dots, n \\ & && u_i + 1 - (n - 1)(1 - x_{ij}) \leq u_j && i = 1, \dots, n, \\ & && && j = 2, \dots, n : i \neq j \\ & && 0 \leq u_i \leq n - 1 && i = 1, \dots, n \\ & && x_{ij} \in \{0, 1\} && \forall i \neq j \end{aligned}$$

- First and second constraints: assignment constraints  $\rightarrow$  ensure that each vertex is incident to one outgoing arc and one incoming arc.
- Third constraint: Miller-Tucker-Zemlin or **potential** constraint  $\rightarrow$  define the order in which each vertex  $i$  is visited on a tour.
  - $u_i \rightarrow$  potential at vertex  $i$
  - forcing  $u_j = u_i + 1$  only when  $x_{ij} = 1$  in
$$u_i + 1 - (n - 1)(1 - x_{ij}) \leq u_j$$
    - coefficient  $n - 1$  of term  $(1 - x_{ij}) \rightarrow$  "Big M" constraint
- Fourth constraint: upper and lower bounds of the potential.
- This formulation is **much weaker** than subtour elimination constraints

---

```
param n;  
set V := {1..n};  
param c {i in V, j in V};  
  
var x {i in V, j in V: i != j} binary;  
var u {V} >= 0, <= n-1;  
  
minimize z: sum {i in V, j in V: i != j} c[i,j] * x[i,j];  
  
subject to  
OutDegree {i in V}:  
    sum {j in V: j != i} x[i,j] = 1;  
InDegree {i in V}:  
    sum {j in V: j != i} x[j,i] = 1;  
Potential {i in 1..n, j in 2..n : i != j}:  
    u[i] + 1 - (n-1) * (1-x[i,j]) <= u[j];
```

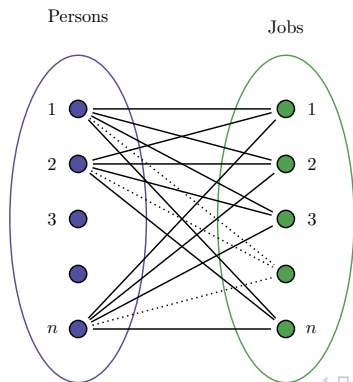
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# Preliminaries for next week:

Assignment problem

# Assignment problem

- There are  $n$  people available to carry out  $n$  jobs
- Each person is assigned to carry out exactly one job
- Some persons are better suited to particular jobs than others
  - cost  $c_{ij}$  if person  $i$  is assigned to job  $j$
- Problem: find a minimum cost / maximum reward assignment



# Assignment problem

Formulation:

- $x_{ij} = 1$  if person  $i$  does job  $j$ , 0 otherwise
- all persons are assigned to a job
- all jobs are assigned to a person

$$\text{minimize } \sum_{i=1}^n \sum_{j=1}^n c_{ij} x_{ij}$$

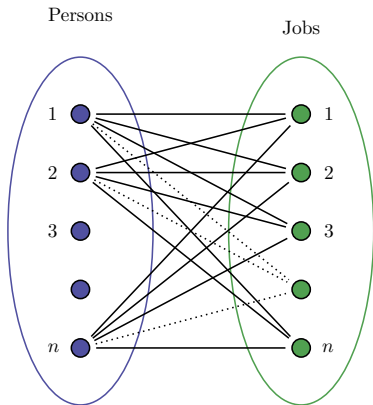
$$\text{subject to } \sum_{j=1}^n x_{ij} = 1 \quad i = 1, \dots, n$$

$$\sum_{i=1}^n x_{ij} = 1 \quad j = 1, \dots, n$$

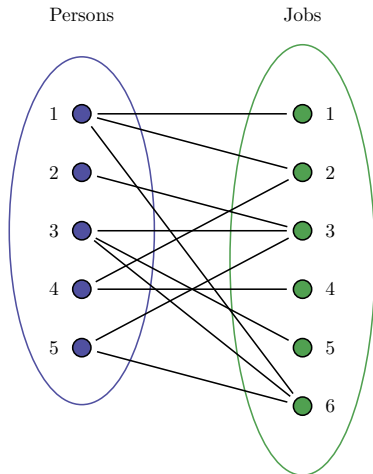
$$x_{ij} \in \{0, 1\} \quad i = 1, \dots, n, j = 1, \dots, n$$

# Bipartite graphs

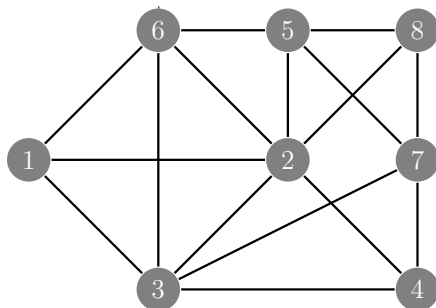
## Complete



## Arbitrary

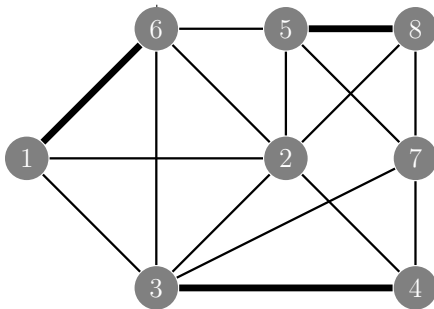


- Given a graph  $G = (V, E)$  a **matching** is a set of disjoint edges



# Matching

- Given a graph  $G = (V, E)$  a **matching** is a set of disjoint edges





# Maximum weight matching

- Simple, polynomial algorithms for the bipartite case
- Not so simple, still polynomial for the general case

**Maximum cardinality matching:** all weights  $c_{ij}$  are 1

# Maximum weight matching

Given:

- graph  $G = (V, E)$
- weight  $c_{ij}$  for all  $ij \in E$

Formulation:

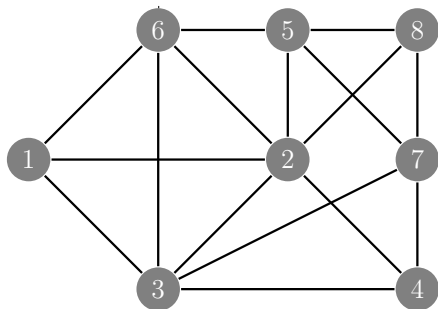
- $x_{ij} = 1$  if edge  $\{i, j\}$  is in the matching, 0 otherwise  $\forall \{i, j\} \in E$

$$\begin{aligned} & \text{maximize} && \sum_{ij \in E} c_{ij} x_{ij} \\ & \text{subject to} && \sum_{j: ij \in E} x_{ij} \leq 1 && \forall i \in V \\ & && x_{ij} \in \{0, 1\} && \forall ij \in E \end{aligned}$$

*Note: I'm abusing notation, and representing an edge  $\{i, j\}$  simply by  $ij$*

# Example: maximum cardinality matching

Consider the graph



Find the maximum cardinality matching

Recall:

$$\begin{array}{ll} \text{maximize} & \sum_{ij \in E} c_{ij} x_{ij} \\ \text{subject to} & \sum_{j:ij \in E} x_{ij} \leq 1 \quad \forall i \in V \\ & x_{ij} \in \{0, 1\} \quad \forall ij \in E \end{array}$$

# Model

Recall:

$$\begin{aligned} & \text{maximize} && \sum_{ij \in E} c_{ij} x_{ij} \\ & \text{subject to} && \sum_{j:ij \in E} x_{ij} \leq 1 && \forall i \in V \\ & && x_{ij} \in \{0, 1\} && \forall ij \in E \end{aligned}$$

---

```
set V;  
set E within {V,V};  
  
var x {E} binary;  
  
maximize z: sum {(i,j) in E} x[i,j];  
  
subject to  
Matched {i in V}:  
    sum {j in V : (i,j) in E} x[i,j] +  
    sum {j in V : (j,i) in E} x[j,i] <= 1;
```

---

```
set V;  
set E within {V,V};  
  
var x {E} binary;  
  
maximize z: sum {(i,j) in E} x[i,j];  
  
subject to  
Matched {i in V}:  
    sum {j in V : (i,j) in E} x[i,j] +  
    sum {j in V : (j,i) in E} x[j,i] <= 1;
```

---

---

```
set V := 1 2 3 4 5 6 7 8;  
set E :=  
    1 2  
    1 3  
    1 6  
    2 3  
    2 4  
    2 5  
    2 6  
    2 8  
    3 4  
    3 6  
    3 7  
    4 7  
    5 6  
    5 7  
    5 8  
    7 8  
  
;
```

---

# Solution

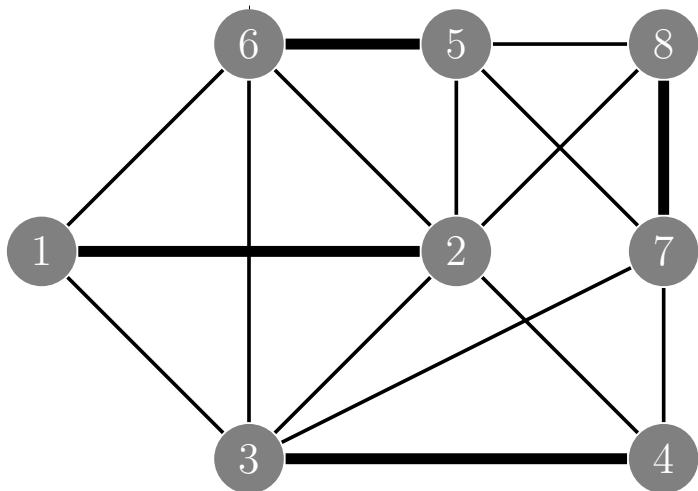
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$z = 4$

$x :=$

1	2	1
1	3	0
1	6	0
2	3	0
2	4	0
2	5	0
2	6	0
2	8	0
3	4	1
3	6	0
3	7	0
4	7	0
5	6	1
5	7	0
5	8	0
7	8	1

---



Given:

- graph  $G = (V, E)$

we can define:

- **covering by nodes**: a set  $R \subseteq V$  such that every edge  $e \in E$  has at least one endpoint in  $R$
- **packing by edges**: a set  $F \subseteq E$  such that every node  $i \in V$  is the endpoint of at most one  $e \in F$ 
  - i.e., a *matching*
  - but we may want to pack with more general structures
    - **kidney exchange problems**

# This lesson

- Graph problems
  - partitioning
  - stable set
- Next lesson:
  - matching problems