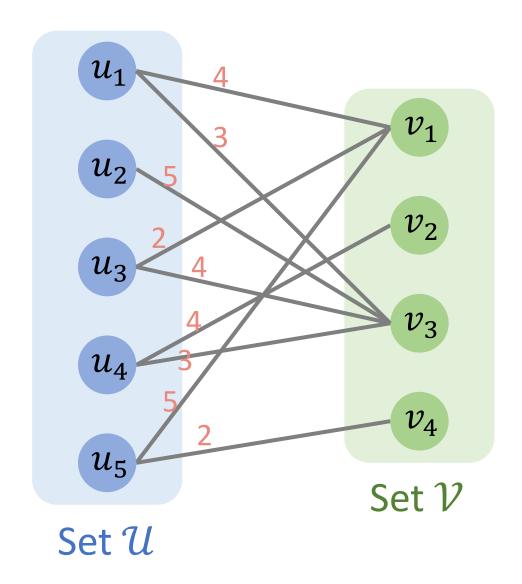
## Maximum-Weight Bipartite Matching

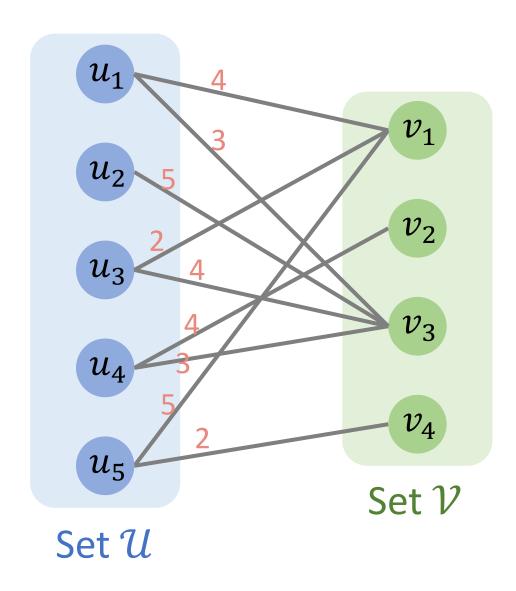
Shusen Wang

## Weighted Bipartite Graph



- Bipartite graph: G = (U, V, E).
- Edges have weights:  $w_{ij}$ .

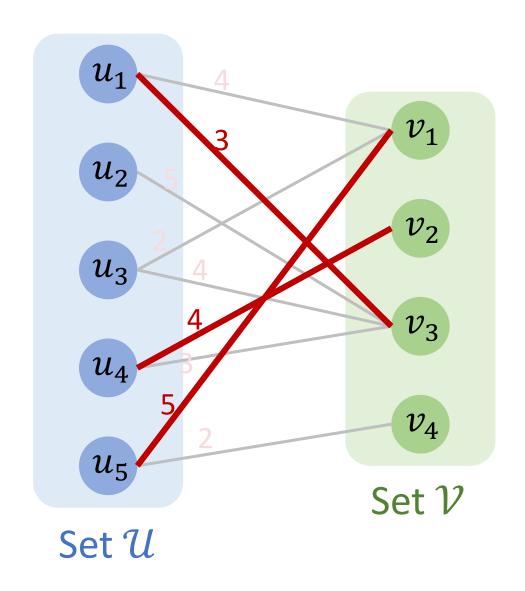
## Weighted Bipartite Graph



- Bipartite graph: G = (U, V, E).
- Edges have weights:  $w_{ij}$ .
- Adjacency matrix:

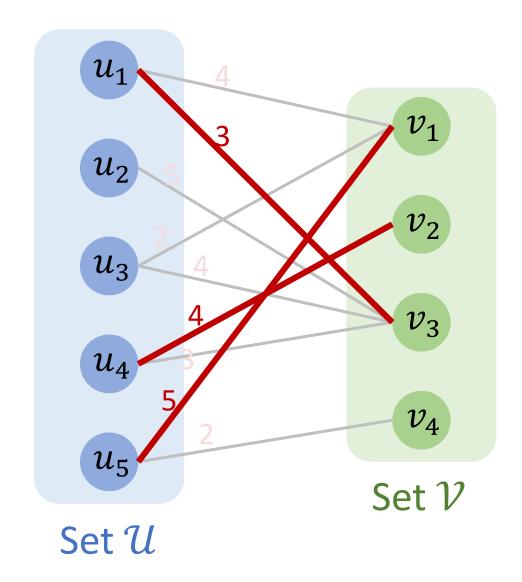
	$v_1$	$v_2$	$v_3$	$v_4$
$u_1$	4	0	3	0
$u_2$	0	0	5	0
$u_3$	2	0	4	0
$u_4$	0	4	3	0
$u_5$	5	0	0	2

#### Bipartite Matching in Weighted Graph



- Bipartite graph: G = (U, V, E).
- Matching is a subset of edges without common vertices.
- Denote the matching by set  $S \subseteq \mathcal{E}$ .

#### Bipartite Matching in Weighted Graph



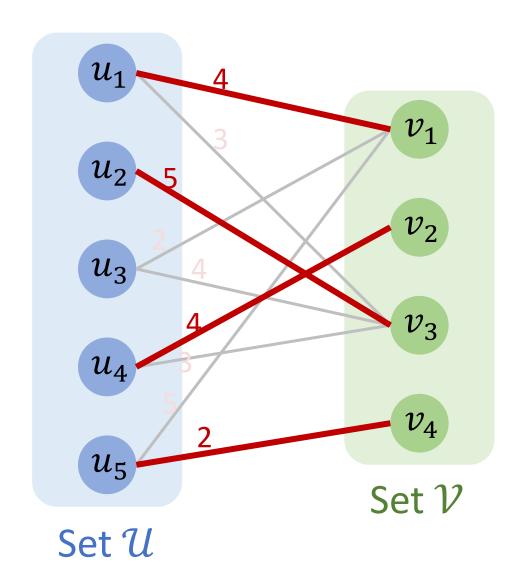
• Sum of weights in matching  $\mathcal{S}$ :

$$f(\mathcal{S}) = \sum_{(u,v)\in\mathcal{S}} w_{uv}.$$

• In this example,

$$f(S) = 3 + 4 + 5 = 12.$$

#### Bipartite Matching in Weighted Graph



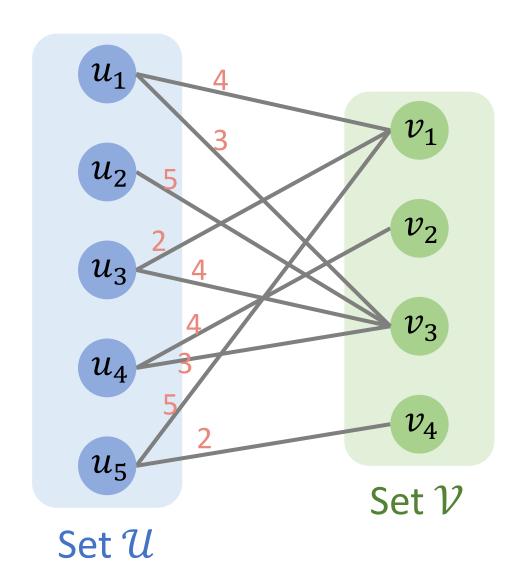
• Sum of weights in matching S:

$$f(\mathcal{S}) = \sum_{(u,v)\in\mathcal{S}} w_{uv}.$$

• In this example,

$$f(S) = 4 + 5 + 4 + 2 = 20.$$

#### Maximum-Weight Bipartite Matching



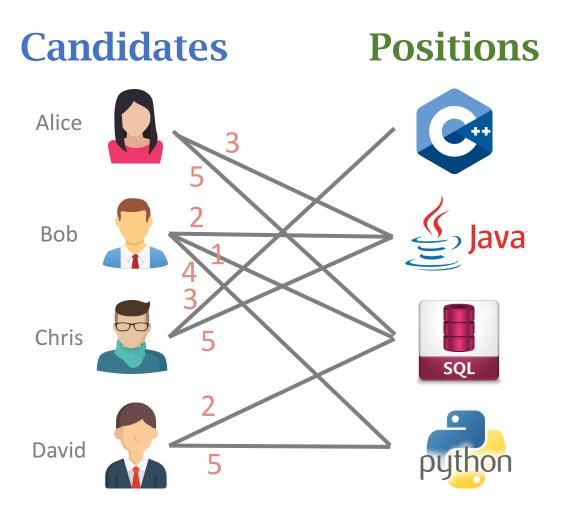
• Sum of weights in matching  $\mathcal{S}$ :

$$f(\mathcal{S}) = \sum_{(u,v)\in\mathcal{S}} w_{uv}.$$

• Objective: Finding matching  $\mathcal{S}$  that has the maximum weight:

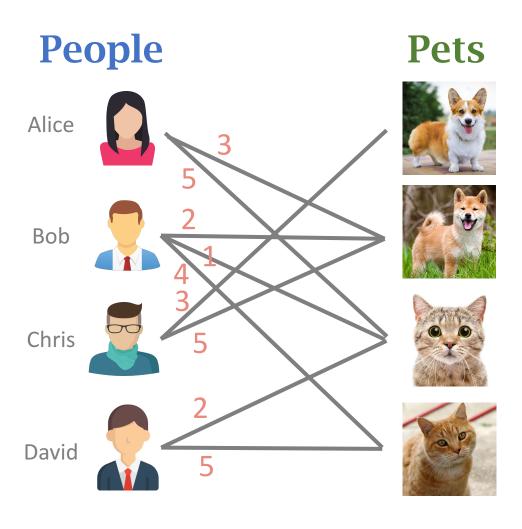
$$\max_{\mathcal{S}} f(\mathcal{S}).$$

#### Application 1: Match candidates and positions



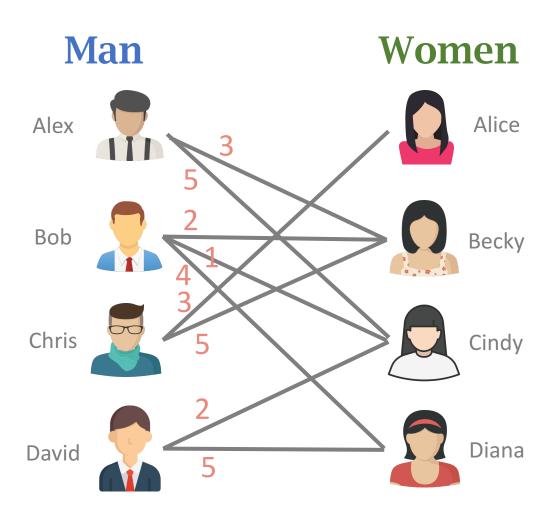
- Edge weights quantify candidates' skills.
- Maximize the weights of matching. (Match the right person with the right job position to maximize the company's interest.)

## **Application 2: Pet adoptions**



- Edge weight quantifies how much a person loves a pet.
- Maximize the weights of matching. (Maximize people's happiness.)

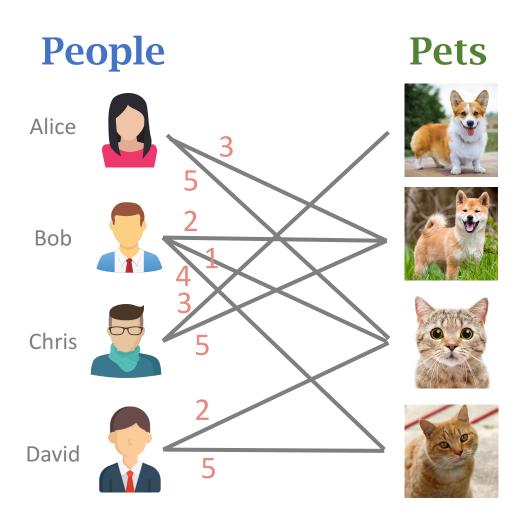
## **Application 3: Dating**



- Edge weights quantify how well two persons match (e.g., similar hoppy).
- Maximize the weights of matching.
   (Maximize the change of success.)

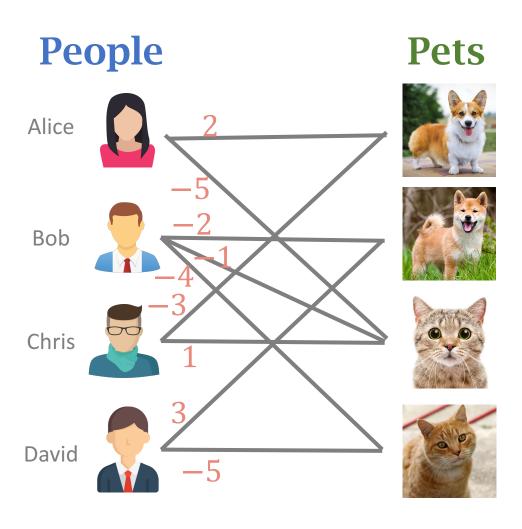
## 

## **Maximum Matching**



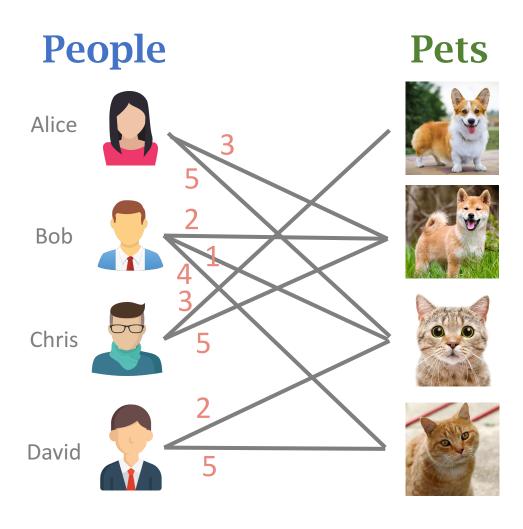
- Adopting a pet can bring happiness to people.
- A weight quantifies how much a person loves a pet.
- Maximize the weights of matching. (Maximize people's happiness.)

## **Minimum Matching**



- Adopting a pet can cost time and money.
- A weight quantifies how a person dislike a pet.
- Minimize the weights of matching. (Maximize people's happiness.)

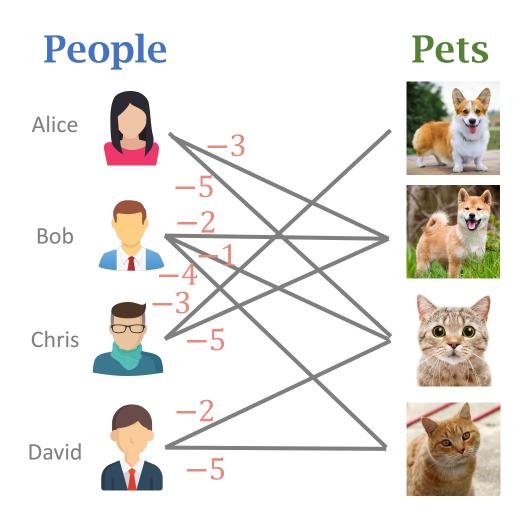
# 



- If we have an algorithm for finding minimum matching.
- Then we can use it for finding maximum matching.

## 

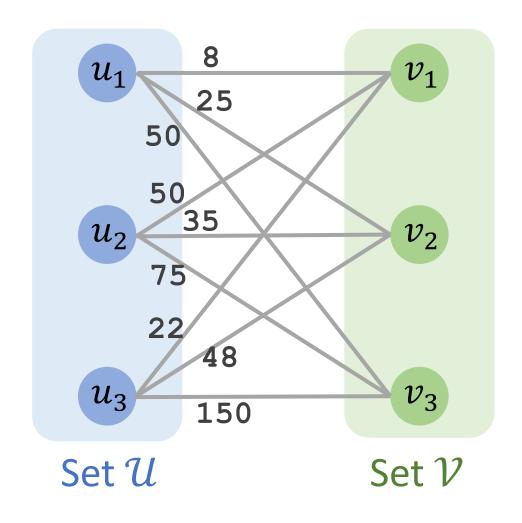




- If we have an algorithm for finding minimum matching.
- Then we can use it for finding maximum matching.
  - 1. Flip the signs of all the weights.
  - 2. Run the minimum matching algorithm.

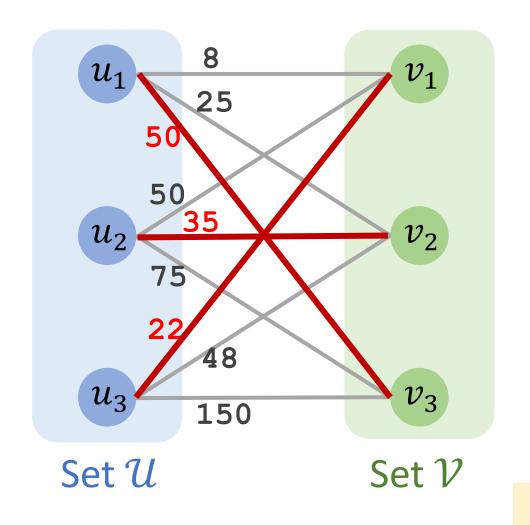
#### Hungarian Algorithm for Minimum-Weight Bipartite Matching

## Minimum-Weight Bipartite Matching



	$v_1$	$v_2$	$v_3$
$u_1$	8	25	50
$u_2$	50	35	75
$u_3$	22	48	150

## Minimum-Weight Bipartite Matching



	$v_1$	$v_2$	$v_3$
$u_1$	8	25	50
$u_2$	50	35	75
$u_3$	22	48	150

The minimum sum of weight is 50 + 35 + 22 = 107.

- Hungarian algorithm is for finding the minimum-weight bipartite matching.
- In the graph, the cardinality of  $\mathcal{U}$  and  $\mathcal{V}$  must be the same:

$$|\mathcal{U}| = |\mathcal{V}| = n$$
.

• Time complexity:  $O(n^3)$ .

	$v_1$	$v_2$	$v_3$
$u_1$	8	25	50
$u_2$	50	35	75
$u_3$	22	48	150

#### Reference:

 Harold W. Kuhn. The Hungarian Method for the assignment problem. Naval Research Logistics Quarterly, 2: 83–97, 1955.

• Step 1: Subtract row minima.

	$v_1$	$v_2$	$v_3$
$u_1$	8	25	50
$u_2$	50	35	75
$u_3$	22	48	150

- Step 1: Subtract row minima.
  - Subtract the smallest entry of each row from all the entries in the row.

	$v_1$	$v_2$	$v_3$
$u_1$	8 -8	25 -8	50 -8
$u_2$	50	35	75
	-35	-35	-35
$u_3$	22	48	150
	-22	-22	-22

- Step 1: Subtract row minima.
  - Subtract the smallest entry of each row from all the entries in the row.
  - The minimum of the row is equal to 0.

	$v_1$	$v_2$	$v_3$
$u_1$	0	17	42
$u_2$	15	0	40
$u_3$	0	26	128

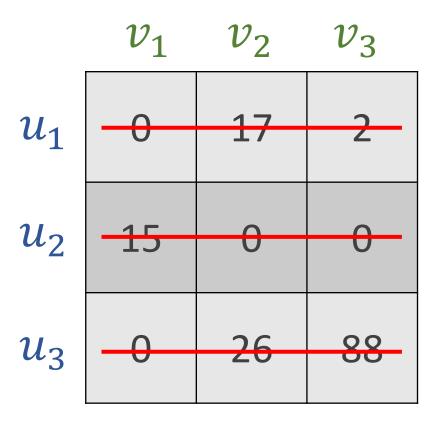
- Step 1: Subtract row minima.
  - Subtract the smallest entry of each row from all the entries in the row.
  - The minimum of the row is equal to 0.
- Step 2: Subtract column minima.
  - Subtract the smallest entry of each column from all the entries in the column.

	$v_1$	$v_2$	$v_3$
$u_1$	0 -0	17 -0	42 -40
$u_2$	15 -0	0 -0	40 -40
$u_3$	0 -0	26 -0	128 -40

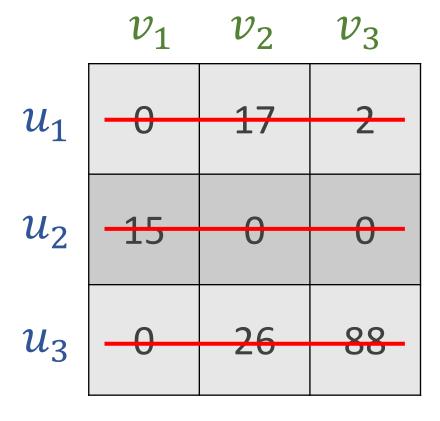
- Step 1: Subtract row minima.
  - Subtract the smallest entry of each row from all the entries in the row.
  - The minimum of the row is equal to 0.
- Step 2: Subtract column minima.
  - Subtract the smallest entry of each column from all the entries in the column.
  - The minimum of the column is equal to
    0.

	$v_1$	$v_2$	$v_3$
$u_1$	0	17	2
$u_2$	15	0	0
$u_3$	0	26	88

- Step 3A: Cover all zeros with a minimum number of lines.
  - Use either horizontal or vertical lines.
  - Minimize the total number of lines.

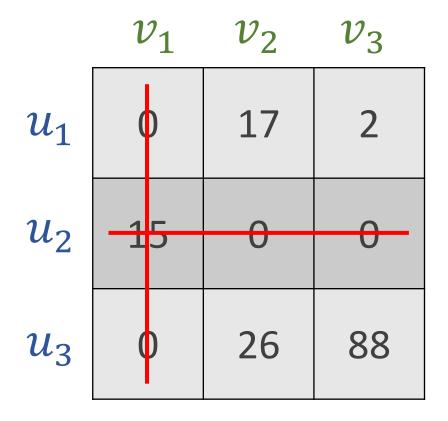


- Step 3A: Cover all zeros with a minimum number of lines.
  - Use either horizontal or vertical lines.
  - Minimize the total number of lines.



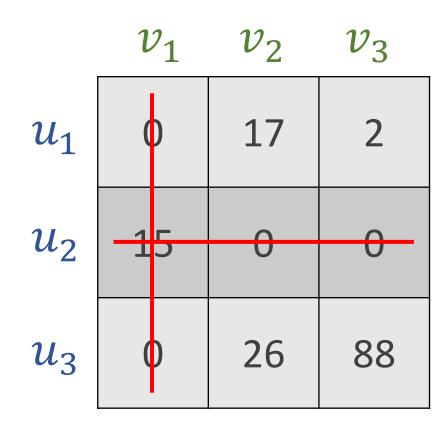
- The number of line is 3.
- It is NOT the minimum.

- Step 3A: Cover all zeros with a minimum number of lines.
  - Use either horizontal or vertical lines.
  - Minimize the total number of lines.

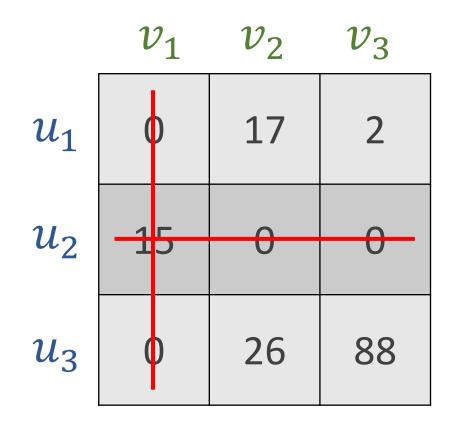


- The number of line is 2.
- It is the minimum.

- Step 3A: Cover all zeros with a minimum number of lines.
- Step 3B: Decide whether to stop.
  - If *n* lines are required, the algorithm stops.
  - If less than *n* lines are required, continue with Step 3C.
- The number of line is 2.
- Number of vertices is n = 3.
- Thus continue to Step 3C.

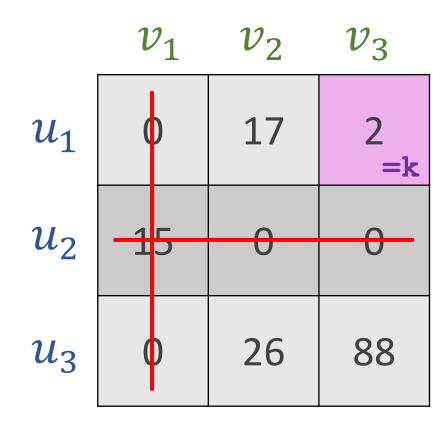


- Step 3A: Cover all zeros with a minimum number of lines.
- Step 3B: Decide whether to stop.
- Step 3C: Create additional zeros.
  - Find the smallest element (call it *k*) that is not covered by a line.



#### Step 3: Repeat the following:

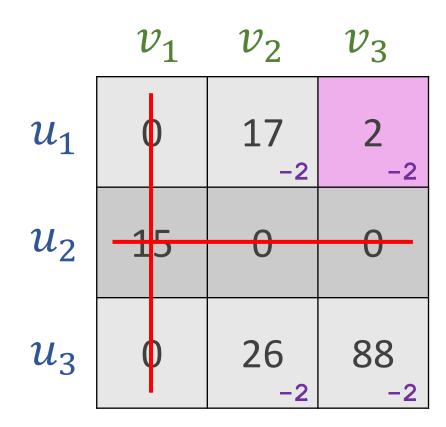
- Step 3A: Cover all zeros with a minimum number of lines.
- Step 3B: Decide whether to stop.
- Step 3C: Create additional zeros.
  - Find the smallest element (call it *k*) that is not covered by a line.



The smallest element that is not covered is k = 2.

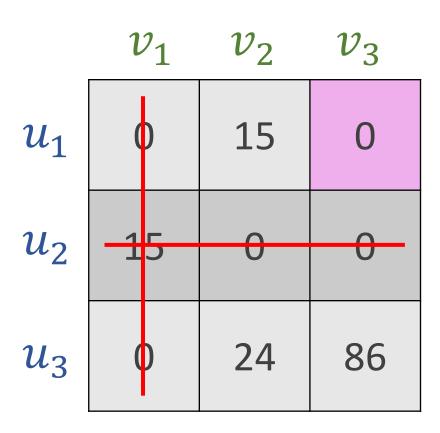
#### Step 3: Repeat the following:

- Step 3A: Cover all zeros with a minimum number of lines.
- Step 3B: Decide whether to stop.
- Step 3C: Create additional zeros.
  - Find the smallest element (call it k) that is not covered by a line.
  - Subtract k from all uncovered elements.

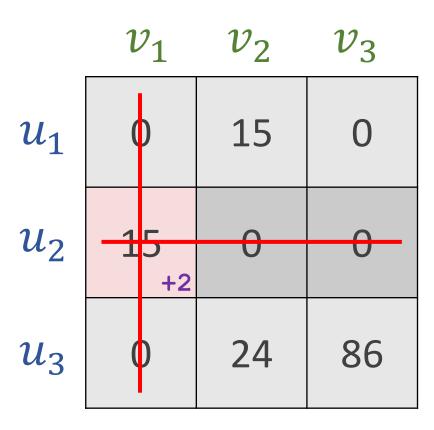


The smallest element that is not covered is k = 2.

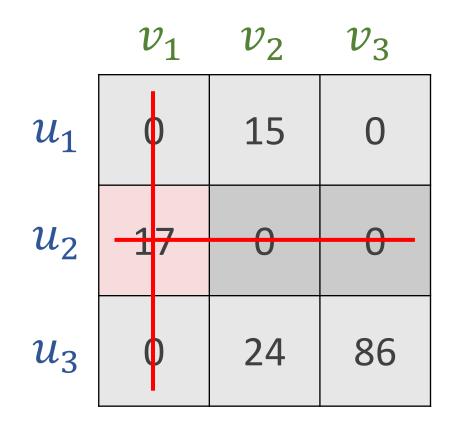
- Step 3A: Cover all zeros with a minimum number of lines.
- Step 3B: Decide whether to stop.
- Step 3C: Create additional zeros.
  - Find the smallest element (call it k) that is not covered by a line.
  - Subtract k from all uncovered elements.



- Step 3A: Cover all zeros with a minimum number of lines.
- Step 3B: Decide whether to stop.
- Step 3C: Create additional zeros.
  - Find the smallest element (call it k) that is not covered by a line.
  - Subtract k from all uncovered elements.
  - Add k to all elements that are covered twice.



- Step 3A: Cover all zeros with a minimum number of lines.
- Step 3B: Decide whether to stop.
- Step 3C: Create additional zeros.
  - Find the smallest element (call it *k*) that is not covered by a line.
  - Subtract k from all uncovered elements.
  - Add k to all elements that are covered twice.



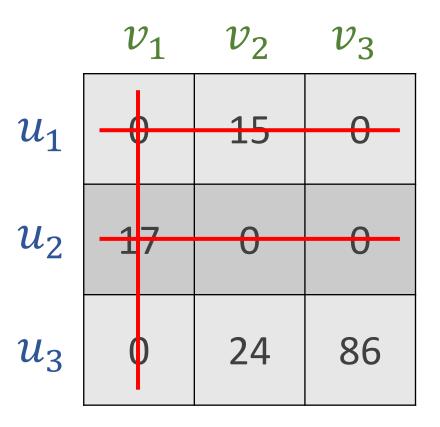
#### Step 3: Repeat the following:

• Step 3A: Cover all zeros with a minimum number of lines.

	$v_1$	$v_2$	$v_3$
$u_1$	0	15	0
$u_2$	17	0	0
$u_3$	0	24	86

#### **Step 3:** Repeat the following:

 Step 3A: Cover all zeros with a minimum number of lines.

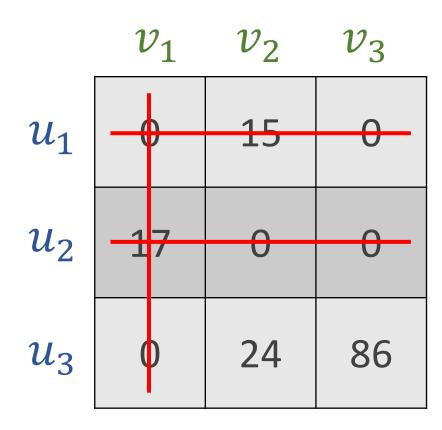


At least 3 lines are needed.

#### **Step 3:** Repeat the following:

- Step 3A: Cover all zeros with a minimum number of lines.
- Step 3B: Decide whether to stop.
  - If *n* lines are required, the algorithm stops.

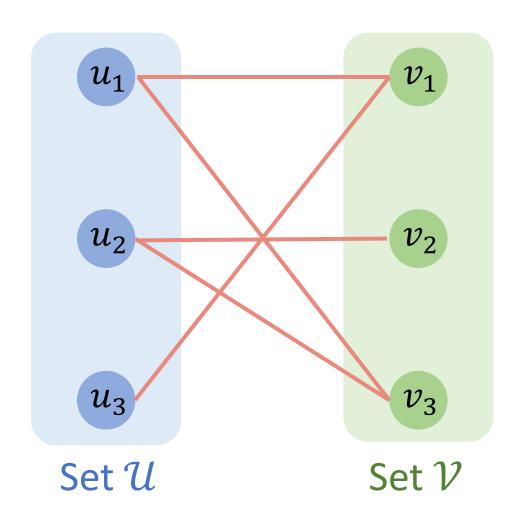
The algorithm stops.



	$v_1$	$v_2$	$v_3$
$u_1$	0	15	0
$u_2$	17	0	0
$u_3$	0	24	86

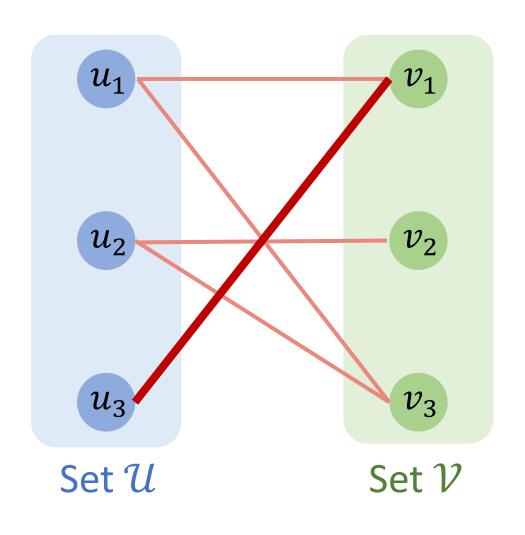
	$v_1$	$v_2$	$v_3$
$u_1$	0	15	0
$u_2$	17	0	0
$u_3$	0	24	86

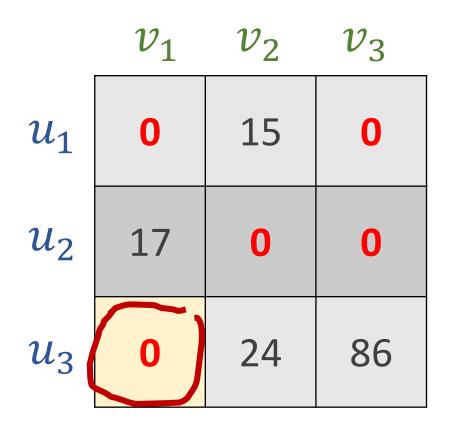
- Choose a matching among the zeros.
- Think of the zeros as edges.



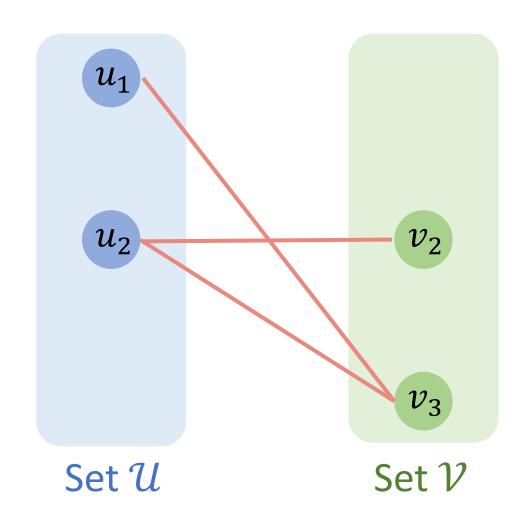
	$v_1$	$v_2$	$v_3$
$u_1$	0	15	0
$u_2$	17	0	0
$u_3$	0	24	86

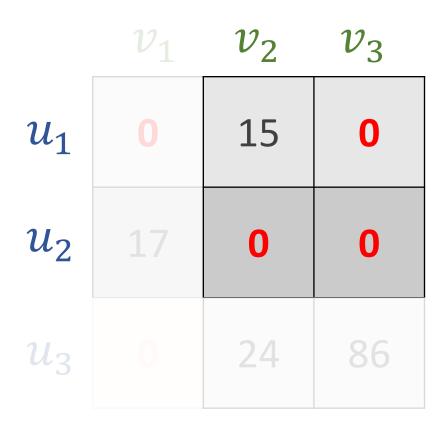
- Choose a matching among the zeros.
- Think of the zeros as edges.



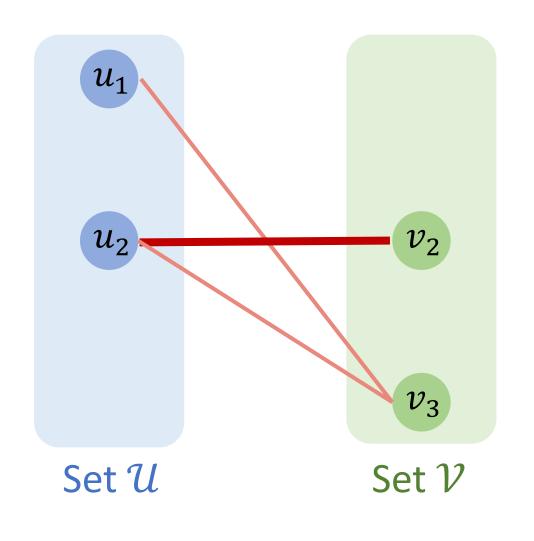


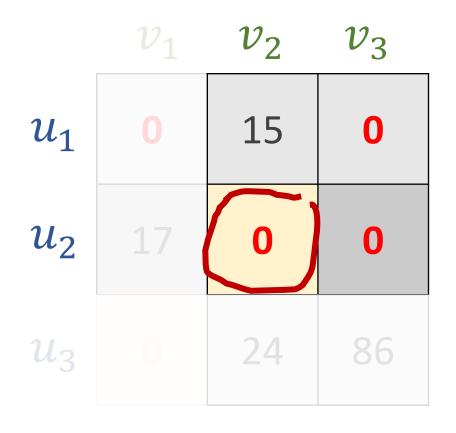
- The edge  $(u_3, v_1)$  must be chosen.
- Because it is the only zero in the row.



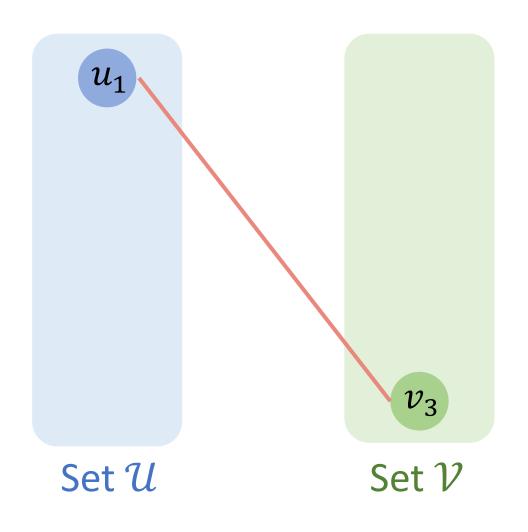


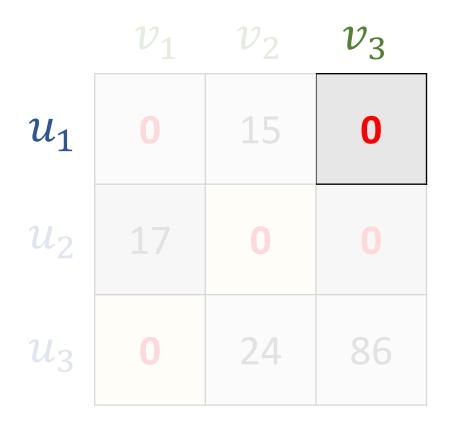
- Cover the row of  $u_3$ .
- Cover the column of  $v_1$ .



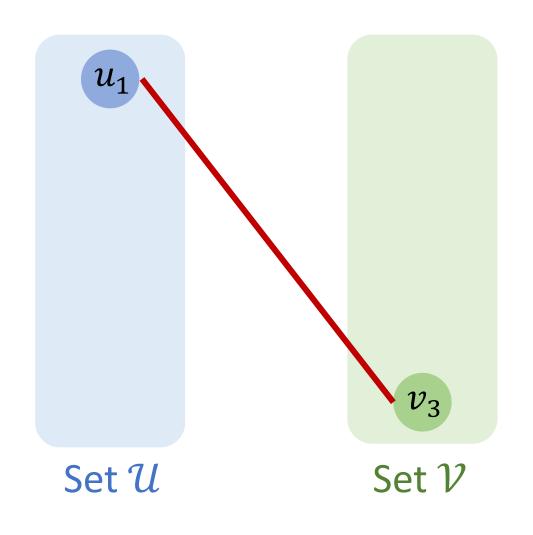


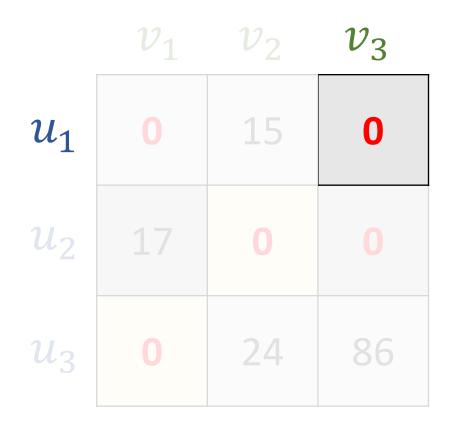
- The edge  $(u_2, v_2)$  must be chosen.
- Because it is the only zero in the column.



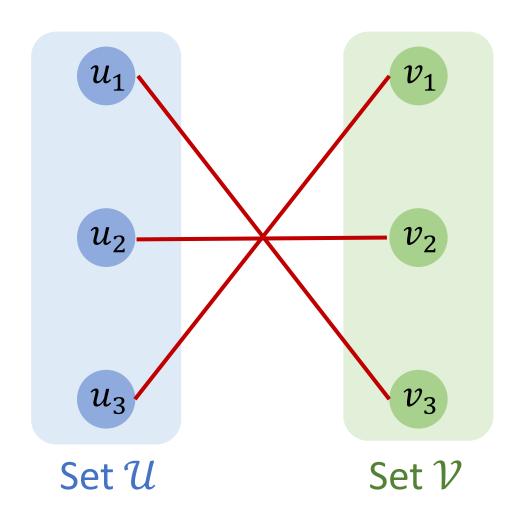


- Cover the row of  $u_2$ .
- Cover the column of  $v_2$ .





- The edge  $(u_1, v_3)$  must be chosen.
- Because it is the only zero.



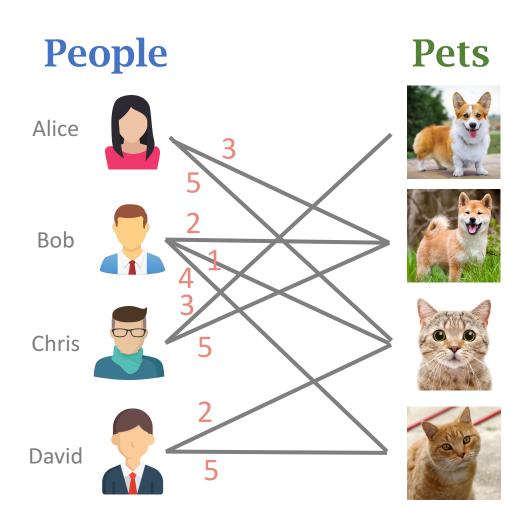
	$v_1$	$v_2$	$v_3$
$u_1$	0	15	0
$u_2$	17	0	0
$u_3$	0	24	86

The matching is

$$S = \{(u_3, v_1), (u_1, v_3), (u_2, v_2)\}.$$

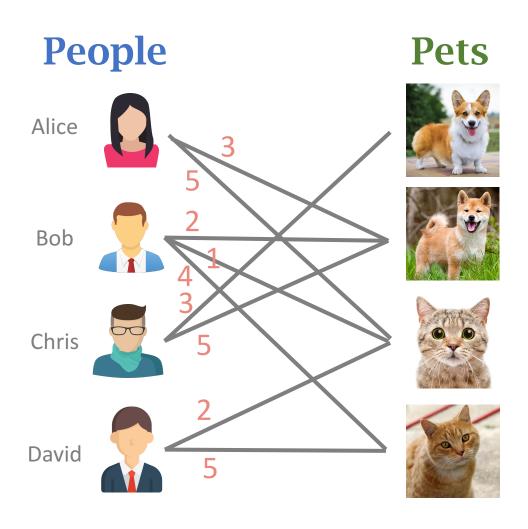


## **Maximum Matching**



- Pet adoption is a max matching problem.
- A weight quantifies how much a person loves a pet.
- Maximize the weights of matching. (Maximize people's happiness.)

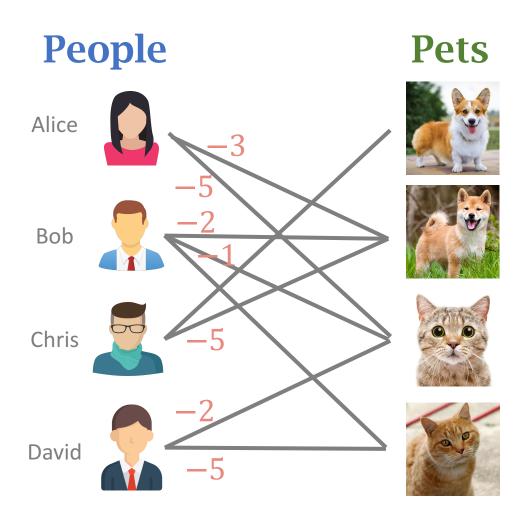
## **Hungarian Algorithm for Maximum Matching**



Idea: Max Matching → Min Matching

- Flip the signs of all the weights.
- It is equivalent to the minimum matching.
- Run the Hungarian algorithm.

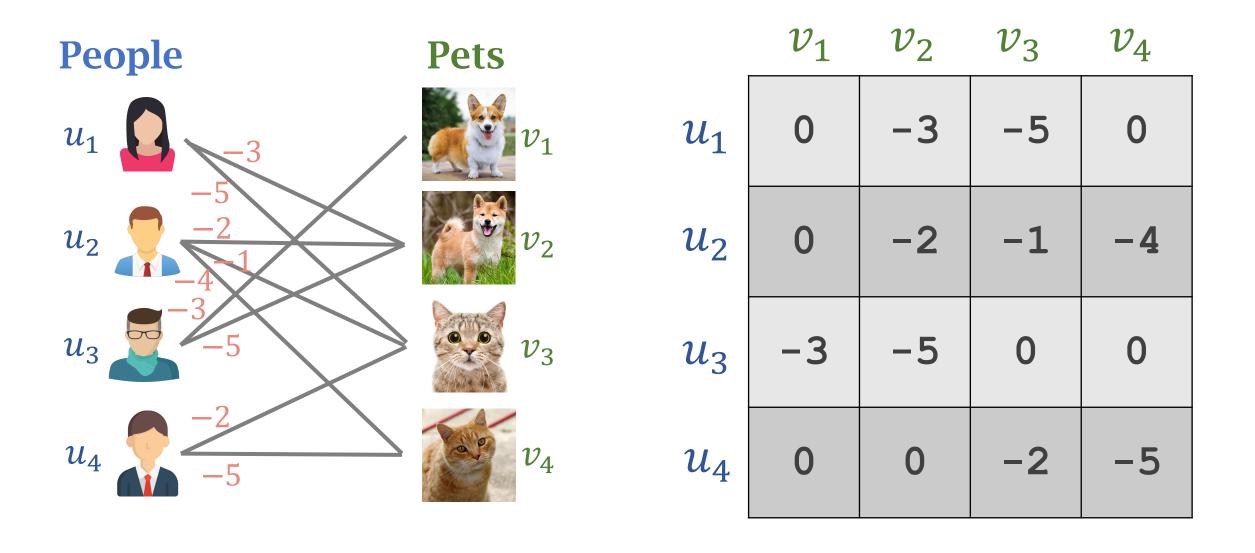
## **Hungarian Algorithm for Maximum Matching**



Idea: Max Matching → Min Matching

- Flip the signs of all the weights.
- It is equivalent to the minimum matching.
- Run the Hungarian algorithm.

## **Hungarian Algorithm for Maximum Matching**



- Step 1: Subtract row minima.
  - Subtract the smallest entry of each row from all the entries in the row.

	$\nu_1$	$\nu_2$	$\nu_3$	$\nu_4$
$\iota_1$	0	-3	<b>-5</b>	0
	- (-5)	-(-5)	-(-5)	- (-5)
$l_2$	0	-2	-1	-4
	-(-4)	-(-4)	-(-4)	-(-4)
$l_3$	-3	<b>-5</b>	0	0
	-(-5)	-(-5)	- (-5)	- (-5)
$\iota_4$	0	0	-2	-5
	- (-5)	- (-5)	-(-5)	-(-5)

11

- Step 1: Subtract row minima.
  - Subtract the smallest entry of each row from all the entries in the row.
  - The minimum of the row is equal to 0.

	$v_1$	$v_2$	$v_3$	$v_4$
$u_1$	5	2	0	5
$u_2$	4	2	3	0
$u_3$	2	0	5	5
$u_4$	5	5	3	0

- Step 1: Subtract row minima.
  - Subtract the smallest entry of each row from all the entries in the row.
  - The minimum of the row is equal to 0.
- Step 2: Subtract column minima.
  - Subtract the smallest entry of each column from all the entries in the column.

	$v_1$	$v_2$	$v_3$	$v_4$
$u_1$	5 -2	2 -0	0 -0	5 -0
$u_2$	4 -2	2 -0	3 -0	0 -0
$u_3$	2 -2	0 -0	5 -0	5 -0
$u_4$	5 -2	5 -0	3 -0	0 -0

 $u_1$ 

 $u_2$ 

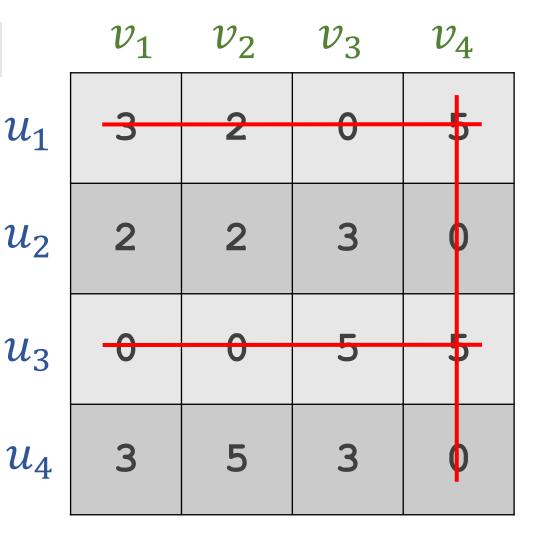
 $u_3$ 

 $u_4$ 

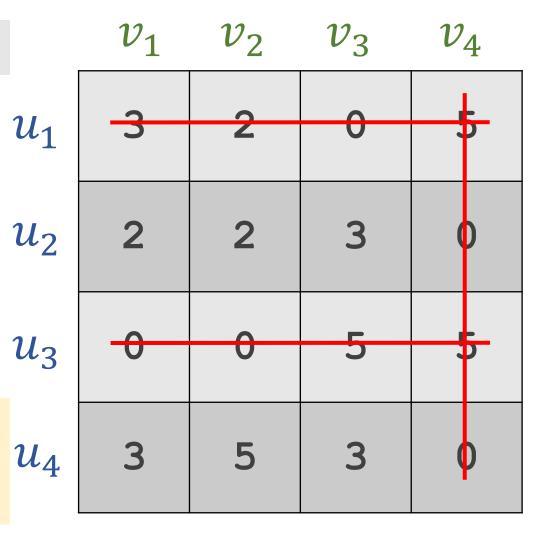
- Step 1: Subtract row minima.
  - Subtract the smallest entry of each row from all the entries in the row.
  - The minimum of the row is equal to 0.
- Step 2: Subtract column minima.
  - Subtract the smallest entry of each column from all the entries in the column.
  - The minimum of the column is equal to
    0.

$v_1$	$v_2$	$v_3$	$v_4$
3	2	0	5
2	2	3	0
0	0	5	5
3	5	3	0

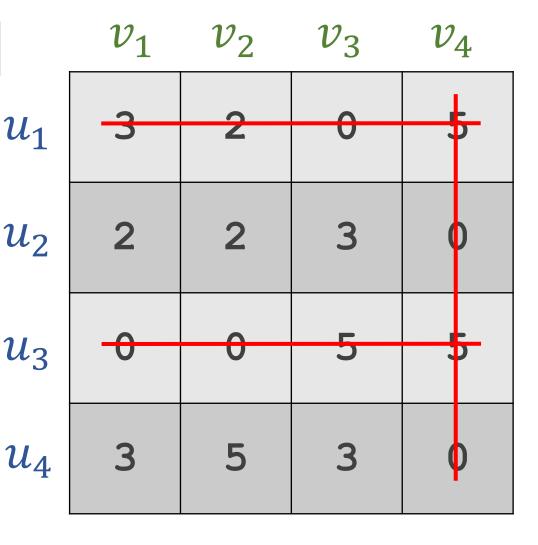
- Step 3A: Cover all zeros with a minimum  $u_1$  number of lines.
  - Use either horizontal or vertical lines.
  - Minimize the total number of lines.



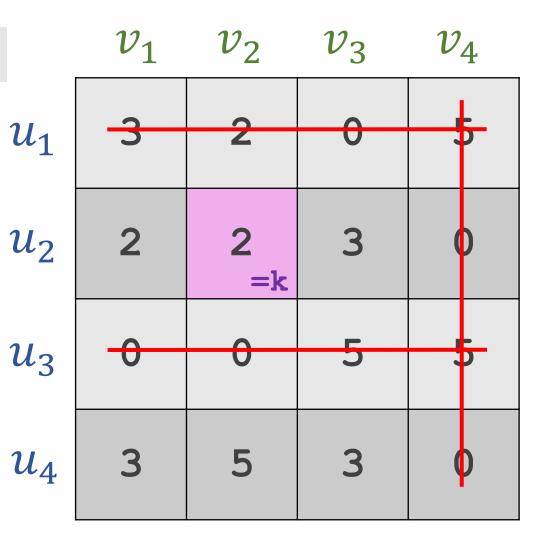
- Step 3A: Cover all zeros with a minimum  $u_1$  number of lines.
- Step 3B: Decide whether to stop.
  - If *n* lines are required, the algorithm stops.
  - If less than *n* lines are required, continue with Step 3C.
- The number of line is 3.
- Number of vertices is n = 4.
- Thus continue to Step 3C.



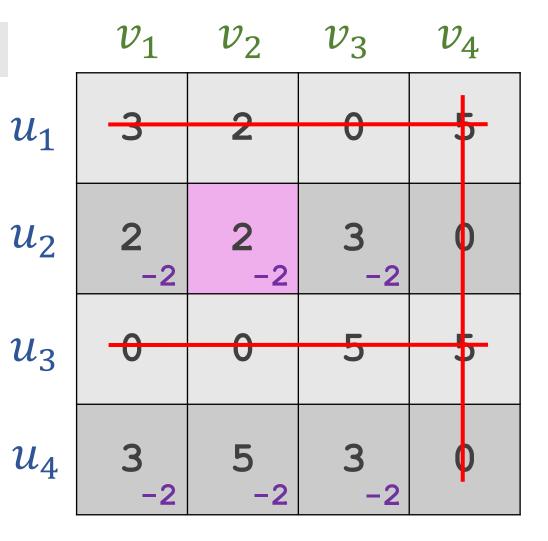
- Step 3A: Cover all zeros with a minimum  $u_1$  number of lines.
- Step 3B: Decide whether to stop.
- Step 3C: Create additional zeros.
  - Find the smallest element (call it *k*) that is not covered by a line.



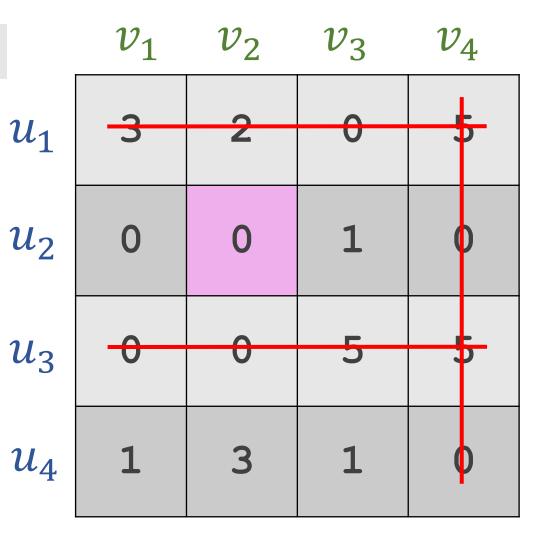
- Step 3A: Cover all zeros with a minimum  $u_1$  number of lines.
- Step 3B: Decide whether to stop.
- Step 3C: Create additional zeros.
  - Find the smallest element (call it *k*) that is not covered by a line.



- Step 3A: Cover all zeros with a minimum number of lines.
- Step 3B: Decide whether to stop.
- Step 3C: Create additional zeros.
  - Find the smallest element (call it *k*) that is not covered by a line.
  - Subtract k from all uncovered elements.



- Step 3A: Cover all zeros with a minimum  $u_1$  number of lines.
- Step 3B: Decide whether to stop.
- Step 3C: Create additional zeros.
  - Find the smallest element (call it *k*) that is not covered by a line.
  - Subtract k from all uncovered elements.



 $u_2$ 

 $u_3$ 

- Step 3A: Cover all zeros with a minimum  $u_1$  number of lines.
- Step 3B: Decide whether to stop.
- Step 3C: Create additional zeros.
  - Find the smallest element (call it k) that is not covered by a line.
  - Subtract k from all uncovered elements.
  - Add k to all elements that are covered twice.

$v_1$	$v_2$	$v_3$	$v_4$
	0		Į.
3	2	0	+2
0	0	1	•
0	0		
U	0	5	+2
1	3	1	ф

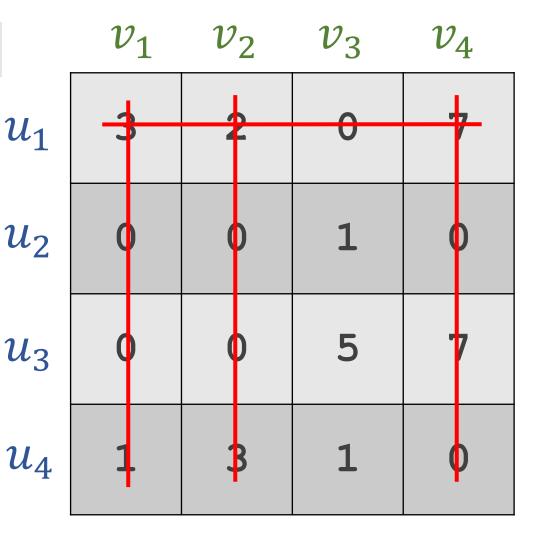
 $u_2$ 

 $u_3$ 

- Step 3A: Cover all zeros with a minimum  $u_1$  number of lines.
- Step 3B: Decide whether to stop.
- Step 3C: Create additional zeros.
  - Find the smallest element (call it k) that is not covered by a line.
  - Subtract k from all uncovered elements.
  - Add k to all elements that are covered twice.

${v}_1$	$v_2$	$v_3$	$v_4$
2	2		+,_
3	2	U	
0	0	1	0
0	0	F	• 7
0	U	3	
1	3	1	ф

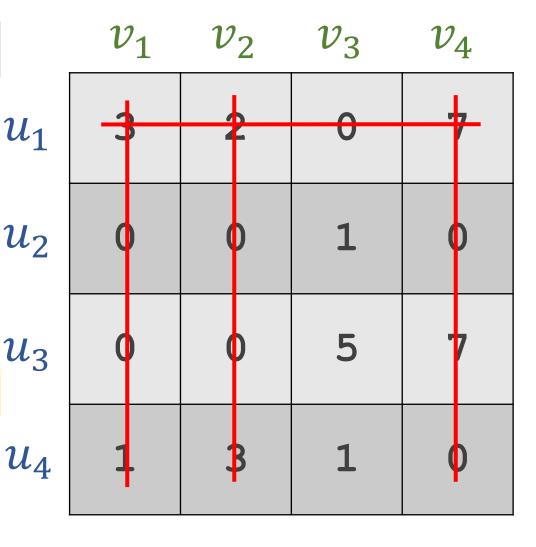
- Step 3A: Cover all zeros with a minimum  $u_1$  number of lines.
  - Use either horizontal or vertical lines.
  - Minimize the total number of lines.

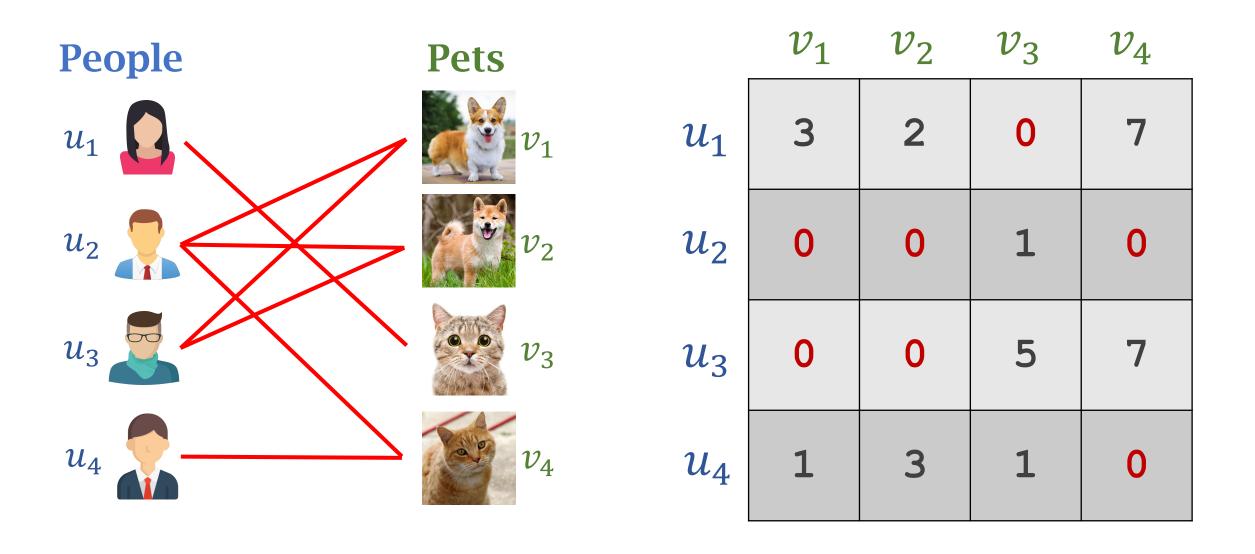


#### **Step 3:** Repeat the following:

- Step 3A: Cover all zeros with a minimum u number of lines.
- Step 3B: Decide whether to stop.
  - If *n* lines are required, the algorithm stops.

The algorithm stops.





- The edge  $(u_1, v_3)$  must be selected.
  - Otherwise,  $u_1$  would have no matching.

	$v_1$	$v_2$	$v_3$	$v_4$
$u_1$	3	2		7
$u_2$	0	0	1	0
$u_3$	0	0	5	7
$u_4$	1	3	1	0

- The edge  $(u_1, v_3)$  must be selected.
  - Otherwise,  $u_1$  would have no matching.
  - Cover the row of  $u_1$ .
  - Cover the column of  $v_3$ .

	$v_1$	$v_2$	$v_3$	$v_4$
$u_1$	3	2		7
$u_2$	0	0		0
$u_3$	0	0	5	7
$u_4$	1	3	1	0

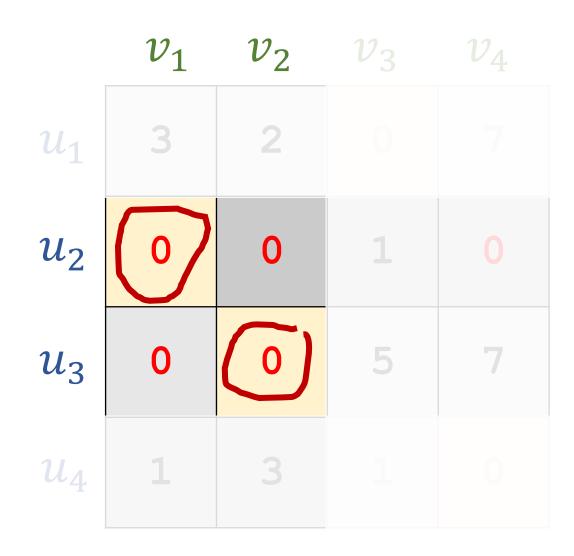
- The edge  $(u_4, v_4)$  must be selected.
  - Otherwise,  $u_4$  would have no matching.

	$v_1$	$v_2$	$v_3$	$v_4$
$u_1$	3	2		7
$u_2$	0	0	1	0
$u_3$	0	0	5	7
$u_4$	1	3	1	0

- The edge  $(u_4, v_4)$  must be selected.
  - Otherwise,  $u_4$  would have no matching.
  - Cover the row of  $u_4$ .
  - Cover the column of  $v_4$ .

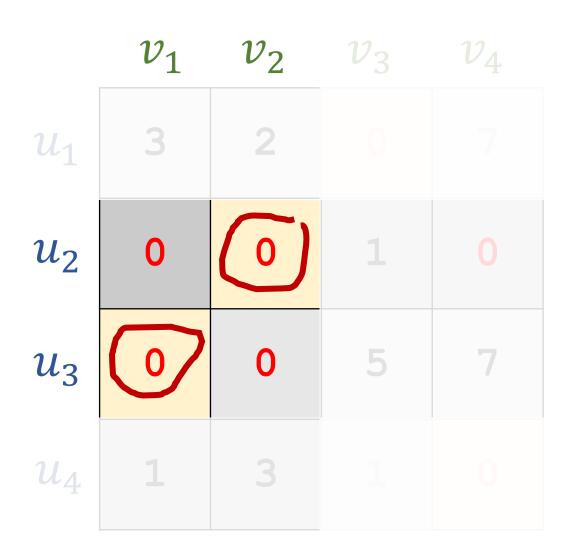
	$v_1$	$v_2$	$v_3$	
$u_1$	3	2		
$u_2$	0	0	1	0
$u_3$	0	0	5	7
$u_4$	1	3		

• Select edges  $(u_2, v_1)$  and  $(u_3, v_2)$ .



• Select edges  $(u_2, v_1)$  and  $(u_3, v_2)$ .

• Or select edges  $(u_3, v_1)$  and  $(u_2, v_2)$ .



• Return the matching:

$$S = \{(u_1, v_3), (u_4, v_4), (u_2, v_1), (u_3, v_2)\}.$$

• Or return the matching:

$$S = \{(u_1, v_3), (u_4, v_4), (u_3, v_1), (u_2, v_2)\}.$$

	$v_1$	$v_2$	$v_3$	$v_4$
$u_1$	3	2	(0)	7
$u_2$	$\bigcirc$	0	1	0
$u_3$	0	0	5	7
$u_4$	1	3	1	0

Return the matching:

$$S = \{(u_1, v_3), (u_4, v_4), (u_2, v_1), (u_3, v_2)\}.$$

• Or return the matching:

$$S = \{(u_1, v_3), (u_4, v_4), (u_3, v_1), (u_2, v_2)\}.$$

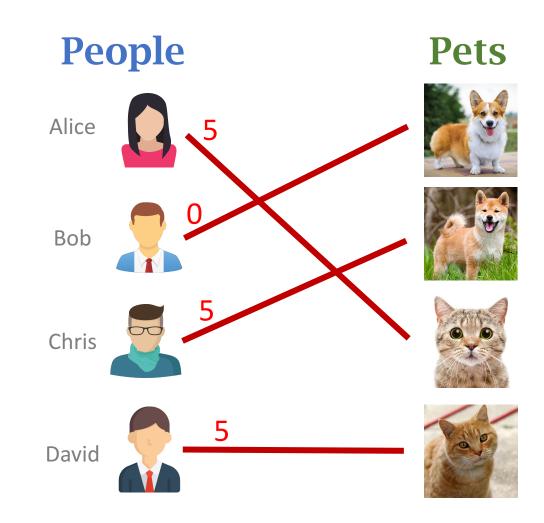
	$v_1$	$v_2$	$v_3$	$v_4$
$u_1$	3	2	(o)	7
$u_2$	0	(0)	1	0
$u_3$	6	0	5	7
$u_4$	1	3	1	0

• Return the matching:

$$S = \{(u_1, v_3), (u_4, v_4), (u_2, v_1), (u_3, v_2)\}.$$

- The matching is equal to 15.
- Or return the matching:

$$S = \{(u_1, v_3), (u_4, v_4), (u_3, v_1), (u_2, v_2)\}.$$



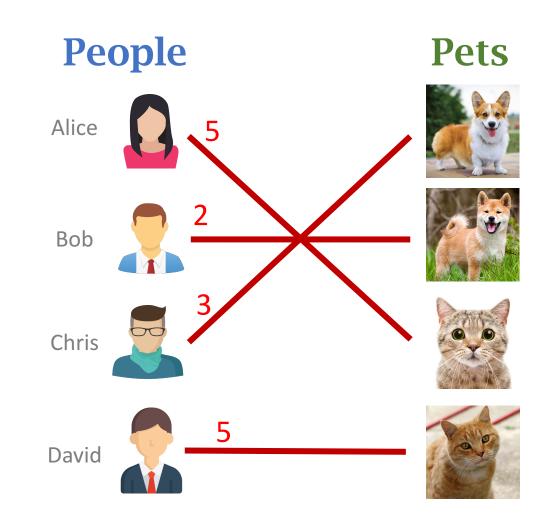
Return the matching:

$$S = \{(u_1, v_3), (u_4, v_4), (u_2, v_1), (u_3, v_2)\}.$$

- The matching is equal to 15.
- Or return the matching:

$$S = \{(u_1, v_3), (u_4, v_4), (u_3, v_1), (u_2, v_2)\}.$$

• The matching is equal to 15.



# **Summary**

### **Maximum-Weight Bipartite Matching**

- Weighted bipartite graph:  $G = (U, V, \mathcal{E})$ . (Edges have weights:  $w_{uv}$ .)
- Matching is a subset of edges without common vertices.
- Denote the matching by set  $S \subseteq \mathcal{E}$ .
- Sum of weights in matching S:

$$f(\mathcal{S}) = \sum_{(u,v)\in\mathcal{S}} w_{uv}.$$

• Find matching S that has the maximum weight:

$$\max_{\mathcal{S}} f(\mathcal{S}).$$

# 

- Maximum matching:  $\max_{\mathcal{S}} f(\mathcal{S})$ .
- Minimum matching:  $\min_{\mathcal{S}} f(\mathcal{S})$ .
- The maximum matching problem can be reduced to minimum matching problem by flipping the signs of weights.
- Algorithms that solve minimum matching can also solve the maximum matching problem.

### **Hungarian Algorithm**

- Hungarian algorithm finds minimum-weight bipartite matching.
- It requires  $|\mathcal{U}| = |\mathcal{V}| = n$ .
- Time complexity:  $O(n^3)$ .

# Questions

## **Question 1**

- The right is the adjacency matrix of a bipartite graph.
- Find the minimum matching in the graph.

	$v_1$	$v_2$	$v_3$	$v_4$	$v_5$
$u_1$	20	15	18	24	25
$u_2$	18	20	12	14	15
$u_3$	21	23	25	27	26
$u_4$	17	18	21	23	22
$u_5$	19	22	16	21	20

#### Question 2

- The right is the adjacency matrix of a bipartite graph.
- Find the maximum matching in the graph.

	$v_1$	$v_2$	$v_3$	$v_4$	$v_5$
$u_1$	20	15	18	24	25
$u_2$	18	20	12	14	15
$u_3$	21	23	25	27	26
$u_4$	17	18	21	23	22
$u_5$	19	22	16	21	20

## Thank You!