#### Minimum cost bipartite matching Complements of Operations Research

Giovanni Righini

Università degli Studi di Milano

## Definitions

Given a graph  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$  a matching is an edge subset  $\mathcal{M} \subseteq \mathcal{E}$  such that it is not incident to any vertex more than once.

A matching is maximal if and only if there is no other matching containing it.

A matching has maximum cardinality if and only if it contains the maximum number of edges of  $\mathcal{E}$ .

A graph  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$  is bipartite when  $\mathcal{V}$  is formed by two disjoint subsets  $\mathcal{S}$  and  $\mathcal{T}$  and all edges  $[i, j] \in \mathcal{E}$  have an endpoint in  $\mathcal{S}$  and the other in  $\mathcal{T}$ .

A bipartite graph  $\mathcal{G} = (\mathcal{S}, \mathcal{T}, \mathcal{E})$  is complete when there are all possible edges between  $\mathcal{S}$  and  $\mathcal{T}$ , i.e.  $\mathcal{E} = \mathcal{S} \times \mathcal{T}$ .



## The problem

We consider the problem of finding the *matching* of maximum cardinality and minimum cost between two vertex subsets S and T defining a weighted bipartite graph.

Data:

- a bipartite graph  $\mathcal{G} = (\mathcal{S}, \mathcal{T}, \mathcal{E})$ ,
- a cost function  $c : \mathcal{E} \mapsto \Re$ .

**Problem (Minimum Cost Bipartite Matching Problem).** Find a minimum cost *matching* between  $S \in T$  among all those with maximum cardinality.



### Graph pre-processing

We assume:

- the two partitions are balanced: |S| = |T| = n
- the graph is complete:  $\mathcal{E} = \mathcal{S} \times \mathcal{T}$ .

**Observation.** If these conditions do not hold, it is always possible to reformulate the problem in an equivalent way on a complete balanced bipartite graph.



### Graph pre-processing

#### Balancing the graph.

If the given bipartite graph is not balanced, we insert dummy vertices in the partition of smaller cardinality, to make it balanced.

No matching is affected by this operation.

#### Completing the graph.

If the given graph is not complete, we insert dummy edges with a very large cost ("Big-M") to make it complete.

Infeasible matchings are now feasible but they have very large cost. Maximum cardinality feasible matchings in the original graph correspond to matchings with the smallest number of dummy edges in the new graph.

Among them, optimality only depends on the costs of the original edges.



### The reformulated problem

After pre-processing, we can reformulate the problem as follows.

**Minimum Cost Bipartite Matching Problem (reformulated).** Find a minimum cost complete *matching* between the two vertex subsets of a given weighted bipartite graph.

Every solution is represented by an *assignment matrix* where S is the row set and T is the column set.

An assignment matrix is a binary square matrix with exactly one entry equal to 1 for each row and each column.

**Linear Assignment Problem.** Find a minimum cost assignment in a given square matrix.



#### Reformulation as a flow problem

A "trivial" way of solving the problem is to transform it into a min cost max flow problem.

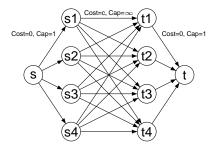


Figure: Network flow reformulation.



## A mathematical model (ILP)

We use a binary variable  $x_{ij}$  for each edge  $[i, j] \in \mathcal{E}$ , to indicate whether the edge is in the solution or not.

$$\begin{array}{ll} \text{minimize } \boldsymbol{z} = \sum_{i \in \mathcal{S}} \sum_{j \in \mathcal{T}} c_{ij} \boldsymbol{x}_{ij} \\ \text{s.t. } \sum_{j \in \mathcal{T}} \boldsymbol{x}_{ij} = 1 & \forall i \in \mathcal{S} \\ & \sum_{i \in \mathcal{S}} \boldsymbol{x}_{ij} = 1 & \forall j \in \mathcal{T} \\ & \boldsymbol{x}_{ij} \in \{0, 1\} & \forall i \in \mathcal{S}, \ \forall j \in \mathcal{T}. \end{array}$$

Does it have the integrality property?



## A mathematical model (ILP)

We use a binary variable  $x_{ij}$  for each edge  $[i, j] \in \mathcal{E}$ , to indicate whether the edge is in the solution or not.

$$\begin{array}{ll} \text{minimize } \mathbf{z} = \sum_{i \in \mathcal{S}} \sum_{j \in \mathcal{T}} c_{ij} \mathbf{x}_{ij} \\ \text{s.t. } \sum_{j \in \mathcal{T}} \mathbf{x}_{ij} = 1 & \forall i \in \mathcal{S} \\ & \sum_{i \in \mathcal{S}} \mathbf{x}_{ij} = 1 & \forall j \in \mathcal{T} \\ & \mathbf{x}_{ij} \in \{0, 1\} & \forall i \in \mathcal{S}, \ \forall j \in \mathcal{T}. \end{array}$$

- The constraint matrix is totally unimodular.
- The right-hand sides are integer numbers.

Hence all solutions of the linear relaxation have integer coordinates.



## A mathematical model (LP)

Relaxing the integrality constraints, the following model is obtained:

$$\begin{array}{ll} \text{minimize } \textbf{\textit{z}} = \sum_{i \in \mathcal{S}} \sum_{j \in \mathcal{T}} c_{ij} \textbf{\textit{x}}_{ij} \\ \text{s.t. } \sum_{j \in \mathcal{T}} \textbf{\textit{x}}_{ij} = 1 & \forall i \in \mathcal{S} \\ & \sum_{i \in \mathcal{S}} \textbf{\textit{x}}_{ij} = 1 & \forall j \in \mathcal{T} \\ & \textbf{0} \leq \textbf{\textit{x}}_{ij} \leq 1 & \forall i \in \mathcal{S}, \ \forall j \in \mathcal{T}. \end{array}$$

Upper bounds  $x_{ij} \leq 1$  are redundant because assignment constraints and non-negativity constraints imply them.



## A mathematical model (LP)

So we are left with the following model:

$$\begin{array}{ll} \text{minimize } \textbf{\textit{z}} = \sum_{i \in \mathcal{S}} \sum_{j \in \mathcal{T}} \textbf{\textit{c}}_{ij} \textbf{\textit{x}}_{ij} \\ \text{s.t. } \sum_{j \in \mathcal{T}} \textbf{\textit{x}}_{ij} = 1 & \forall i \in \mathcal{S} \\ & \sum_{i \in \mathcal{S}} \textbf{\textit{x}}_{ij} = 1 & \forall j \in \mathcal{T} \\ & \textbf{\textit{x}}_{ij} \geq 0 & \forall i \in \mathcal{S}, \ \forall j \in \mathcal{T}. \end{array}$$

This is a LP problem, hence it has a dual problem and it forms a *strong dual pair* with it.



### The primal-dual pair

Primal problem:

$$\begin{array}{ll} \text{minimize } \textbf{\textit{z}} = \sum_{i \in \mathcal{S}} \sum_{j \in \mathcal{T}} c_{ij} \textbf{\textit{x}}_{ij} \\ \text{s.t. } \sum_{j \in \mathcal{T}} \textbf{\textit{x}}_{ij} = \textbf{1} & \forall i \in \mathcal{S} \\ & \sum_{i \in \mathcal{S}} \textbf{\textit{x}}_{ij} = \textbf{1} & \forall j \in \mathcal{T} \\ & \textbf{\textit{x}}_{ij} \geq \textbf{0} & \forall i \in \mathcal{S}, \ \forall j \in \mathcal{T}. \end{array}$$

Write the dual.



### The primal-dual pair

Primal problem:

$$\begin{array}{ll} \text{minimize } \textbf{z} = \sum_{i \in \mathcal{S}} \sum_{j \in \mathcal{T}} c_{ij} \textbf{x}_{ij} \\ \text{s.t.} \ \sum_{j \in \mathcal{T}} \textbf{x}_{ij} = 1 & \forall i \in \mathcal{S} \\ & \sum_{i \in \mathcal{S}} \textbf{x}_{ij} = 1 & \forall j \in \mathcal{T} \\ & \textbf{x}_{ij} \geq 0 & \forall i \in \mathcal{S}, \ \forall j \in \mathcal{T}. \end{array}$$

Dual problem:

$$\begin{array}{ll} \text{maximize } \mathbf{w} = \sum_{i \in \mathcal{S}} \mathbf{u}_i + \sum_{j \in \mathcal{T}} \mathbf{v}_j \\ \text{s.t. } \mathbf{u}_i + \mathbf{v}_j \leq \mathbf{c}_{ij} & \forall i \in \mathcal{S} \; \forall j \in \mathcal{T}. \end{array}$$



#### The dual problem

$$\begin{array}{ll} \text{maximize } \mathbf{w} = \sum_{i \in \mathcal{S}} \mathbf{u}_i + \sum_{j \in \mathcal{T}} \mathbf{v}_j \\ \text{s.t. } \mathbf{u}_i + \mathbf{v}_j \leq \mathbf{c}_{ij} & \forall i \in \mathcal{S} \; \forall j \in \mathcal{T}. \end{array}$$

Dual variables u and v are unrestricted in sign.

The dual slack variables (primal reduced costs) are:

$$\overline{\mathbf{C}}_{ij}=\mathbf{C}_{ij}-\mathbf{U}_i-\mathbf{V}_j.$$

For optimality, complementary slackness conditions impose that:

$$\overline{c}_{ij}\mathbf{x}_{ij} = 0 \quad \forall i \in S \; \forall j \in \mathcal{T}.$$



### Partial assignments and primal feasibility

We call partial assignment an assignment satisfying

$$\sum_{j \in \mathcal{T}} \mathbf{x}_{ij} \leq \mathbf{1} \quad \forall i \in \mathcal{S} \ \sum_{i \in \mathcal{S}} \mathbf{x}_{ij} \leq \mathbf{1} \quad \forall j \in \mathcal{T}.$$

Primal infeasibility is measured by the number of missing assignments.

CSCs impose that in each primal/dual pair of base solutions we may have  $x_{ij} > 0$  only for edges [i, j] for which  $\overline{c}_{ij} = 0$ .

We call admissible cells of the assignment matrix those where  $\overline{c}_{ii} = 0$ .



### Primal-dual algorithms

A primal-dual algorithm solves linear programming problems exploiting duality theory and in particular the CSCs.

The algorithm is initialized with a dual feasible solution and a corresponding primal solution (in general, infeasible) satisfying the CSCs.

After every iteration the algorithm keeps a pair of primal (infeasible) and dual (feasible) solutions, satisfying the CSCs.

The algorithm alternates two types of iterations, and it monotonically decreases primal infeasibility until it achieves primal feasibility.

- Primal iteration: keeping the current dual feasible solution fixed, find a primal solution minimizing primal infeasibility among those satisfying the CSCs;
- Dual iteration: keeping the current primal solution fixed, modify the dual solution, keeping it feasible and the CSCs satisfied.

## Hungarian algorithm (Kuhn 1955)

The hungarian algorithm is a primal-dual algorithm.

- Primal iteration: keeping u<sub>i</sub> and v<sub>j</sub> fixed, and hence c
  <sub>ij</sub> fixed, determine x maximizing the number of assignments (x<sub>ij</sub> = 1), using only admissible cells;
- Dual iteration: update u<sub>i</sub> and v<sub>j</sub>, keeping c
  <sub>ij</sub> = 0 where x<sub>ij</sub> = 1 and making some inadmissible cells admissible.

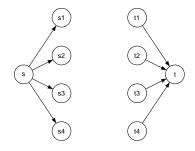


## Hungarian algorithm: pseudo-code

```
Begin
Step 1: Dual initialization of u and v;
Step 2: Primal initialization of x:
while (x is infeasible) do
  Step 3.1: Path initialization
  Path:=nil:
  while (Path = nil) do
    while (Path = nil) \land (L \neq \emptyset) do
       Step 3.2: Labeling procedure
    end while
    if Path = nil then
       Step 4: Dual iteration: Modify u and v;
    end if
  end while
  Step 5: Primal iteration: Modify x;
end while
End
```



#### Hungarian algorithm: visualization



C	t1	t2	t3	t4	u
s1	15	22	13	4	0
s2	12	21	15	7	0
s3	16	20	22	6	0
s4	6	11	8	5	0
V	0	0	0	0	0

The values in the bottom-right corners are:

	t1	t2	t3	t4	Mate
s1	15	22	13	4	nil
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	nil
Mate	nil	nil	nil	nil	0

Initially:  $\mathbf{x} = 0$ ,  $\mathbf{z} = 0$ , Card = 0.

$$\mathbf{Z} = \sum_{[i,j]\in\mathcal{E}} \mathbf{C}_{ij} \mathbf{X}_{ij}.$$

$$\mathbf{W} = \sum_{i \in S} \mathbf{U}_i + \sum_{j \in T} \mathbf{V}_j.$$

Initially: u = v = 0,  $\overline{c} = c$ , w = 0.

UNIVERSITÀ DEGLI STUDI DI MILANO

## Step 1: Dual initialization

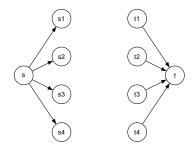
This can be done with a dual ascent procedure.

Begin Step 1 for  $i \in S$  do  $u_i := \min_{j \in T} \{ C_{ij} \}$ ; end for for  $j \in T$  do  $v_j := \min_{i \in S} \{ C_{ij} - u_i \}$ ; end for End Step 1 The dual variables are raised one at a time from 0 up to the minimum value that makes a dual constraint active.

This guarantees that the dual solution remains feasible.

**Complexity:**  $O(n^2)$ .



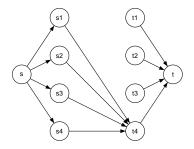


C	t1	t2	t3	t4	u
s1	15	22	13	4	0
s2	12	21	15	7	0
s3	16	20	22	6	0
s4	6	11	8	5	0
V	0	0	0	0	0

	t1	t2	t3	t4	Mate
s1	15	22	13	4	nil
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	nil
Mate	nil	nil	nil	nil	0

$$x = 0, Card = 0.$$

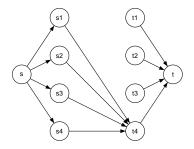




C	t1	t2	t3	t4	u
s1	11	18	9	0	4
s2	5	14	8	0	7
s3	10	14	16	0	6
s4	1	6	3	0	5
V	0	0	0	0	22

	t1	t2	t3	t4	Mate
s1	15	22	13	4	nil
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	nil
Mate	nil	nil	nil	nil	0



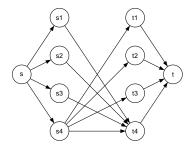


C	t1	t2	t3	t4	и
s1	11	18	9	0	4
s2	5	14	8	0	7
s3	10	14	16	0	6
s4	1	6	3	0	5
V	0	0	0	0	22

	t1	t2	t3	t4	Mate
s1	15	22	13	4	nil
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	nil
Mate	nil	nil	nil	nil	0

$$x = 0, Card = 0.$$





C	t1	t2	t3	t4	u
s1	10	12	6	0	4
s2	4	8	5	0	7
s3	9	8	13	0	6
s4	0	0	0	0	5
V	1	6	3	0	32

	t1	t2	t3	t4	Mate
s1	15	22	13	4	nil
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	nil
Mate	nil	nil	nil	nil	0



UNIVERSITÀ DEGLI STUDI DI MILANO

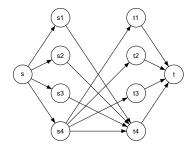
## Step 2: Primal initialization

```
Begin Step 2
for k \in S \cup T do
  Mate(k):=nil;
end for
Card := 0;
while (\exists [i, j] : (c_{ij} - u(i) - v(j) = 0) \land (Mate(i) = nil) \land (Mate(j) = nil))
do
  x_{ii} := 1;
  Card := Card + 1:
  Mate(i) := i; Mate(i) := i;
end while
End Step 2
```

A maximal partial matching is computed, using only admissible cells. This requires scanning a square  $(n \times n)$  matrix.

**Complexity:**  $O(n^2)$ .





t2

22

21

t1

15

12

s1

s2

C	t1	t2	t3	t4	u
s1	10	12	6	0	4
s2	4	8	5	0	7
s3	9	8	13	0	6
s4	0	0	0	0	5
V	1	6	3	0	32

There are 7 admissible cells.

Scanning them in lexicographic order by rows and columns, edge [1,4] is chosen first.



s3	16	20	22	6	nil
s4	6	11	8	5	nil
Mate	nil	nil	nil	nil	0

t3

13

15

t4

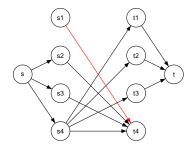
4

7

Mate

nil

nil

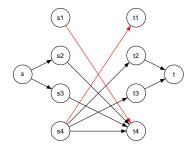


C	t1	t2	t3	t4	u
s1	10	12	6	0	4
s2	4	8	5	0	7
s3	9	8	13	0	6
s4	0	0	0	0	5
V	1	6	3	0	32

We still have three admissible cells: edge [4, 1] is chosen next.

	t1	t2	t3	t4	Mate
s1	15	22	13	4	4
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	nil
Mate	nil	nil	nil	1	4





	t1	t2	t3	t4	Mate
s1	15	22	13	4	4
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	1
Mate	4	nil	nil	1	10

C	t1	t2	t3	t4	u
s1	10	12	6	0	4
s2	4	8	5	0	7
s3	9	8	13	0	6
s4	0	0	0	0	5
V	1	6	3	0	32

No admissible cells are left on unmatched rows and columns. The current partial matching is maximal.



$$x_{14} = x_{41} = 1$$
, Card = 2.

#### Primal feasibility test

It consists of counting how many edges have been inserted into the primal solution (partial matching).

Primal feasibility test: ??.



#### Primal feasibility test

It consists of counting how many edges have been inserted into the primal solution (partial matching).

Primal feasibility test: Card = n.



## Step 3: Search for an augmenting path

Step 3 consists of searching for an *augmenting path*, which is also an *alternating path* since the graph is bipartite. This is a path a unit of flow can follow to go from s to t.



## Step 3: Search for an augmenting path

Step 3 consists of searching for an *augmenting path*, which is also an *alternating path* since the graph is bipartite. This is a path a unit of flow can follow to go from s to t.

Every time an *s*-*t* path is found, the cardinality of the current partial matching can be increased by 1 (primal iteration). To find the *s*-*t* path it may be necessary to execute at most O(n) dual iterations, because each of them allows to reach one more node in T.



## Step 3: Search for an augmenting path

Step 3 consists of searching for an *augmenting path*, which is also an *alternating path* since the graph is bipartite. This is a path a unit of flow can follow to go from s to t.

Every time an *s*-*t* path is found, the cardinality of the current partial matching can be increased by 1 (primal iteration). To find the *s*-*t* path it may be necessary to execute at most O(n) dual iterations, because each of them allows to reach one more node in T.

The path starts from *s*; every node in S and T that can be reached is labeled. *Label* of a node is its predecessor. *L* is the set of labels to be used to generate others.

Vector *p* stores the minimum reduced cost value for each unlabeled column among those in labeled rows. Vector  $\pi$  stores the corresponding row.

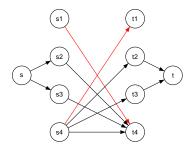


Step 3.1: Path initialization

Begin Step 3.1  $L:=\emptyset$ : for  $k \in S \cup T$  do Label(k):=nil;end for for  $j \in \mathcal{T}$  do  $p(j):=\infty; \pi(j):=nil;$ end for for  $i \in S$  : (*Mate*(*i*) = *nil*) do Label(i):=s; $L:=L \cup \{i\};$ for  $j \in \mathcal{T}$ : (Label(j) = nil) do if c(i, j) - u(i) - v(j) < p(j) then  $p(i):=c(i, j) - u(i) - v(j); \pi(j):=i;$ end if end for end for End Step 3.1

**Complexity:**  $O(n^2)$ .





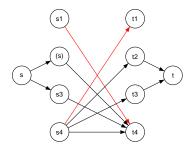
ſ	C	t1	t2	t3	t4	и
	s1	10	12	6	0	4
	s2	4	8	5	0	7
	s2 s3 s4	9	8	13	0	6
	s4	0	0	0	0	5
	V	1	6	3	0	32
	р	$\infty$	$\infty$	$\infty$	$\infty$	
	$\pi$	nil	nil	nil	nil	

	t1	t2	t3	t4	Mate
s1	15	22	13	4	4
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	1
Mate	4	nil	nil	1	10

There are two unmatched nodes in  $\ensuremath{\mathcal{S}}.$ 



$$x_{14} = x_{41} = 1$$
, Card = 2.

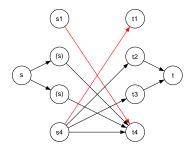


C	t1	t2	t3	t4	u
s1	10	12	6	0	4
s2	4	8	5	0	7
s3	9	8	13	0	6
s4	0	0	0	0	5
V	1	6	3	0	32
р	4	8	5	0	
$\pi$	s2	s2	s2	s2	

	t1	t2	t3	t4	Mate
s1	15	22	13	4	4
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	1
Mate	4	nil	nil	1	10

Insert  $L = \{s2\}$ .

$$x_{14} = x_{41} = 1$$
, Card = 2.



C	t1	t2	t3	t4	u
s1	10	12	6	0	4
s2	4	8	5	0	7
s3	9	8	13	0	6
s4	0	0	0	0	5
V	1	6	3	0	32
р	4	8	5	0	
$\pi$	s2	s2	s2	s2	

	t1	t2	t3	t4	Mate
s1	15	22	13	4	4
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	1
Mate	4	nil	nil	1	10

Insert: 
$$L = \{s2, s3\}$$
.



$$x_{14} = x_{41} = 1$$
, Card = 2.

### Step 3.2: Label propagation

```
Begin Step 3.2
Extract k from L;
if k \in S then
  Step 3.2.A: Propagation from k \in S to T
else
  if (Mate(k) \neq nil) then
    Step 3.2.B: Propagation from k \in \mathcal{T} to S
  else
    Path := k:
  end if
end if
End Step 3.2
```

Propagation stops when an unmatched node  $k \in T$  is labeled.

Each node is inserted/extracted in/from L at most once.



# Step 3.2.A: Label propagation from ${\cal S}$ to ${\cal T}$

Begin Step 3.2.A for  $j \in \mathcal{T}$ :  $(Label(j) = nil) \land (c(k, j) - u(k) - v(j) = 0)$  do Label(j):=k; $L := L \cup \{j\};$ end for End Step 3.2.A

Propagation from  $k \in S$  to T occurs along edges that:

- correspond to admissible cells;
- do not belong to the current partial matching.

**Complexity:** O(n).



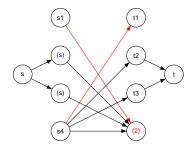
# Step 3.2.B: Label propagation from ${\mathcal T}$ to ${\mathcal S}$

```
Begin Step 3.2.B
if (Label(Mate(k)) = nil) then
  Label(Mate(k)):=k;
  L := L \cup \{Mate(k)\};
  for j \in \mathcal{T}: (Label(j) = nil) do
    if c(Mate(k), j) - u(Mate(k)) - v(j) < p(j) then
       p(j) := c(Mate(k), j) - u(Mate(k)) - v(j);
       \pi(i) := Mate(k);
    end if
  end for
end if
End Step 3.2.B
```

Propagation from  $k \in T$  to S occurs along edges of the partial matching.



**Complexity:** O(n).



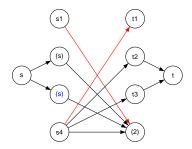
C	t1	t2	t3	t4	u
s1	10	12	6	0	4
s2	4	8	5	0	7
s3	9	8	13	0	6
s4	0	0	0	0	5
V	1	6	3	0	32
р	4	8	5	0	
$\pi$	s2	s2	s2	s2	

	t1	t2	t3	t4	Mate
s1	15	22	13	4	4
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	1
Mate	4	nil	nil	1	10

Extract: 
$$L = \{s2, s3\}$$
.  
Insert:  $L = \{s3, t4\}$ .



$$x_{14} = x_{41} = 1$$
, Card = 2.



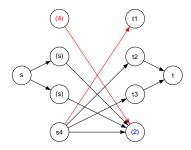
C	t1	t2	t3	t4	u
s1	10	12	6	0	4
s2	4	8	5	0	7
s3	9	8	13	0	6
s4	0	0	0	0	5
V	1	6	3	0	32
р	4	8	5	0	
$\pi$	s2	s2	s2	s2	

	t1	t2	t3	t4	Mate
s1	15	22	13	4	4
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	1
Mate	4	nil	nil	1	10

Extract:  $L = \{s3, t4\}$ . Insert:  $L = \{t4\}$ .



$$x_{14} = x_{41} = 1$$
, Card = 2.



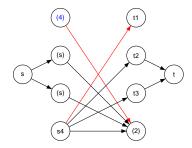
C	t1	t2	t3	t4	u
s1	10	12	6	0	4
s2	4	8	5	0	7
s3	9	8	13	0	6
s4	0	0	0	0	5
V	1	6	3	0	32
р	4	8	5	0	
$\pi$	s2	s2	s2	s2	

	t1	t2	t3	t4	Mate
s1	15	22	13	4	4
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	1
Mate	4	nil	nil	1	10

Extract: 
$$L = \{t4\}$$
.  
Insert:  $L = \{s1\}$ .



$$x_{14} = x_{41} = 1$$
, Card = 2.



	t1	t2	t3	t4	Mate
s1	15	22	13	4	4
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	1
Mate	4	nil	nil	1	10

C	t1	t2	t3	t4	u
s1	10	12	6	0	4
s2	4	8	5	0	7
s3	9	8	13	0	6
s4	0	0	0	0	5
V	1	6	3	0	32
р	4	8	5	0	
$\pi$	s2	s2	s2	s2	

Extract:  $L = \{s1\}$ . Insert:  $L = \{\}$ .

No s-t path has been found. Nodes s1, s2, s3, t4 are labeled. Nodes s4, t1, t2, t3 are not.



# Step 4: Dual iteration

#### **Begin Step 4**

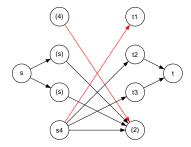
 $\delta := \min_{j \in \mathcal{T}} \{ p(j) : Label(j) = nil \};$ for  $i \in S$  : Label(i)  $\neq$  nil do  $u(i) := u(i) + \delta;$ end for for  $j \in \mathcal{T}$  : Label(j)  $\neq$  nil do  $\mathbf{v}(i) := \mathbf{v}(i) - \delta;$ end for for  $i \in \mathcal{T}$ : Label(i) = nil do  $p(i) := p(i) - \delta;$ end for for  $j \in \mathcal{T}$ :  $(Label(j) = nil) \land (p(j) = 0)$  do *Label*(*j*) :=  $(\pi(j))$ ;  $L := L \cup \{j\};$ end for End Step 4

The value  $\delta$  is the minimum reduced cost in the sub-matrix of labeled rows and unlabeled columns. However, owing to the vector *p*, finding  $\delta$  takes O(n) instead of  $O(n^2)$ .

Updating u takes O(n). Updating v takes O(n). Updating p takes O(n). Re-initializing L takes O(n).

At least one more cell becomes admissible and it is used to label one more node in  $\mathcal{T}$ .





C	t1	t2	t3	t4	u
s1	10	12	6	0	4
s2	4	8	5	0	7
s3	9	8	13	0	6
s4	0	0	0	0	5
V	1	6	3	0	32
р	4	8	5	0	
$\pi$	s2	s2	s2	s2	

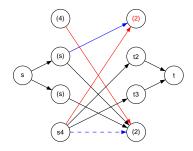
	t1	t2	t3	t4	Mate
s1	15	22	13	4	4
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	1
Mate	4	nil	nil	1	10

We consider edges joining labeled nodes in  ${\mathcal S}$  with unlabeled nodes in  ${\mathcal T}.$ 

We find  $\delta = 4$ .



$$x_{14} = x_{41} = 1$$
, Card = 2.



	t1	t2	t3	t4	Mate
s1	15	22	13	4	4
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	1
Mate	4	nil	nil	1	10

$$x_{14} = x_{41} = 1$$
, Card = 2.

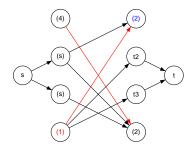
ſ	C	t1	t2	t3	t4	и
	s1	6	8	2	0	8
	s2	0	4	1	0	11
	s3	5	4	9	0	10
	s4	0	0	0	4	5
	V	1	6	3	-4	<b>40</b>
	р	0	4	1	0	
	$\pi$	s2	s2	s2	s2	

Increase  $u_1$ ,  $u_2$  and  $u_3$  by  $\delta$ . Decrease  $v_4$  by  $\delta$ . Decrease  $p_1$ ,  $p_2$  and  $p_3$  by  $\delta$ .

Cell [4,4] is no longer admissible. Cell [2,1] becomes admissible.

Label *t*1 from s2. Re-initialize:  $L = \{t1\}$ .



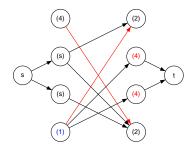


C	t1	t2	t3	t4	u
s1	6	8	2	0	8
s2	0	4	1	0	11
s3	5	4	9	0	10
s4	0	0	0	4	5
V	1	6	3	-4	<b>40</b>
р	0	0	0	0	
$\pi$	s2	s4	s4	s2	

	t1	t2	t3	t4	Mate
s1	15	22	13	4	4
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	1
Mate	4	nil	nil	1	10

Extract: 
$$L = \{t1\}$$
.  
Insert:  $L = \{s4\}$ .





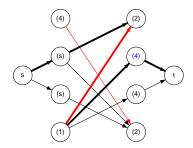
C	t1	t2	t3	t4	u
s1	6	8	2	0	8
s2	0	4	1	0	11
s3	5	4	9	0	10
s4	0	0	0	4	5
V	1	6	3	-4	40
р	0	0	0	0	
$\pi$	s2	s4	s4	s2	

	t1	t2	t3	t4	Mate
s1	15	22	13	4	4
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	1
Mate	4	nil	nil	1	10

Extract:  $L = \{s4\}$ . Insert:  $L = \{t2, t3\}$ .



$$x_{14} = x_{41} = 1$$
, Card = 2.



Γ	C	t1	t2	t3	t4	и
ſ	s1	6	8	2	0	8
	s2	0	4	1	0	11
	s3	5	4	9	0	10
	s4	0	0	0	4	5
Γ	V	1	6	3	-4	<b>40</b>
	р	0	0	0	0	
	$\pi$	s2	s4	s4	s2	

	t1	t2	t3	t4	Mate
s1	15	22	13	4	4
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	1
Mate	4	nil	nil	1	10

Extract:  $L = \{t^2, t^3\}$ .

*t*2 is not matched: an *s*-*t* path has been found.



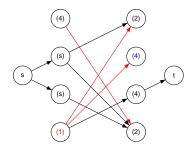
# Step 5: primal iteration

```
Begin Step 5
i := Path;
repeat
   i := Label(j);
   Mate(j) := i; Mate(i) := j;
   x_{ii} := 1; z := z + c_{ii};
   Card := Card + 1:
  i := Label(i);
   if Label(i) \neq s then
      \mathbf{x}_{ii} := 0; \mathbf{z} := \mathbf{z} - \mathbf{c}_{ii};
      Card := Card - 1:
   end if
until (i = s);
End Step 5
```

The path is reconstructed backward from t to s. It has O(n) edges.

**Complexity:** O(n).

UNIVERSITÀ DEGLI STUDI DI MILANO

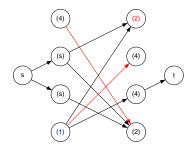


C	t1	t2	t3	t4	u
s1	6	8	2	0	8
s2	0	4	1	0	11
s3	5	4	9	0	10
s4	0	0	0	4	5
V	1	6	3	-4	<b>40</b>
р	0	0	0	0	
$\pi$	s2	s4	s4	s2	



	t1	t2	t3	t4	Mate
s1	15	22	13	4	4
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	2
Mate	4	4	nil	1	21

$$x_{14} = x_{41} = x_{42} = 1$$
, Card = 3.



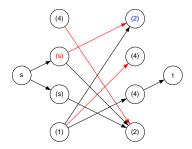
C	t1	t2	t3	t4	u
s1	6	8	2	0	8
s2	0	4	1	0	11
s3	5	4	9	0	10
s4	0	0	0	4	5
V	1	6	3	-4	<b>40</b>
р	0	0	0	0	
$\pi$	s2	s4	s4	s2	

The pro	edecessor	of	s4	is	<i>t</i> 1.
---------	-----------	----	----	----	-------------



	t1	t2	t3	t4	Mate
s1	15	22	13	4	4
s2	12	21	15	7	nil
s3	16	20	22	6	nil
s4	6	11	8	5	2
Mate	nil	4	nil	1	15

$$x_{14} = x_{42} = 1$$
, Card = 2.

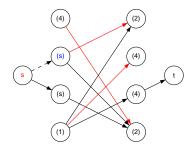


C	t1	t2	t3	t4	u
s1	6	8	2	0	8
s2	0	4	1	0	11
s3	5	4	9	0	10
s4	0	0	0	4	5
V	1	6	3	-4	<b>40</b>
р	0	0	0	0	
$\pi$	s2	s4	s4	s2	



	t1	t2	t3	t4	Mate
s1	15	22	13	4	4
s2	12	21	15	7	1
s3	16	20	22	6	nil
s4	6	11	8	5	2
Mate	2	4	nil	1	27

$$x_{14} = x_{42} = x_{21} = 1$$
, Card = 3.



C	t1	t2	t3	t4	и
s1	6	8	2	0	8
s2	0	4	1	0	11
s3	5	4	9	0	10
s4	0	0	0	4	5
V	1	6	3	-4	<b>40</b>
р	0	0	0	0	
$\pi$	s2	s4	s4	s2	

	t1	t2	t3	t4	Mate
s1	15	22	13	4	4
s2	12	21	15	7	1
s3	16	20	22	6	nil
s4	6	11	8	5	2
Mate	2	4	nil	1	27

 $x_{14} = x_{42} = x_{21} = 1$ , Card = 3.

The predecessor of s2 is s.

The primal solution has been updated.

Card < n.

We are ready for another stage.



# Hungarian algorithm: complexity

#### Begin

Step 1: Dual initialization; Step 2: Primal initialization; while [1] (x is infeasible) do Step 3.1: Initialization Path:=nil: while [2] (Path = nil) do while [3] (*Path* = *nil*)  $\land$  ( $L \neq \emptyset$ ) do Step 3.2: Labeling procedure end while if Path = nil then Step 4: Dual iteration: end if end while Step 5: Primal iteration: end while End

Step 1:  $O(n^2)$ . Step 2:  $O(n^2)$ . Loop 1: O(n) times. Step 3.1: O(n). Loop 2 (stage): O(n) times. Loop 3: O(n) times. Step 3.2:  $O(n) \forall$  node, i.e.  $O(n^2) \forall$  stage. Step 4: O(n). Step 5: O(n).

#### **Overall complexity:** $O(n^3)$ .

