

Dual problems in linear programming

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Outline

- Dual LP-problems
- Maximum-weight matching
- Assignment problem
- Minimum-cost flow

Dual LP-problems

Initial problem

Objective function

- $L(X) = \sum_{i=1}^n c_i x_i \rightarrow \max$

Subject to

- $\sum_{i=1}^n a_{j,i} x_i \leq b_j, j = \overline{1, k};$
- $\sum_{i=1}^n a_{j,i} x_i = b_j, j = \overline{k+1, m};$
- $x_i \geq 0, i = \overline{1, l}.$

Dual problem

Objective function

- $l(Y) = \sum_{j=1}^m b_j y_j \rightarrow \min$

Subject to

- $\sum_{j=1}^m a_{j,i} y_j \geq c_i, i = \overline{1, l};$
- $\sum_{j=1}^m a_{j,i} y_j = c_i, i = \overline{l+1, n};$
- $y_j \geq 0, j = \overline{1, k}.$

Dual LP-problems

- **Complementary slackness conditions**

$$\begin{aligned}
 L(X) &= \sum_{i=1}^n c_i x_i = \sum_{i=1}^l c_i x_i + \sum_{i=l+1}^n c_i x_i \leq \\
 &\leq \sum_{i=1}^l x_i \sum_{j=1}^m a_{i,j} y_j + \sum_{i=l+1}^n x_i \sum_{j=1}^m a_{i,j} y_j = \\
 &= \sum_{j=1}^k y_j \sum_{i=1}^n a_{j,i} x_i + \sum_{j=k+1}^m y_j \sum_{i=1}^n a_{j,i} x_i \leq \\
 &\leq \sum_{j=1}^k b_j y_j + \sum_{j=k+1}^m b_j y_j = \sum_{j=1}^m b_j y_j = l(Y)
 \end{aligned}$$

So, $L(X) \leq l(Y)$ and

$L(X) = l(Y)$ iff **(all summands are non-negative!)**:

(a) $\sum_{i=1}^l x_i (\sum_{j=1}^m a_{i,j} y_j - c_i) = 0;$

(b) $\sum_{j=1}^k y_j (b_j - \sum_{i=1}^n a_{j,i} x_i) = 0.$

Dual LP-problems

- **Complementary slackness conditions**

So, $L(X) = l(Y)$ iff :

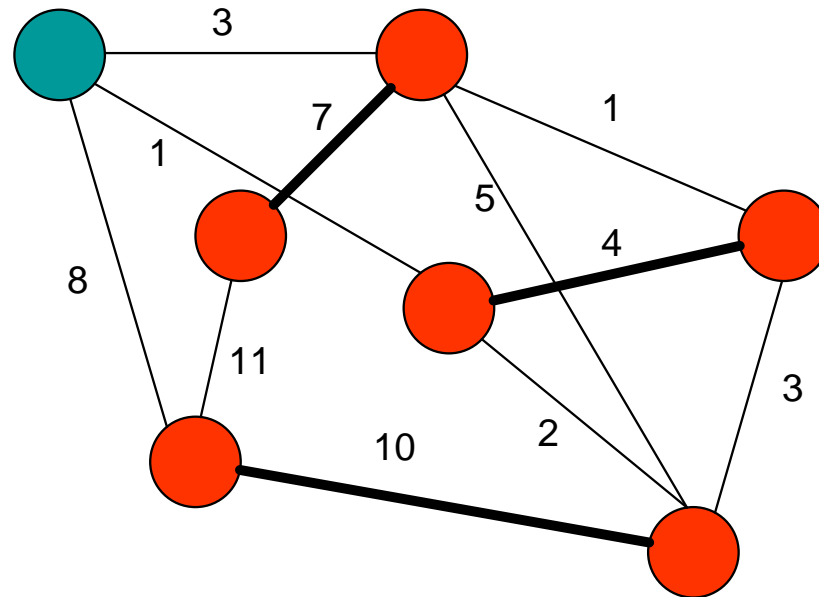
(a) For all $i = \overline{1, l}$, either $x_i = 0$ or the corresponding condition in the dual problem turns into the equality $\sum_{j=1}^m a_{j,i} y_j = c_i$;

(b) For all $j = \overline{1, k}$, either $y_j = 0$ or the corresponding condition in the initial problem turns into the equality $\sum_{i=1}^n a_{j,i} x_i = b_j$.

To solve the LP problem, we can find values (X, Y) , which satisfy the complementary slackness conditions. These values provide the maximum value of the objective function $L(X)$ and the minimum value of the objective function $l(Y)$.

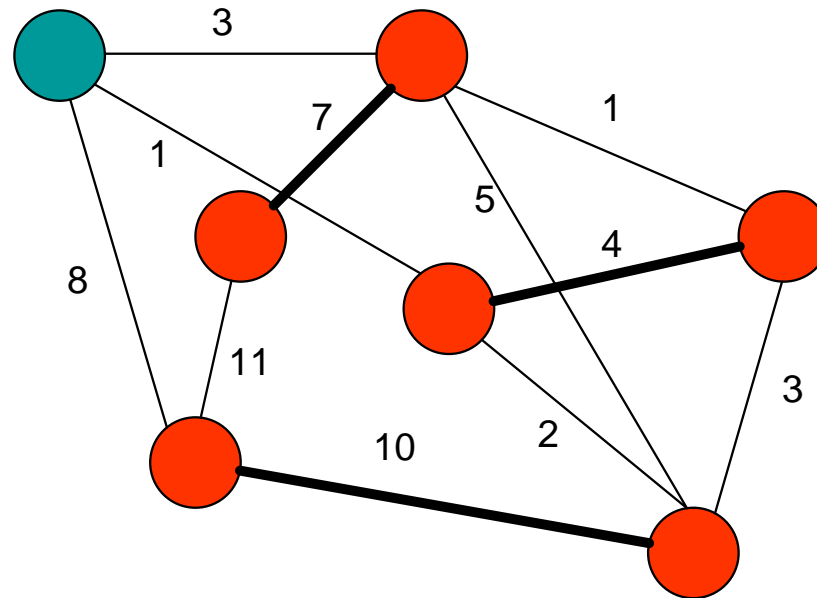
Maximum-weight matching

- **Maximum-weight matching** in a weighted graph $G(V,E)$ is a matching with the maximum sum of the edge weights.



Maximum-weight matching

- **Vertices – people.**
- Two persons, working in pair, produce a certain amount of a product. How to match people, to produce the greatest amount of product?



Maximum-weight matching

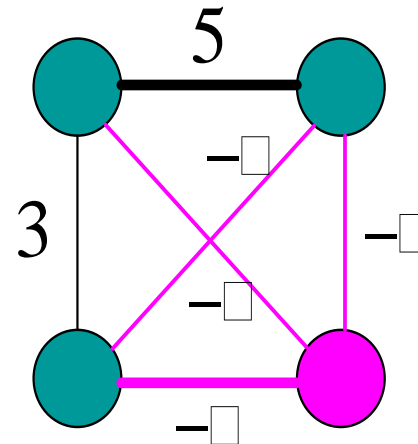
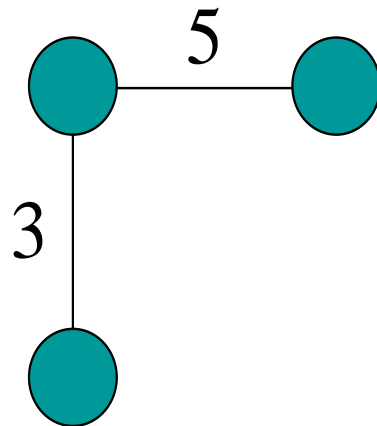
- **Problem statement**

$$\begin{aligned} & \sum_{j=1}^q c_j \xi_j \rightarrow \max; \\ & \sum_{j=1}^q I_{kj} \xi_j \leq 1, \quad \forall k = 1, \dots, p; \\ & \xi_j \in \{0, 1\}. \end{aligned}$$

Maximum-weight matching

The problem can be transformed into the perfect maximum-weight matching problem:

- add artificial edges with the weight $-\infty$;
- add an artificial vertex if the number of vertices is odd.

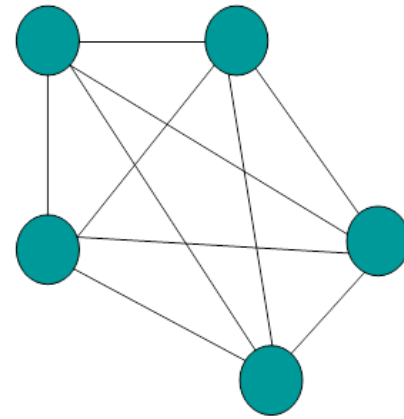


Maximum-weight matching

- **Problem statement**
- Solutions: $\xi_j \in \{0, 1, 1/2\}$

$$\sum_{j=1}^q c_j \xi_j \rightarrow \max;$$
$$\sum_{j=1}^q I_{kj} \xi_j = 1, \quad \forall k = 1, \dots, p;$$
$$\xi_j \in \{0, 1\}.$$

$G(V_r, E_r): V_r = 2p_r + 1$



Maximum-weight matching

- **Problem statement (non-discrete!)**

$$\sum_{j=1}^q c_j \xi_j \rightarrow \max;$$

$$(a) \sum_{j \in T_k} \xi_j = 1, \quad k = \overline{1, p}, \quad T_k = \{j = (i, k)\};$$

$$(b) \sum_{j \in E_r} \xi_j \leq p_r, \quad V_r \subseteq V, \quad |V_r| = 2p_r + 1;$$

$$\xi_j \geq 0, \quad j = \overline{1, q}.$$

- (a) every vertex is covered by an edge;
- (b) in every odd-cardinality set of vertices, every vertex is covered by at most one edge.

Maximum-weight matching

Dual problem statement

- π_i – variables associated with vertices
- λ_r – variables associated with odd-cardinality sets of vertices

$$\sum_{j=1}^p \pi_j + \sum_{V_r} p_r \lambda_r \rightarrow \min;$$
$$\pi_i + \pi_k + \sum_{V_r \in F_j} \lambda_r \geq c_j, \quad \forall j = 1, \dots, q, \quad e_j = (v_i, v_k), \quad F_j = \{V_r | e_j \in E_r\};$$
$$\lambda_j \geq 0, \quad \forall V_r.$$

Maximum-weight matching

Complementary slackness conditions

- For every $\xi_j, j = (i, k)$ either $\xi_j = 0$, or the equality holds in the corresponding constraint $\pi_i + \pi_k + \sum_{V_r \in F_j} \lambda_r = c_j$ (the edge $j = (i, k)$ can belong to a matching only if the equality holds);
- For every λ_r , either $\lambda_r = 0$, or the equality holds in the corresponding constraint $\sum_{j \in E_r} \xi_j = p_r$ ($\lambda_r > 0$ only if the odd-cardinality set of vertices has a maximum-cardinality matching).

Maximum-weight matching

- **Start. Choose initial values of the variables.**
- For every λ_r , $\lambda_r = 0$;
- For every edge $j = (i, k)$, $\pi_i + \pi_k \geq c_j$.
- Equality partial graph $G' = (V, E')$: $E' = \{j = (i, k), \pi_i + \pi_k = c_j\}$.
- **An optimal matching contains only edges from the EPG.**

Maximum-weight matching

- Let the current graph (after a number of blossom shrinkings), be G_{f-1} and its equality partial graph be G'_f . In G'_f , start growing an alternating tree T until the tree either:
 - A. blossoms;
 - B. augments;
 - C. becomes Hungarian.

Maximum-weight matching

- **Case A.** In this case the blossom is shrunk thus obtaining a new graph G_f and its equality partial graph G'_{f+1} .
- The shrinking of a blossom leaves the remaining T with the correct structure of an alternating tree.
- T could be retained and growing of the tree can continue.

Maximum-weight matching

- **Case B.** In this case a better matching with larger cardinality is obtained.
- T must now be discarded and a new tree grown in the same G'_f by choosing some remaining unsaturated vertex of G'_f as the new root.
- Once T is discarded and a new tree is started to be grown in G'_f some of the pseudo-vertices of G'_f (formed by the shrinking of earlier blossoms) can now appear labeled inner in the new tree.

Maximum-weight matching

- **Case C.** In this case, the weight vector $[\pi_i, \lambda_r]$ is changed for the current graph G_{f+1} so that a new EPG is obtained. The changes in $[\pi_i, \lambda_r]$ are chosen so that the new G'_f will:
 - (a) Continue to satisfy the complementary slackness conditions; and
 - (b) Either:
 - the current alternating tree in the old G'_f can be grown further, or it blossoms, or augments—using new links that enter the new G'_f ;
 - Or: some pseudo-vertex of the current alternating tree which was labelled inner is disposed of;
 - Or: we prove that no perfect matching exists in G_0 .

Maximum-weight matching

- For the links $j = (i, k)$ in G_{f+1} but not in G'_f , with one terminal vertex in the current alternating tree T and labeled **outer**, and the other terminal vertex **not in T** calculate:

$$\Delta_1 = \min_j \{ \pi_i + \pi_k - c_j \}$$

- Then,

$$\Delta = \min\{\Delta_1, \Delta_2, \Delta_3\}$$

Maximum-weight matching

- For the links $j = (i, k)$ in G_{f+1} but not in G'_f , with both terminal vertices in T and both labelled **outer** calculate:

$$\Delta_2 = \frac{1}{2} \min_j \{ \pi_i + \pi_k - c_j \}$$

Maximum-weight matching

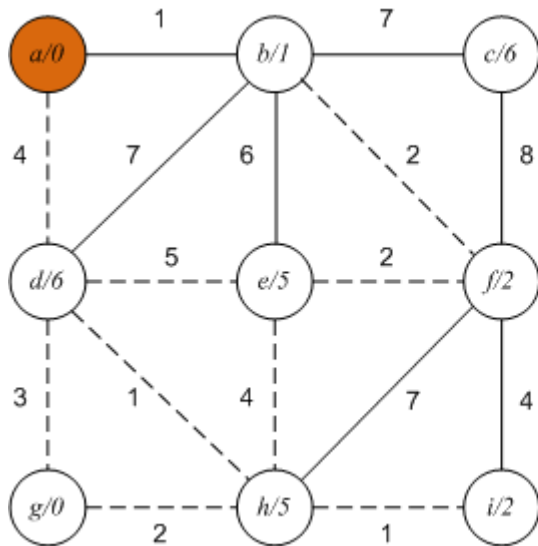
- For the sets S_r of vertices of G forming an outermost pseudo-vertex of T labelled inner calculate:

$$\Delta_3 = \frac{1}{2} \min_{S_r} \{\lambda_r\}$$

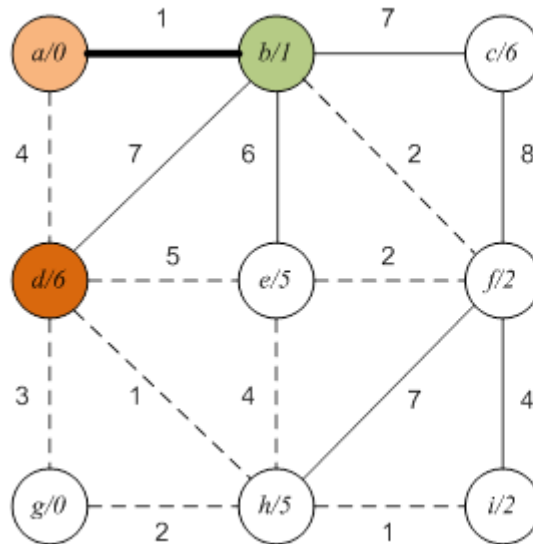
Maximum-weight matching

- For every vertex k of G which is an **outer** vertex of T or is contained in a **pseudo-vertex of T labelled outer**, decrease π_k to $\pi_k - \Delta$.
- For every vertex k of G which is an **inner** vertex of T or is contained in a **pseudo-vertex of T labelled inner**, increase π_k to $\pi_k + \Delta$.
- For every set S_r of vertices of G which forms an **outermost pseudo-vertex of T labelled outer**, increase λ_r to $\lambda_r + 2\Delta$.
- For every set S_r of vertices of G which forms an **outermost pseudo-vertex of T labelled inner** decrease λ_r to $\lambda_r - 2\Delta$.

Maximum-weight matching



a – root
ab – augmenting path



d - root

Hungarian tree

D1:

(d,e): $6+5-5=6$

(d,h): $6+5-1=10$

(d,g): $6+0-3=2$ min

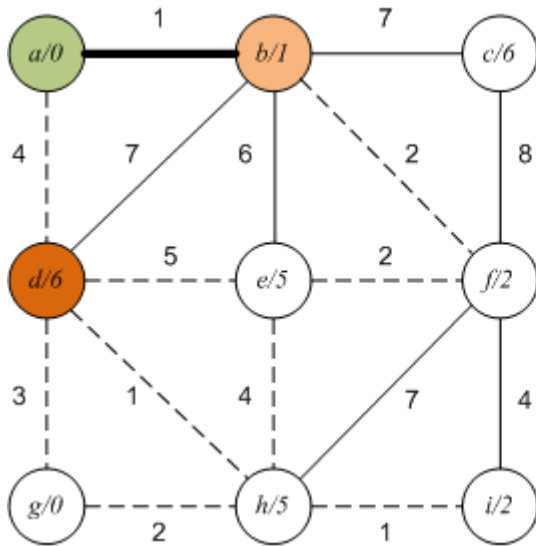
D2:

(a,d): $(0+6-4)/2=1$ min

D3: no

D=1=D2

Maximum-weight matching



d - root

Hungarian tree

D1:

(d,e): $6+5-5=6$

(d,h): $6+5-1=10$

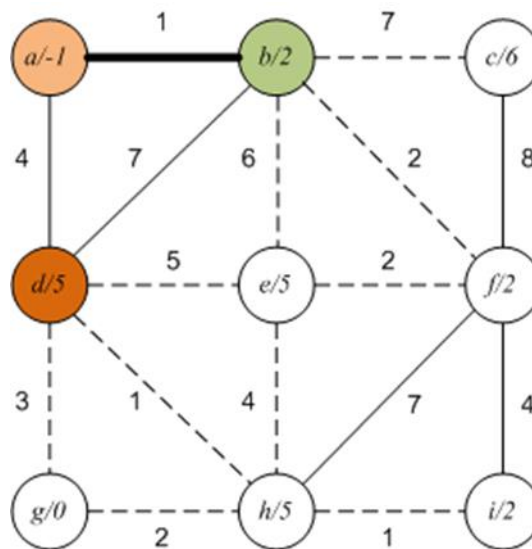
(d,g): $6+0-3=2$ min

D2:

(a,d): $(0+6-4)/2=1$ min

D3: no

$D=1=D2$

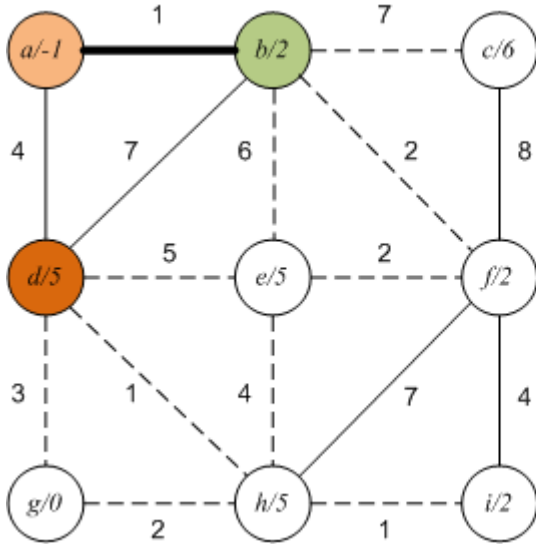


d - root

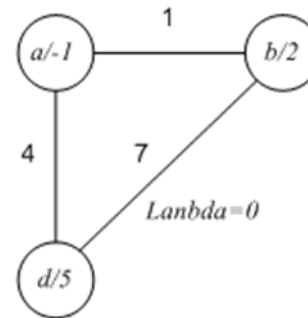
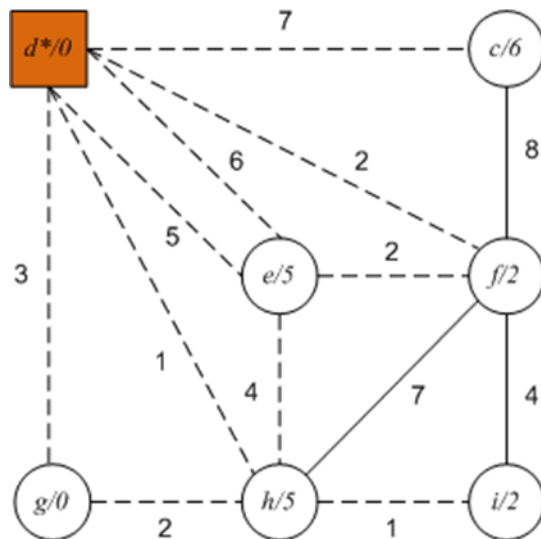
dba - blossom

$\text{Lambda}(dba) = 0$

Maximum-weight matching



d - root
 dba - blossom
 $\text{Lambda}(\text{dba}) = 0$



d* - root

Hungarian tree

D1:

(d,g): $5+0-3=2$

(d,h): $5+5-1=9$

(d,e): $5+5-5=5$

(b,e): $2+5-6=1$ min

(b,f): $2+2-2=2$

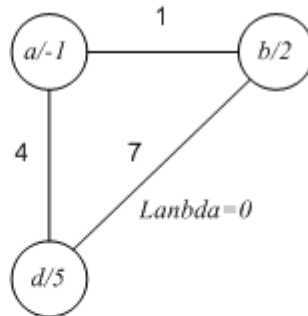
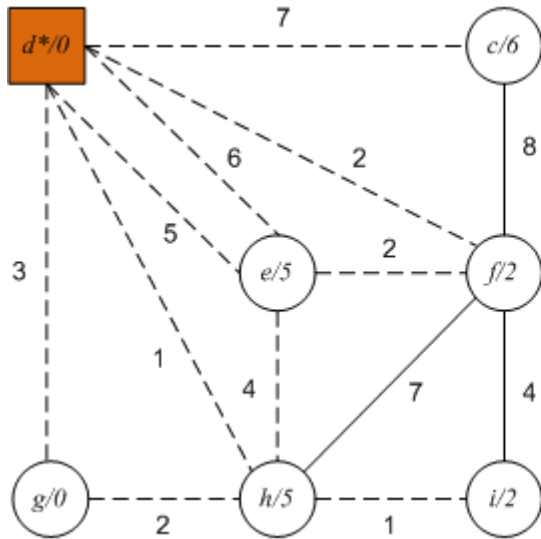
(b,c): $2+6-7=1$ min

D2: no

D3: no

D=1=D1

Maximum-weight matching



d^* - root

Hungarian tree

D1:

$$(d,g): 5+0-3=2$$

$$(d,h): 5+5-1=9$$

$$(d,e): 5+5-5=5$$

$$(b,e): 2+5-6=1 \text{ min}$$

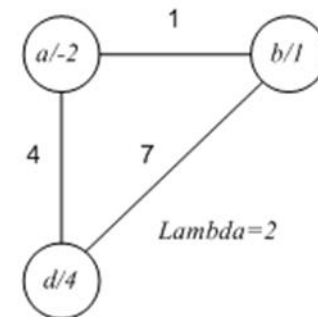
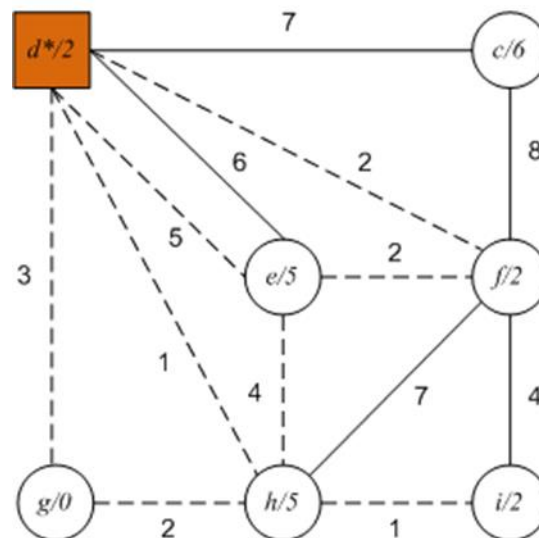
$$(b,f): 2+2-2=2$$

$$(b,c): 2+6-7=1 \text{ min}$$

D2: no

D3: no

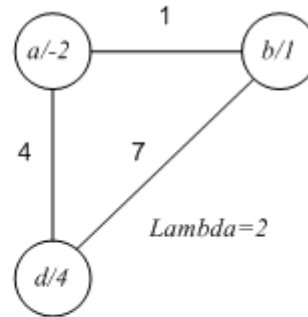
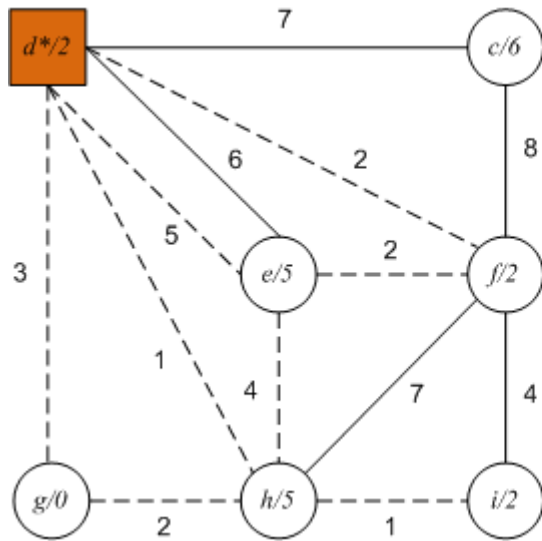
$D=1=D1$



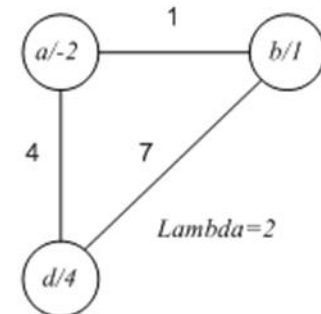
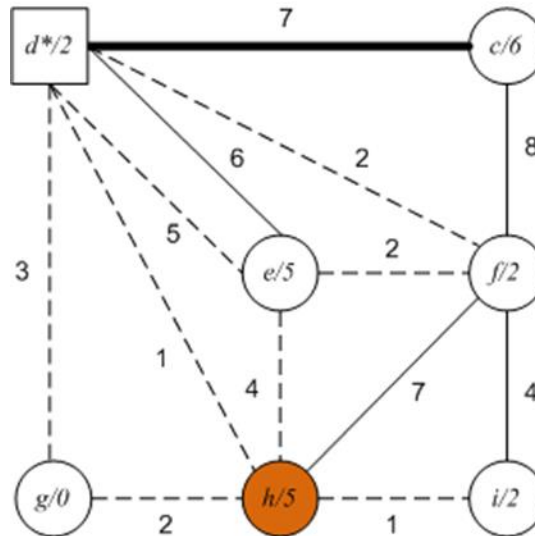
d^* - root

d^*c - augmenting path

Maximum-weight matching

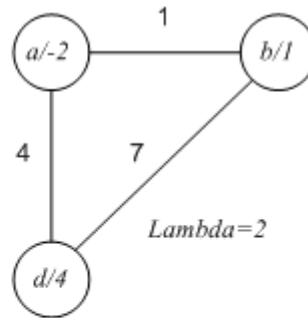
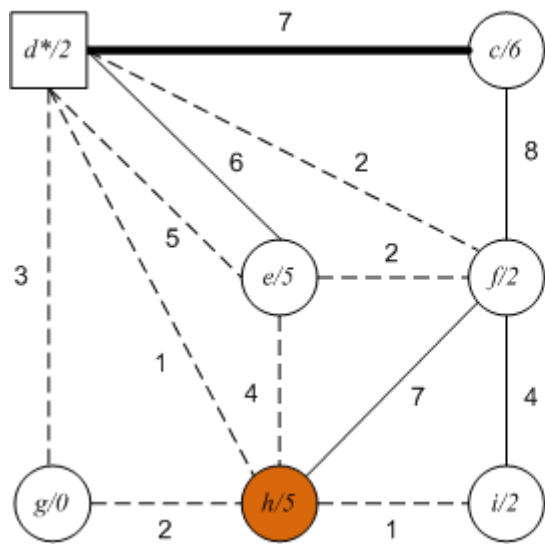


d* - root
d*c - augmenting path

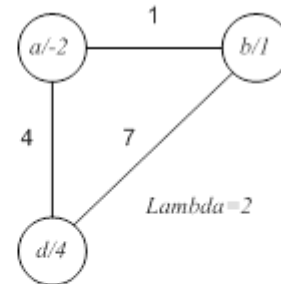
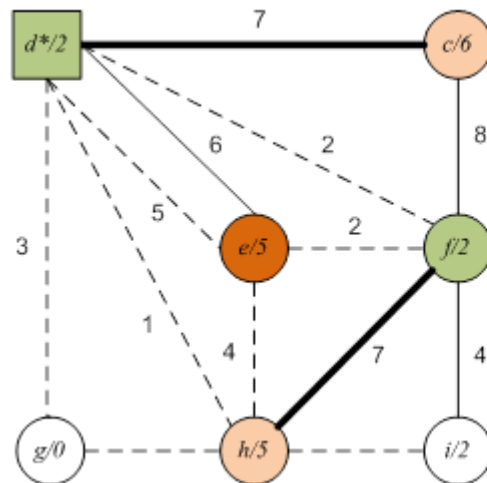


h - root
hf - augmenting path

Maximum-weight matching

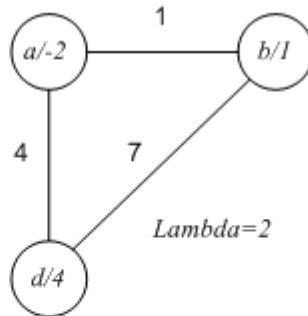
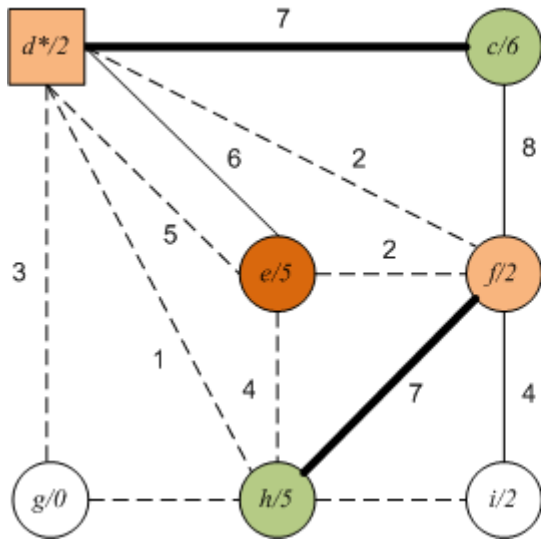


h - root
 hf - augmenting path

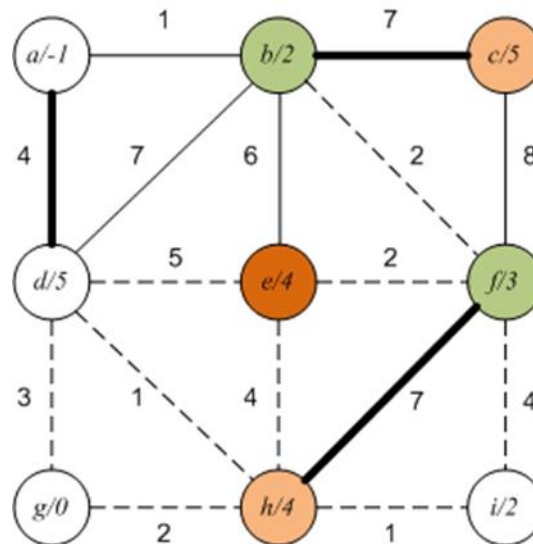


e - root
 Hungarian tree
 D1:
 (h,i): $5+2-1=6$
 (h,g): $5+0-2=3$ min
 D2:
 (h,c): $(5+5-4)/2=3$ min
 D3:
 d^* : $2/2=1$ min
 $D=1=D3$

Maximum-weight matching

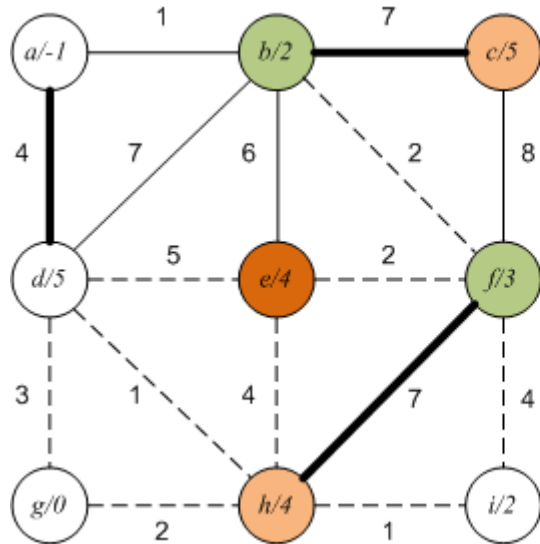


e - root
 Hungarian tree
 D1:
 (h,i): $5+2-1=6$
 (h,g): $5+0-2=3$ min
 D2:
 (h,e): $(5+5-4)/2=3$ min
 D3:
 d*: $2/2=1$ min
 D=1=D3

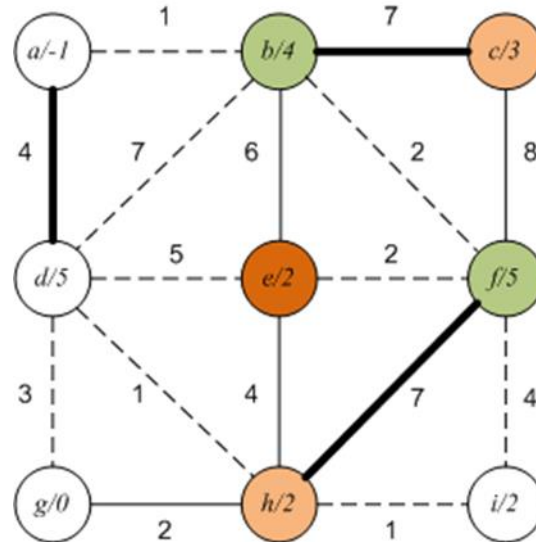


e - root
 Hungarian tree
 D1:
 (h,i): $4+2-1=5$
 (h,g): $4+0-2=2$ min
 (e,d): $4+5-5=4$
 D2:
 (h,e): $(4+4-4)/2=2$ min
 D3: no
 D=1=D1=D2

Maximum-weight matching

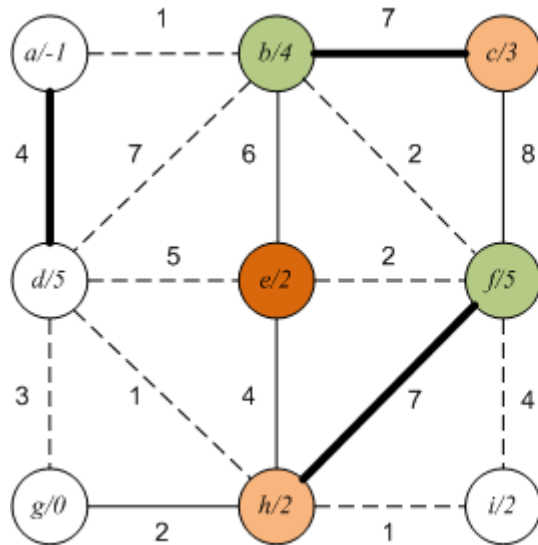


e - root
 Hungarian tree
 D1:
 (h,i): $4+2-1=5$
 (h,g): $4+0-2=2$ min
 (e,d): $4+5-5=4$
 D2:
 (h,e): $(4+4-4)/2=2$ min
 D3: no
 D=1=D1=D2

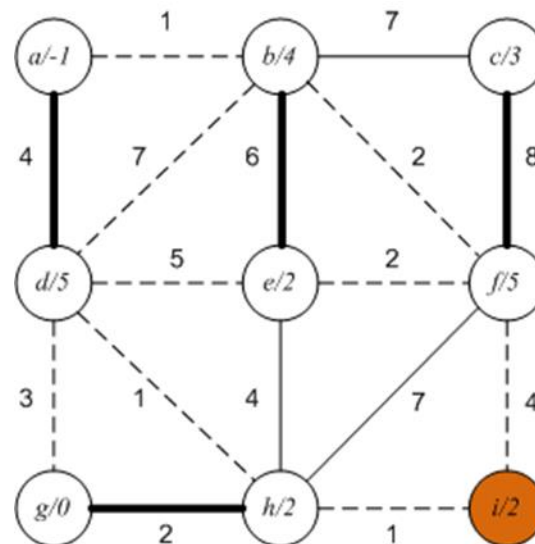


e - root
 ebcfhg – augmenting path

Maximum-weight matching

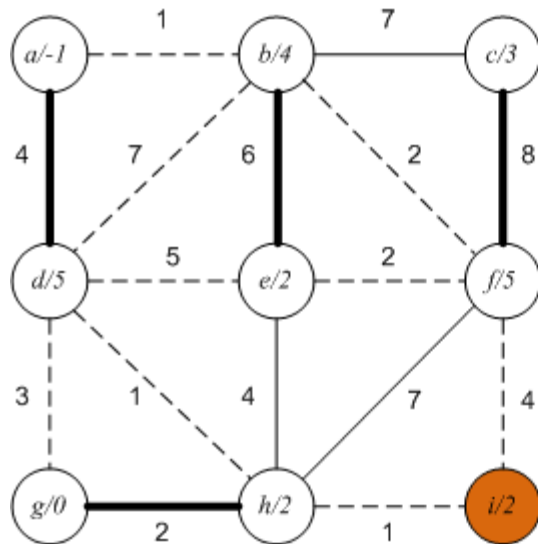


e - root
 ebcfhg – augmenting path

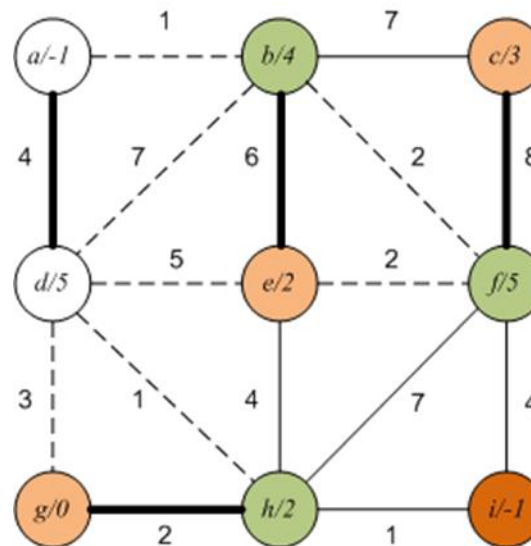


i - root
 Hungarian tree
 D1:
 (i,f): $2+5-4=3$ min
 (i,h): $2+2-1=3$ min
 D2: no
 D3: no
 D=3=D1

Maximum-weight matching

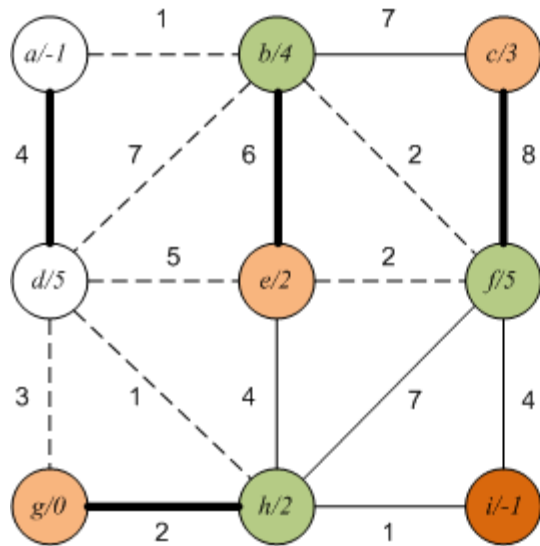


i - root
 Hungarian tree
 D1:
 (i,f): $2+5-4=3$ min
 (i,h): $2+2-1=3$ min
 D2: no
 D3: no
 D=3=D1

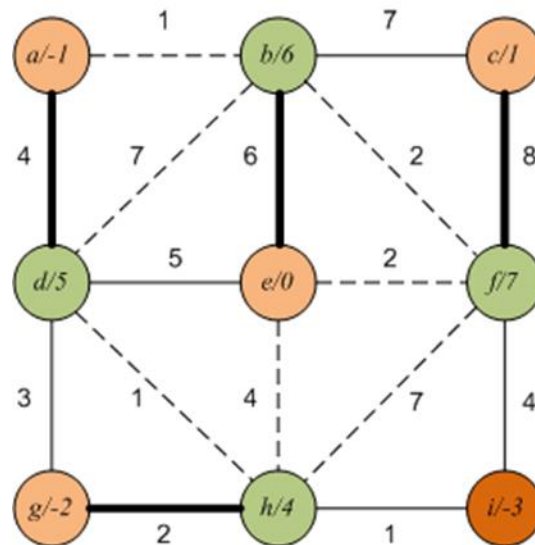


i - root
 Hungarian tree
 D1:
 (e,d): $2+5-5=2$ min
 (g,d): $0+5-3=2$ min
 D2: no
 D3: no
 D=1=D1

Maximum-weight matching



i - root
 Hungarian tree
 D1:
 (e,d): $2+5-5=2$ min
 (g,d): $0+5-3=2$ min
 D2: no
 D3: no
 D=1=D1

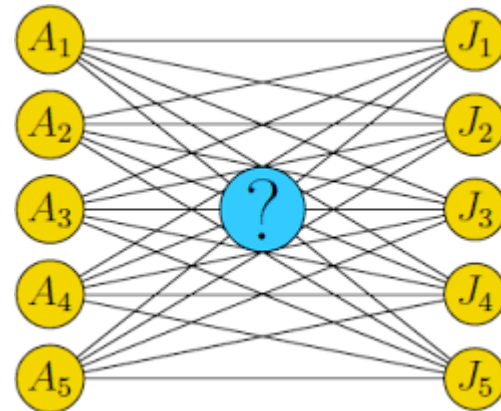


i - root
 Hungarian tree
 D1: no
 D2: no
 D3: no

Assignment problem

- To find, in a **weighted bipartite graph**, a matching in which the sum of weights of the edges is as small (large) as possible.

		Machines				
		A	B	C	D	E
Job	1	5	7	11	6	7
	2	8	5	5	6	5
	3	6	7	10	7	3
	4	10	4	8	2	4



Assignment problem

Bipartite graph:

- No odd-length cycles;
- **No blossoms!**
- Hungarian algorithm.

Assignment problem

Initial problem

Objective function:

$$\sum_{j=1}^q c_j \xi_j \rightarrow \min;$$

Subject to:

$$\sum_{j \in T_k} \xi_j = 1,$$

$$k = \overline{1, p}, T_k = \{j = (i, k)\};$$

$$\xi_j \geq 0, j = \overline{1, q}.$$

Dual problem

Objective function:

$$\sum_{i=1}^p \pi_i \rightarrow \max;$$

Subject to:

$$\pi_i + \pi_k \leq c_j,$$
$$\forall j = (i, k), j = \overline{1, q}.$$

Minimum-cost flow

- Residual network

