

Lecture 5: Matchings

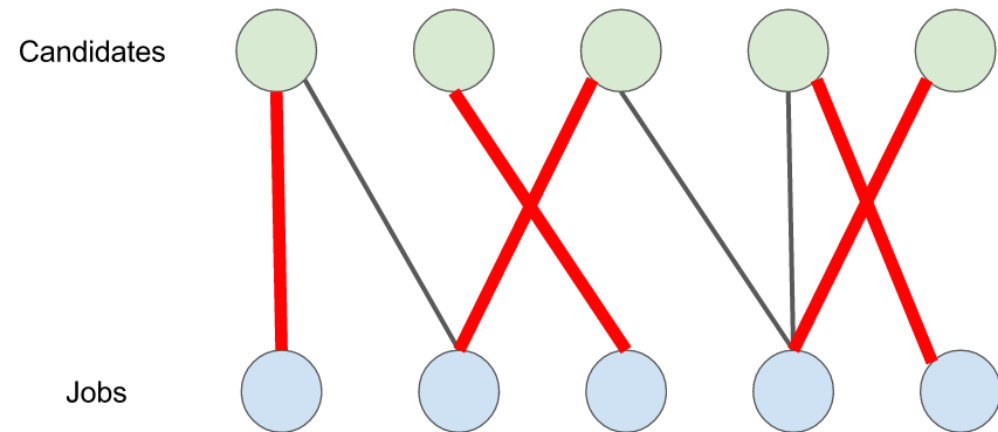
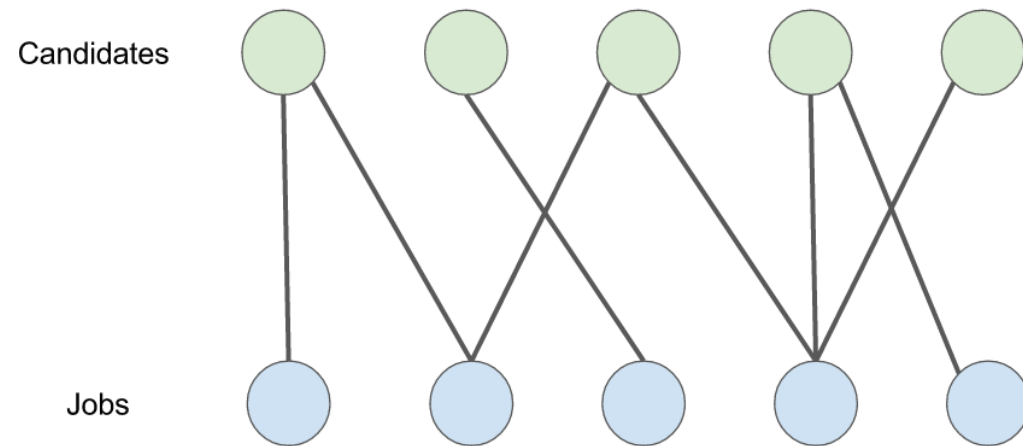
Shuai Li

John Hopcroft Center, Shanghai Jiao Tong University

<https://shuaili8.github.io>

<https://shuaili8.github.io/Teaching/CS3330/index.html>

Motivating example

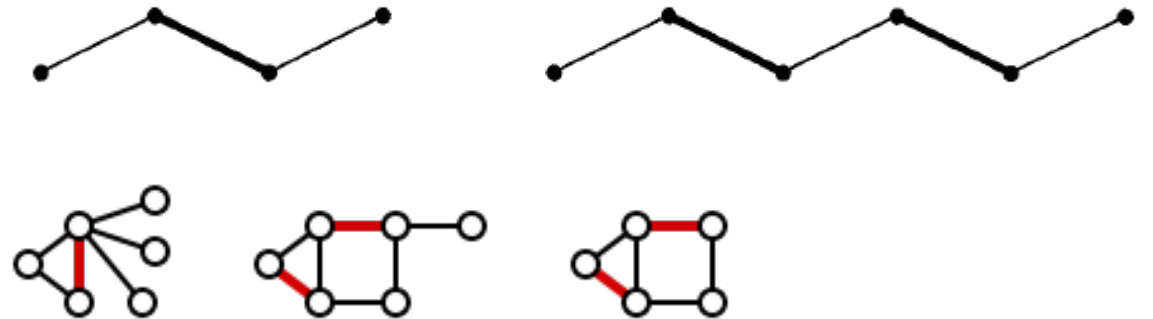


Definitions

- A **matching** is a set of independent edges, in which no pair of edges shares a vertex
- The vertices incident to the edges of a matching M are **M -saturated** (饱和的); the others are **M -unsaturated**
- A **perfect matching** in a graph is a matching that saturates every vertex
- **Example** (3.1.2, W) The number of perfect matchings in $K_{n,n}$ is $n!$
- **Example** (3.1.3, W) The number of perfect matchings in K_{2n} is
$$f_n = (2n - 1)(2n - 3) \cdots 1 = (2n - 1)!!$$

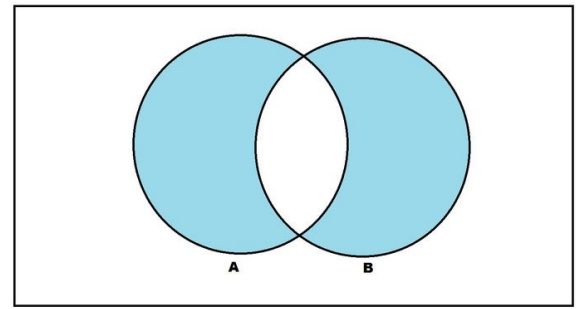
Maximal/maximum matchings 极大/最大

- A **maximal matching** in a graph is a matching that cannot be enlarged by adding an edge
- A **maximum matching** is a matching of maximum size among all matchings in the graph
- Example: P_3, P_5

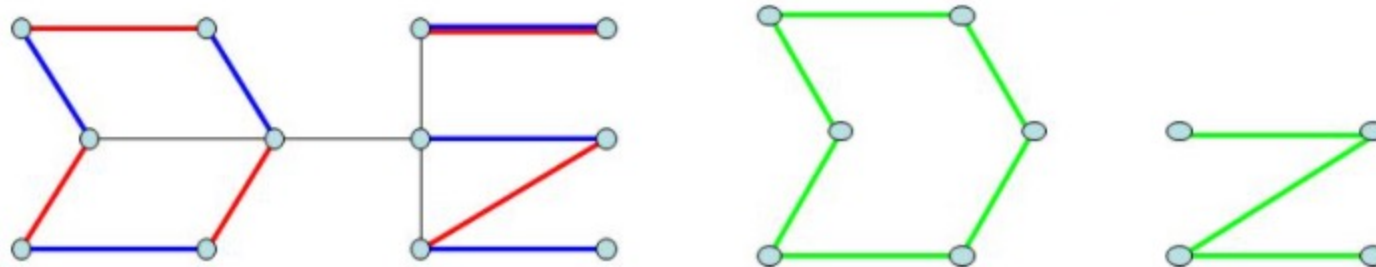


- Every maximum matching is maximal, but not every maximal matching is a maximum matching

Symmetric difference of matchings



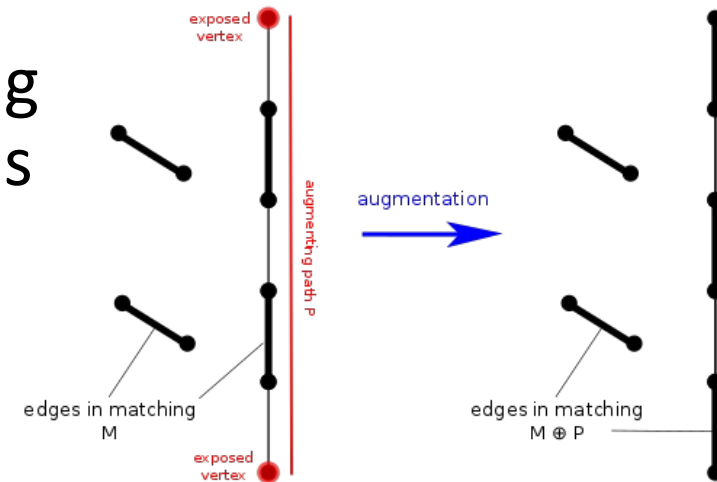
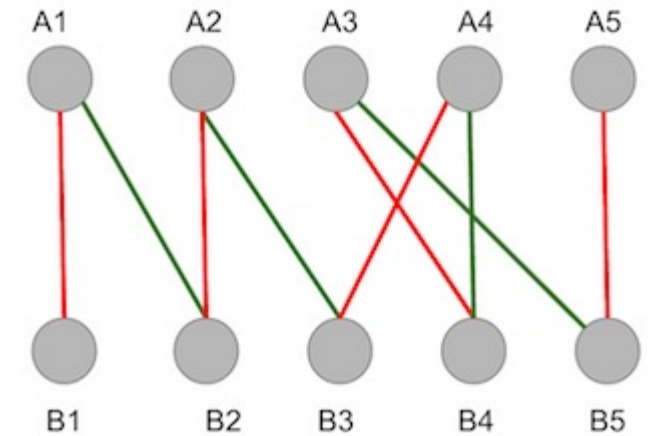
- The **symmetric difference** of M, M' is $M \Delta M' = (M - M') \cup (M' - M)$
- **Lemma** (3.1.9, W) Every component of the symmetric difference of two matchings is a path or an even cycle



Maximum matching and augmenting path

- Given a matching M , an M -alternating path is a path that alternates between edges in M and edges not in M
- An M -alternating path whose endpoints are M -unsaturated is an M -augmenting path
- Theorem** (3.1.10, W; 1.50, H; Berge 1957) A matching M in a graph G is a **maximum** matching in $G \iff G$ has no M -augmenting path

Lemma (3.1.9, W) Every component of the symmetric difference of two matchings is a path or an even cycle



Hall's theorem (TONCAS)

- **Theorem** (3.1.11, W; 1.51, H; 2.1.2, D; Hall 1935) Let G be a bipartite graph with partition X, Y .

G contains a matching of $X \Leftrightarrow |N(S)| \geq |S|$ for all $S \subseteq X$

Theorem (3.1.10, W; 1.50, H; Berge 1957) A matching M in a graph G is a **maximum** matching in $G \Leftrightarrow G$ has no M -augmenting path

- **Exercise**. Read the other two proofs in Diestel.
- **Corollary** (3.1.13, W; 2.1.3, D) Every k -regular ($k > 0$) bipartite graph has a perfect matching

General regular graph

- **Corollary** (2.1.5, D) Every regular graph of positive even degree has a 2-factor
 - A k -regular spanning subgraph is called a **k -factor**
 - A perfect matching is a 1-factor

Theorem (1.2.26, W) A graph G is Eulerian \iff it has at most one nontrivial component and its vertices all have even degree

Corollary (3.1.13, W; 2.1.3, D) Every k -regular ($k > 0$) bipartite graph has a perfect matching

Application to SDR

- Given some family of sets X , a **system of distinct representatives** for the sets in X is a 'representative' collection of distinct elements from the sets of X

$$S_1 = \{2, 8\},$$

$$S_2 = \{8\},$$

$$S_3 = \{5, 7\},$$

$$S_4 = \{2, 4, 8\},$$

$$S_5 = \{2, 4\}.$$

The family $X_1 = \{S_1, S_2, S_3, S_4\}$ does have an SDR, namely $\{2, 8, 7, 4\}$. The family $X_2 = \{S_1, S_2, S_4, S_5\}$ does not have an SDR.

- Theorem**(1.52, H) Let S_1, S_2, \dots, S_k be a collection of finite, nonempty sets. This collection has SDR \Leftrightarrow for every $t \in [k]$, the union of any t of these sets contains at least t elements

Theorem (3.1.11, W; 1.51, H; 2.1.2, D; Hall 1935) Let G be a bipartite graph with partition X, Y .

G contains a matching of $X \Leftrightarrow |N(S)| \geq |S|$ for all $S \subseteq X$

König Theorem

Augmenting Path Algorithm

Vertex cover

- A set $U \subseteq V$ is a **(vertex) cover** of E if every edge in G is incident with a vertex in U
- Example:
 - Art museum is a graph with hallways are edges and corners are nodes
 - A security camera at the corner will guard the paintings on the hallways
 - The minimum set to place the cameras?

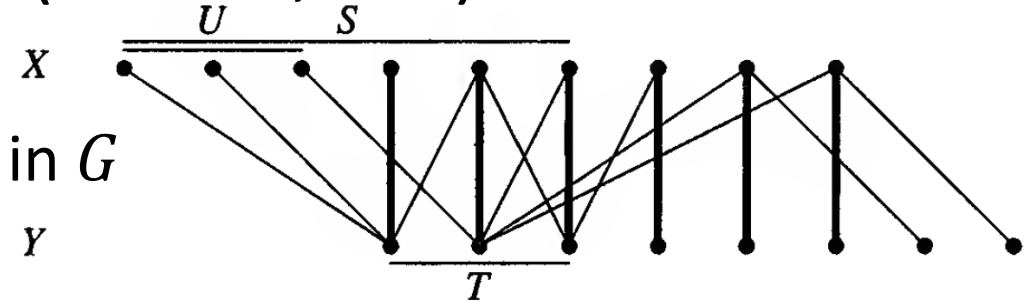
König-Egeváry Theorem (Min-max theorem)

- **Theorem** (3.1.16, W; 1.53, H; 2.1.1, D; König 1931; Egeváry 1931)
Let G be a bipartite graph. The **maximum** size of a matching in G is equal to the **minimum** size of a vertex cover of its edges

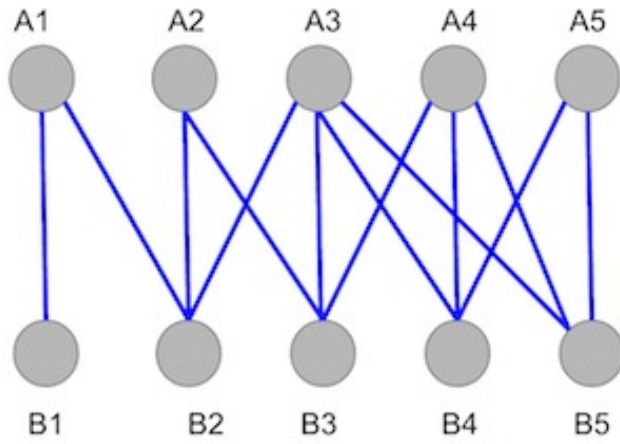
Theorem (3.1.10, W; 1.50, H; Berge 1957) A matching M in a graph G is a **maximum** matching in $G \Leftrightarrow G$ has no M -augmenting path

Augmenting path algorithm (3.2.1, W)

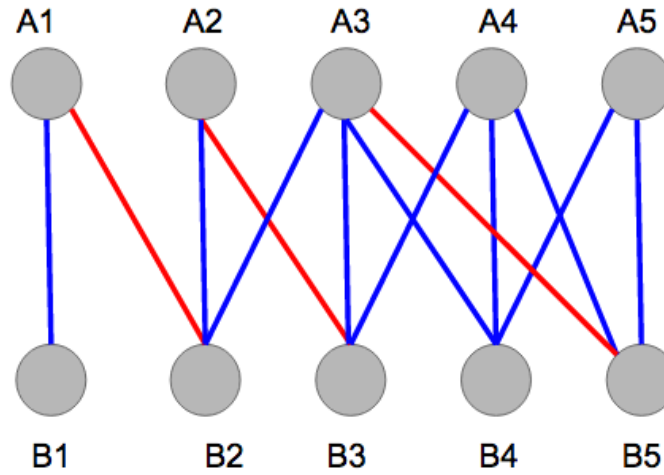
- **Input:** G is Bipartite with X, Y , a matching M in G
 $U = \{M\text{-unsaturated vertices in } X\}$
- **Idea:** Explore M -alternating paths from U
 letting $S \subseteq X$ and $T \subseteq Y$ be the sets of vertices reached
- **Initialization:** $S = U, T = \emptyset$ and all vertices in S are unmarked
- **Iteration:**
 - If S has no unmarked vertex, stop and report $T \cup (X - S)$ as a minimum cover and M as a maximum matching
 - Otherwise, select an unmarked $x \in S$ to explore
 - Consider each $y \in N(x)$ such that $xy \notin M$
 - If y is unsaturated, terminate and report an M -augmenting path from U to y
 - Otherwise, $yw \in M$ for some w
 - include y in T (reached from x) and include w in S (reached from y)
 - After exploring all such edges incident to x , mark x and iterate.



Example



Red: A random matching



Theoretical guarantee for Augmenting path algorithm

- **Theorem** (3.2.2, W) Repeatedly applying the Augmenting Path Algorithm to a bipartite graph produces a matching and a vertex cover of equal size

Weighted Bipartite Matching

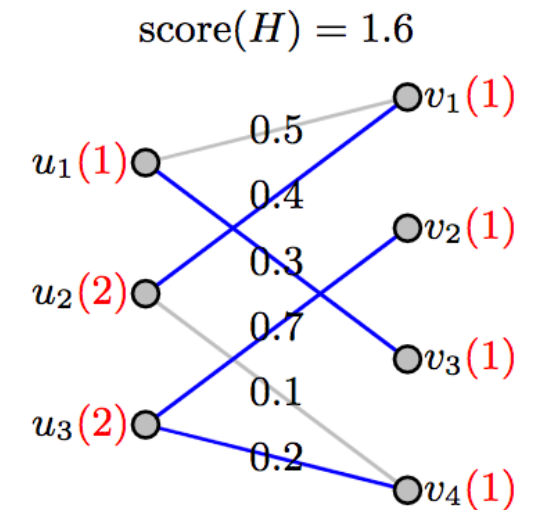
Hungarian Algorithm

Weighted bipartite matching

- The **maximum weighted matching problem** is to seek a perfect matching M to maximize the total weight $w(M)$
- Bipartite graph
 - W.l.o.g. Assume the graph is $K_{n,n}$ with $w_{i,j} \geq 0$ for all $i, j \in [n]$
 - Optimization:

$$\begin{aligned} \max \quad & w(M_a) = \sum_{i,j} a_{i,j} w_{i,j} \\ \text{s.t.} \quad & a_{i,1} + \dots + a_{i,n} = 1 \text{ for any } i \\ & a_{1,j} + \dots + a_{n,j} = 1 \text{ for any } j \\ & a_{i,j} \in \{0,1\} \end{aligned}$$

- Integer programming
- General IP problems are NP-Complete



(Weighted) cover

- A (weighted) **cover** is a choice of labels u_1, \dots, u_n and v_1, \dots, v_n such that $u_i + v_j \geq w_{i,j}$ for all i, j
 - The **cost** $c(u, v)$ of a cover (u, v) is $\sum_i u_i + \sum_j v_j$
 - The **minimum weighted cover problem** is that of finding a cover of minimum cost
- Optimization problem

$$\begin{aligned} \min \quad & c(u, v) = \sum_i u_i + \sum_j v_j \\ \text{s. t.} \quad & u_i + v_j \geq w_{i,j} \text{ for any } i, j \end{aligned}$$

Duality

(IP)

$$\begin{aligned} \max \quad & \sum_{i,j} a_{i,j} w_{i,j} \\ \text{s.t.} \quad & a_{i,1} + \cdots + a_{i,n} = 1 \text{ for any } i \\ & a_{1,j} + \cdots + a_{n,j} = 1 \text{ for any } j \\ & a_{i,j} \in \{0,1\} \end{aligned}$$

(Linear programming)

$$\begin{aligned} \max \quad & \sum_{i,j} a_{i,j} w_{i,j} \\ \text{s.t.} \quad & a_{i,1} + \cdots + a_{i,n} = 1 \text{ for any } i \\ & a_{1,j} + \cdots + a_{n,j} = 1 \text{ for any } j \\ & a_{i,j} \geq 0 \end{aligned}$$

(Dual)

$$\begin{aligned} \min \quad & \sum_i u_i + \sum_j v_j \\ \text{s.t.} \quad & u_i + v_j \geq w_{i,j} \text{ for any } i, j \end{aligned}$$

- Weak duality theorem

- For each feasible solution a and (u, v)

$$\sum_{i,j} a_{i,j} w_{i,j} \leq \sum_i u_i + \sum_j v_j$$

$$\text{thus } \max \sum_{i,j} a_{i,j} w_{i,j} \leq \min \sum_i u_i + \sum_j v_j$$

Duality (cont.)

- Strong duality theorem

- If one of the two problems has an optimal solution, so does the other one and that the bounds given by the weak duality theorem are tight

$$\max \sum_{i,j} a_{i,j} w_{i,j} = \min \sum_i u_i + \sum_j v_j$$

- **Lemma** (3.2.7, W) For a perfect matching M and cover (u, v) in a weighted bipartite graph G , $c(u, v) \geq w(M)$.
 $c(u, v) = w(M) \Leftrightarrow M$ consists of edges $x_i y_j$ such that $u_i + v_j = w_{i,j}$
In this case, M and (u, v) are optimal.

Equality subgraph

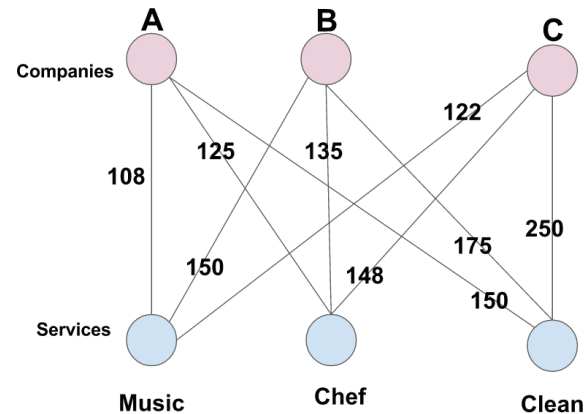
- The **equality subgraph** $G_{u,v}$ for a cover (u, v) is the **spanning** subgraph of $K_{n,n}$ having the edges $x_i y_j$ such that $u_i + v_j = w_{i,j}$
 - So if $c(u, v) = w(M)$ for some perfect matching M , then M is composed of edges in $G_{u,v}$
 - And if $G_{u,v}$ contains a perfect matching M , then (u, v) and M (whose weights are $u_i + v_j$) are both optimal

Hungarian algorithm

- **Input:** Weighted $K_{n,n} = B(X, Y)$
- **Idea:** Iteratively adjusting the cover (u, v) until the equality subgraph $G_{u,v}$ has a perfect matching
- **Initialization:** Let (u, v) be a cover, such as $u_i = \max_j w_{i,j}$, $v_j = 0$

(Dual)

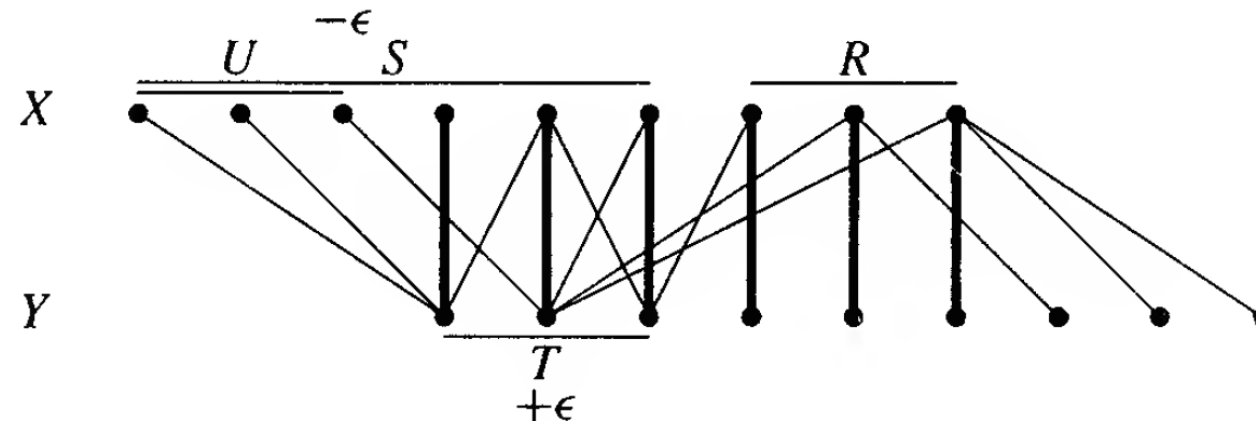
$$\begin{aligned} \min \quad & \sum_i u_i + \sum_j v_j \\ \text{s.t.} \quad & u_i + v_j \geq w_{i,j} \text{ for any } i, j \end{aligned}$$



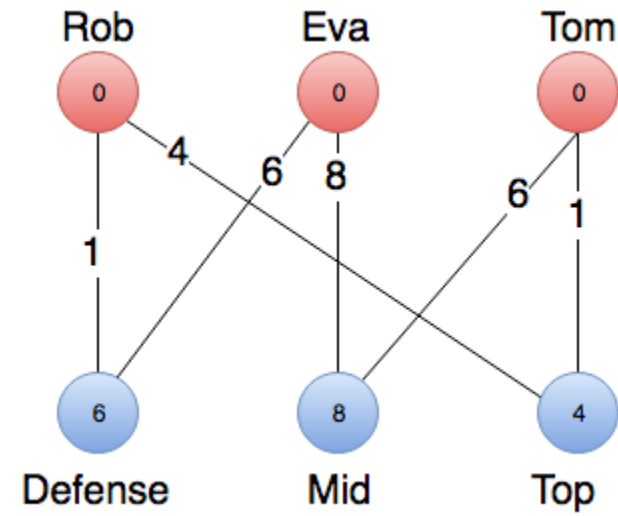
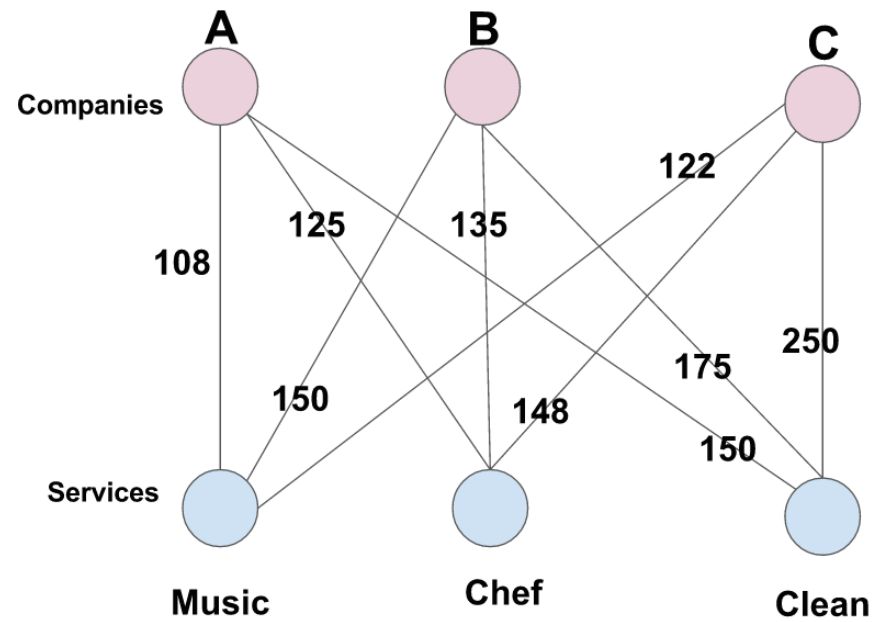
Hungarian algorithm (cont.)

- **Iteration:** Find a maximum matching M in $G_{u,v}$
 - If M is a perfect matching, stop and report M as a maximum weight matching
 - Otherwise, let Q be a vertex cover of size $|M|$ in $G_{u,v}$
 - Let $R = X \cap Q, T = Y \cap Q$

$$\epsilon = \min\{u_i + v_j - w_{i,j} : x_i \in X - R, y_j \in Y - T\}$$
 - Decrease u_i by ϵ for $x_i \in X - R$ and increase v_j by ϵ for $y_j \in T$
 - Form the new equality subgraph and repeat



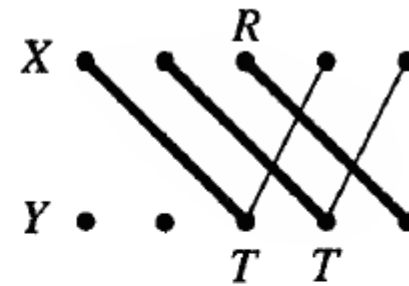
Example



Example 2: Excess matrix

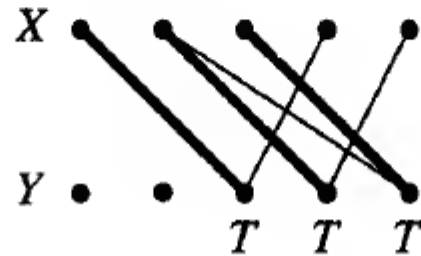
$$\begin{pmatrix} 4 & 1 & 6 & 2 & 3 \\ 5 & 0 & 3 & 7 & 6 \\ 2 & 3 & 4 & 5 & 8 \\ 3 & 4 & 6 & 3 & 4 \\ 4 & 6 & 5 & 8 & 6 \end{pmatrix} \rightarrow \begin{matrix} & 0 & 0 & 0 & 0 & 0 \\ 6 & 2 & 5 & \underline{0} & 4 & 3 \\ 7 & 2 & 7 & 4 & \underline{0} & 1 \\ 8 & 6 & 5 & 4 & 3 & \underline{0} \\ 6 & 3 & 2 & 0 & 3 & 2 \\ 8 & 4 & 2 & 3 & 0 & 2 \end{matrix} \begin{matrix} \\ \\ \\ \\ R \\ \end{matrix}$$

$T \quad T$



$$\begin{matrix} & 0 & 0 & 1 & 1 & 0 \\ 5 & 1 & 4 & \underline{0} & 4 & 2 \\ 6 & 1 & 6 & 4 & \underline{0} & 0 \\ 8 & 6 & 5 & 5 & 4 & \underline{0} \\ 5 & 2 & 1 & 0 & 3 & 1 \\ 7 & 3 & 1 & 3 & 0 & 1 \end{matrix}$$

$T \quad T \quad T$



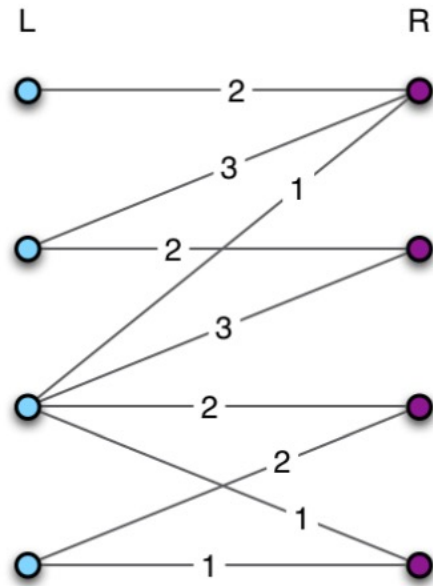
$$\rightarrow \begin{matrix} & 0 & 0 & 2 & 2 & 1 \\ 4 & 0 & 3 & \underline{0} & 4 & 2 \\ 5 & \underline{0} & 5 & 4 & 0 & 0 \\ 7 & 5 & 4 & 5 & 4 & \underline{0} \\ 4 & 1 & \underline{0} & 0 & 3 & 1 \\ 6 & 2 & 0 & 3 & \underline{0} & 1 \end{matrix}$$

Optimal value is the same
But the solution is not unique

Theoretical guarantee for Hungarian algorithm

- **Theorem** (3.2.11, W) The Hungarian Algorithm finds a maximum weight matching and a minimum cost cover

Example 3



Back to (unweighted) bipartite graph

- The weights are binary 0,1
- Hungarian algorithm always maintain integer labels in the weighted cover, thus the solution will always be 0,1
- The vertices receiving label 1 must cover the weight on the edges, thus cover all edges
- So the solution is a minimum vertex cover

Summary

- Matching in bipartite graphs
 - Hall's Theorem (TONCAS)
 - König Theorem: For bipartite graph, the maximum size of a matching is equal to the minimum size of a vertex cover of its edges
 - Augmenting Path Algorithm
- Matchings in weighted bipartite graphs
 - Weighted cover, Hungarian algorithm, equality subgraph, excess matrix

Shuai Li

<https://shuaili8.github.io>

Questions?