



## Chapter 6 Graph Algorithms

## Algorithm Theory WS 2016/17

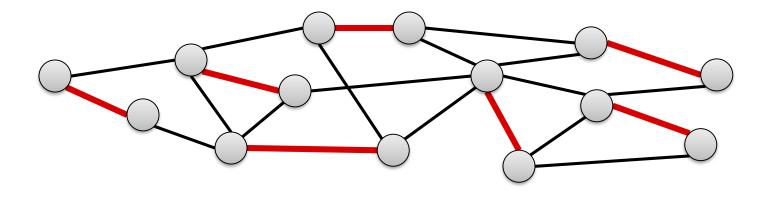
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## Vertex Cover vs Matching



Consider a matching M and a vertex cover S **Claim:**  $|M| \le |S|$  for bipartile scapes :  $|M^*| = |S^*|$ **Proof:** 

- At least one node of every edge  $\{u, v\} \in M$  is in S
- Needs to be a different node for different edges from *M*



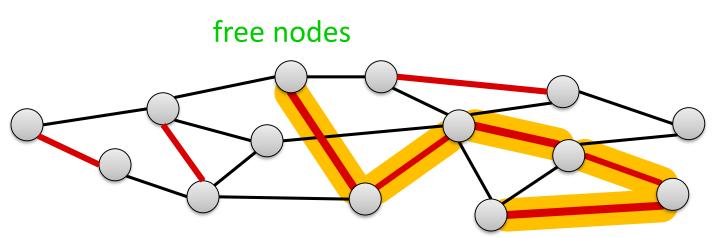
## Augmenting Paths



Consider a matching M of a graph G = (V, E):

• A node  $v \in V$  is called **free** iff it is not matched

**Augmenting Path:** A (odd-length) path that starts and ends at a free node and visits edges in  $E \setminus M$  and edges in M alternatingly.



alternating path

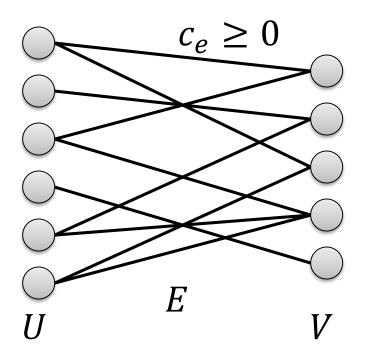
• Matching *M* can be improved using an augmenting path by switching the role of each edge along the path

## Maximum Weight Bipartite Matching

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• Let's again go back to bipartite graphs...

**Given:** Bipartite graph  $G = (U \cup V, E)$  with edge weights  $c_e \ge 0$ **Goal:** Find a matching *M* of maximum total weight



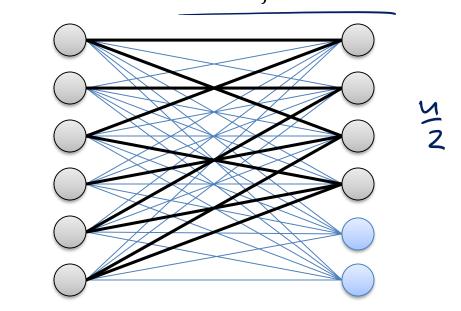
## Minimum Weight Perfect Matching



**Claim:** Max weight bipartite matching is **equivalent** to finding a **minimum weight perfect matching** in a complete bipartite graph.

- 1. Turn into maximum weight perfect matching
  - add dummy nodes to get two equal-sized sides
  - add edges of weight 0 to make graph complete bipartite

2. Replace weights: 
$$c'_e \coloneqq \max_f \{c_f\} - c_e$$



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## As an Integer Linear Program

We can formulate the problem as an integer linear program

Var.  $x_{uv}$  for every edge  $(u, v) \in U \times V$  to encode matching M:

$$x_{uv} = \begin{cases} \underline{1}, & \text{if } \{\underline{u}, v\} \in M \\ \underline{0}, & \text{if } \{u, v\} \notin M \end{cases}$$

# Minimum Weight Perfect Matching $\min_{(u,v)\in U\times V} c_{uv} \cdot x_{uv}$ s.t. $\forall u \in U: \sum_{\substack{v \in V \\ i \in V}} x_{uv} = 1, \quad \forall v \in V: \sum_{u \in U} x_{uv} = 1$ $\forall u \in U, \forall v \in V: x_{uv} \in \{0,1\}$

## **Dual Problem**



- Every linear program has a dual linear program
  - The dual of a minimization problem is a maximization problem
  - Strong duality: primal LP and dual LP have the same objective value

In the case of the minimum weight perfect matching problem

- Assign a variable  $a_u \ge 0$  to each node  $u \in U$ and a variable  $b_v \ge 0$  to each node  $v \in V$
- Condition: for every edge  $(u, v) \in U \times V$ :  $(a_u + b_v \leq c_{uv})$
- Given perfect matching <u>M</u>:

$$\sum_{(u,v)\in M} c_{uv} \ge \sum_{u\in U} a_u + \sum_{v\in V} b_v$$

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 $a_{uv} \ge a_{u} + b_{v}$ 

## **Dual Linear Program**



• Variables  $a_u \ge 0$  for  $u \in U$  and  $b_v \ge 0$  for  $v \in V$ 

$$\max \sum_{u \in U} a_u + \sum_{v \in V} b_v$$
s. t.  

$$\forall u \in U, \forall v \in V: \quad a_u + b_v \leq c_{uv} \quad \Rightarrow$$
For every perfect matching M: (for every feasible dual sol.)  

$$\sum_{\substack{(u,v) \in M}} c_{uv} \geq \sum_{u \in U} a_u + \sum_{v \in V} b_v$$

$$a_i \quad c_i \quad b_i \quad c_{uv} = a_u + b_v \quad for all (u,v) \in M:$$

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## **Complementary Slackness**

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• A perfect matching *M* is optimal if

$$\sum_{(u,v)\in M} c_{uv} = \sum_{u\in U} a_u + \sum_{v\in V} b_v$$

$$C_{uv} \ge a_u + b_v$$

• In that case, for every  $(u, v) \in M$ 

$$\mathbf{w}_{uv} \coloneqq c_{uv} - a_u - b_v = 0$$

- In this case, M is also an optimal solution to the LP relaxation of the problem
- Every optimal LP solution can be characterized by such a property, which is then generally referred to as complementary slackness
- **Goal:** Find a dual solution  $a_u, b_v$  and a perfect matching such that the complementary slackness condition is satisfied!
  - i.e., for every matching edge (u, v), we want  $w_{uv} = 0$
  - We then know that the matching is optimal!

## Algorithm Overview



- Start with any feasible dual solution a<sub>u</sub>, b<sub>v</sub>

  i.e., solution satisfies that for all (u, v): c<sub>uv</sub> ≥ a<sub>u</sub> + b<sub>v</sub>
  for example: a<sub>u</sub> = 0, b<sub>v</sub> = 0 ∀h<sub>v</sub>v

  Let E<sub>0</sub> be the edges for which w<sub>uv</sub> = 0

  Recall that w<sub>uv</sub> = c<sub>uv</sub> a<sub>u</sub> b<sub>v</sub>
- Compute maximum cardinality matching *M* of *E*<sub>0</sub>
- All edges (u, v) of M satisfy  $w_{uv} = 0$ 
  - Complementary slackness if satisfied
  - If M is a perfect matching, we are done  $\sim$

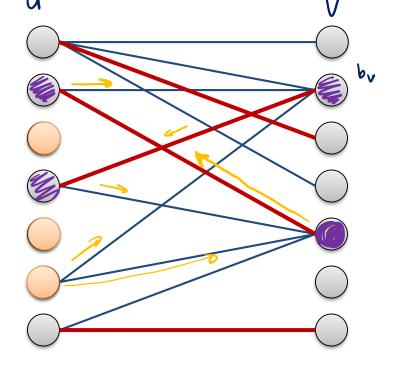
• If *M* is not a perfect matching, dual solution can be improved

## Marked Nodes



#### Define set of marked nodes L:

• Set of nodes which can be reached on alternating paths on edges in  $\underline{E_0}$  starting from unmatched nodes in U



edges  $E_0$  with  $w_{uv} = 0$ 

optimal matching M

L<sub>0</sub>: unmatched nodes in U

L: all nodes that can be reached on alternating paths starting from L<sub>0</sub>

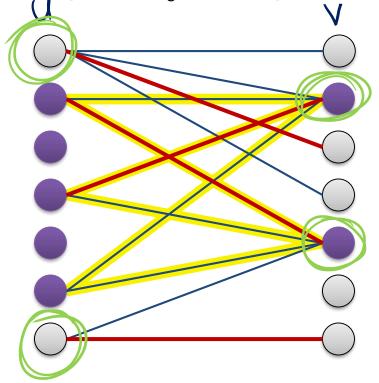
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Servadion: all marked nodes in V are malched

-> otherwise: augun path

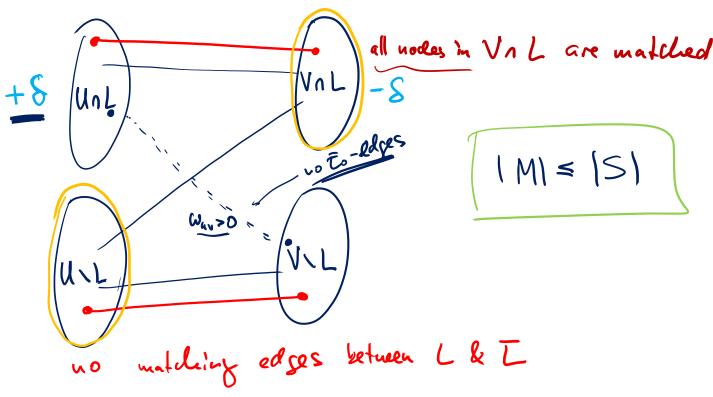
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## Marked Nodes – Vertex Cover



#### Lemma:

- a) There are no  $E_0$ -edges between  $U \cap L$  and  $V \setminus L$
- b) The set  $(U \setminus L) \cup (V \cap L)$  is a vertex cover of size |M| of the graph induced by  $E_0$

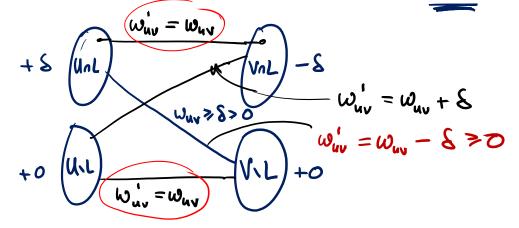


## Improved Dual Solution



**Recall:** all edges (u, v) between  $U \cap L$  and  $V \setminus L$  have  $w_{uv} > 0$ 

**Claim:** New dual solution is feasible (all  $w_{uv}$  remain  $\geq 0$ )

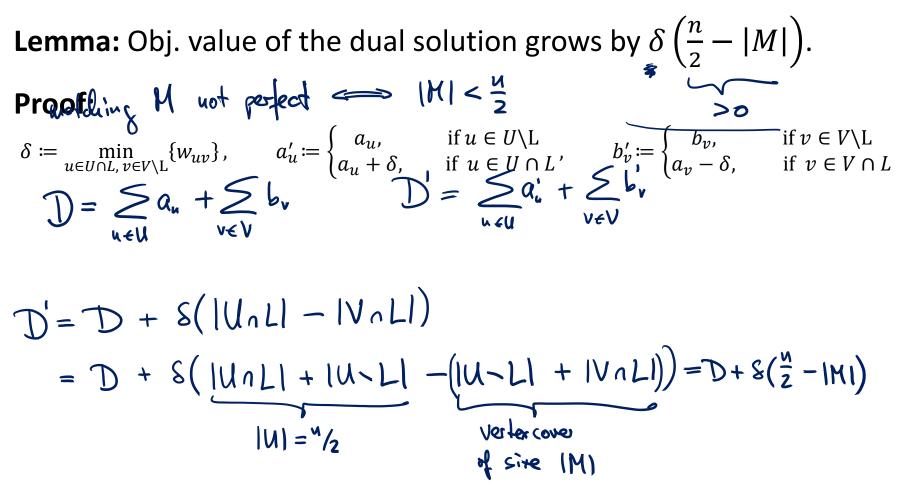


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## Improved Dual Solution





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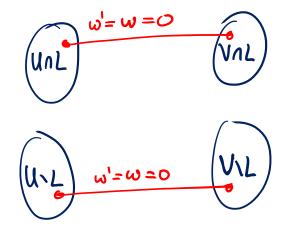
## Termination

#### Some terminology

- Old dual solution:  $a_u$ ,  $b_v$ ,  $w_{uv} \coloneqq c_{uv} a_u b_v$
- New dual solution:  $a'_u$ ,  $b'_v$ ,  $w'_{uv} \coloneqq c_{uv} a'_u b'_v$
- $E_0 \coloneqq \{(u, v) : w_{uv} = 0\}, \quad E'_0 \coloneqq \{(u, v) : w'_{uv} = 0\}$
- $\underline{M}, \underline{M}'$ : max. cardinality matchings of graphs ind. By  $E_0, E'_0$

## **Claim:** $|M'| \ge |M|$ and if |M'| = |M|, we can assume that M = M'.

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## Termination



**Lemma:** The algorithm terminates in at most  $O(n^2)$  iterations.

#### **Proof:** M' = M and $|V \cap L'| > |V \cap L|$ Each iteration: M' > Mor "old altourty path" are still alt. paths $\implies L \subseteq L'$ $W_{uv} = W_{uv}$ VNL UNL $\omega_{uv} = \delta$ $\omega_{uv} = \omega_{uv} - \delta = 0$ becomes a marked nocle! VL 12

## Min. Weight Perfect Matching: Summary



**Theorem:** A minimum weight perfect matching can be computed in time  $O(n^4)$ .

- First dual solution: e.g.,  $a_u = 0$ ,  $b_v = \min_{u \in U} c_{uv}$  $a_u = 0$ ,  $b_v = 0$  also works
- Compute set  $E_0: O(n^2)$

 $O(u^2)$  edge >

• Compute max. cardinality matching of graph induced by  $E_0$ 

- First iteration: 
$$Q(n^2) \cdot Q(n) = O(n^3)$$
  
- Other iterations:  $Q(n^2) \cdot O(1 + |M'| - |M|)$   
total cost when improving matching:  $Q(n^3)$   
total cost when  $|M'| = |M| : Q(n^3)$  for each matching size  
marking  $\cdot Q(n^2)$  (given Eo & M)

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#### We have seen:

- O(mn) time alg. to compute a max. matching in *bipartite graphs*
- $O(mn^2)$  time alg. to compute a max. matching in *general graphs*

#### **Better algorithms:**

• Best known running time (bipartite and general gr.):  $O(m\sqrt{n})$ 

#### Weighted matching:

- Edges have weight, find a matching of **maximum total weight**
- *Bipartite graphs*: polynomial-time primal-dual algorithm
- General graphs: can also be solved in polynomial time (Edmond's algorithms is used as blackbox)