

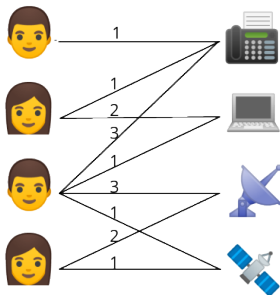
# Maximum Cardinality Matching Problem

Narek Bojikian, Piotr Witkowski, Martin Vogel



June 10, 2019

# Assignment Problem

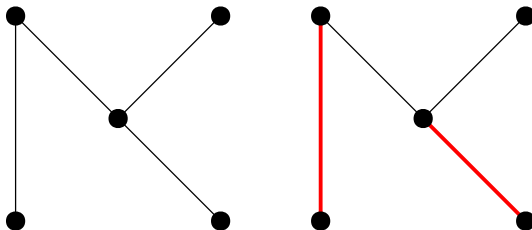


# Definitions

# Matching

## Definition

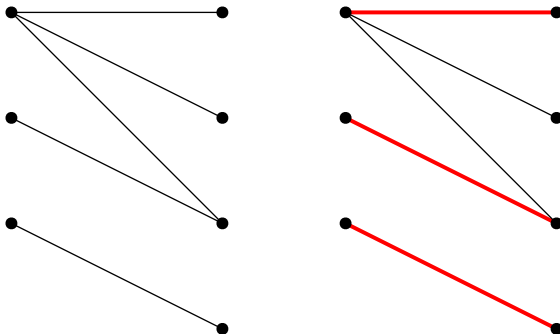
Set of non overlapping edges.



# Bipartite Matching

## Definition

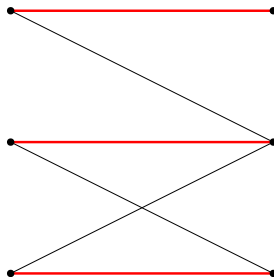
Matching on a bipartite graph.



# Perfect Matching

## Definition

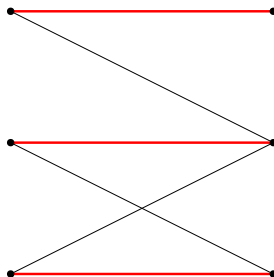
Matching of size  $\frac{|V(G)|}{2}$ .



# Maximum Matching

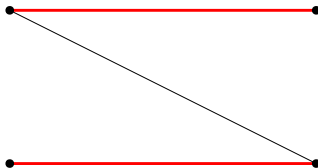
## Definition

Matching of the maximum size.

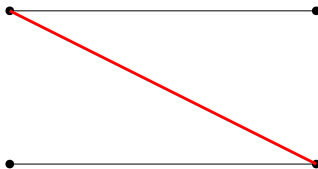


# Maximum Matching vs Maximal Matching

Maximum: matching of the maximum size:



Maximal: no more edges can be added:





# Hall's Theorem

## Definition

A bipartite graph  $G$  consisting of sets  $U$  and  $W$ , has a matching satisfying  $|u|$  if and only if  $|N(x)| \geq |x|$  for every nonempty subset  $X$  of  $U$ .

# Examples - Hall's Theorem

- Examples on the board –

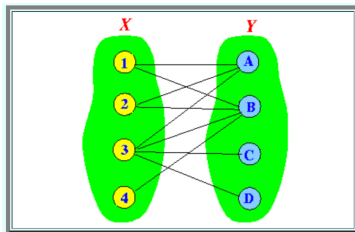
## Definition

The maximum matching for a bipartite graph equals its minimum vertex cover

# Königs Theorem - Example

– Example on the Board –

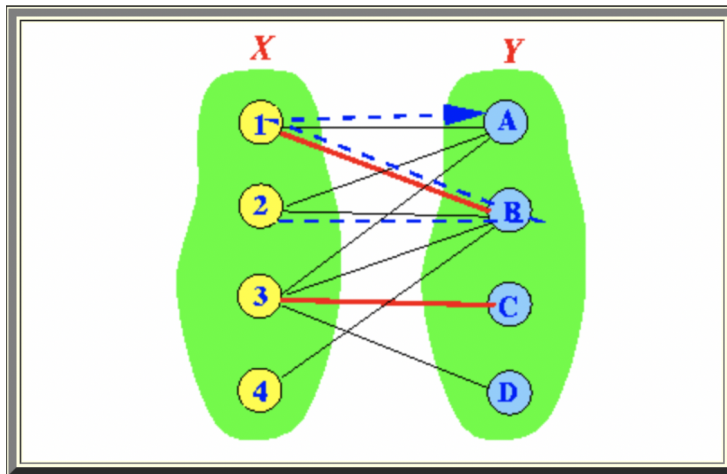
# Alternating Path



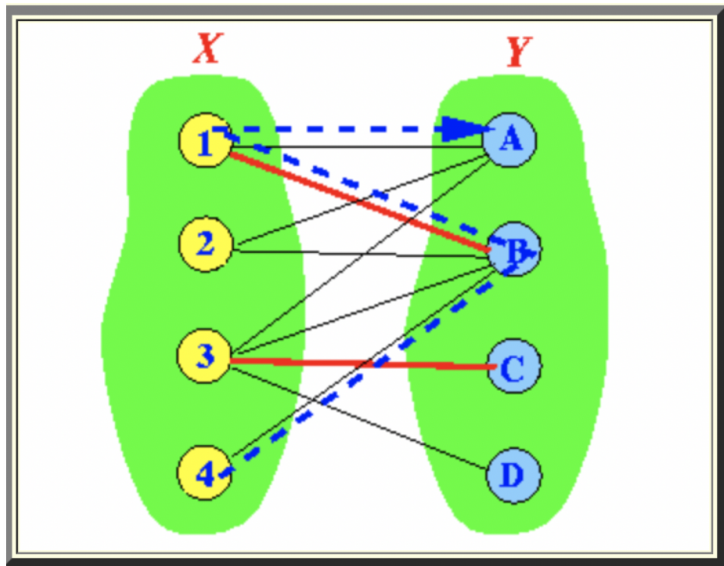
## Definition

Let  $G = (X, Y, E)$  be a bi-partite graph where the vertices are divided into the sets  $X$  and  $Y$  and  $E$  the edges.

# Alternating Path



# Alternating Path



# Alternating Path - Summary

- starts in a vertex element of  $X$  and ends in vertex element of  $Y$
  - must have an odd-number of edges
  - will visit nodes in  $X$  und  $Y$  alternatedly
  - And it starts and ends in free/unmatched vertices
- To go forward, use an edge that is not part of the matching
- To go backward, use an edge that is part of the matching



# Augmenting Path - Definition

## Definition

An augmenting path is an alternating path where the first and last vertex are unmatched.

# Augmenting Path - example

– Example on the Board –

# Breadth First Search - repetition

– Example on the Board –

# Berge Theorem

A matching is a maximum matching if it contains no augmenting path.

# Hungarian Method

Search augmenting paths in the graph until no augmenting path can be found

→ **Time complexity:**  $O(|V|^3)$

Precise information can be found here:

**<https://brilliant.org/wiki/hungarian-matching/>**

**Note:** It is not an algorithm, so it does not specify a implementation

# Maximum Flow Reduction

– Example on the Board –

# Hopcraft-Carp Algorithm

Input: **A bipartite Graph** Initialize Matching

1. Repeat

- Build alternating level graph rooted at unmatched vertices using bfs
- Augment  $M$  via maximal set of vertex disjoint shortest-length paths
- until no augmenting paths exists

2. Return  $M$

**Time complexity:**  $O(|E| * \sqrt{|V|})$

# Hopcraft-Carp Algorithm - bipartite Graph

– Example on the Board –

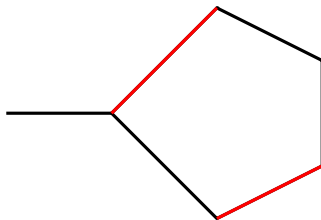


# General graphs

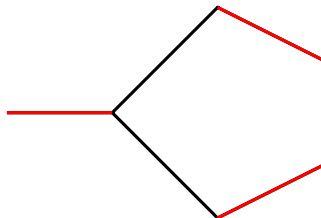
# General graphs

**Problem:** odd-length cycles

Maximal: matching of size 2

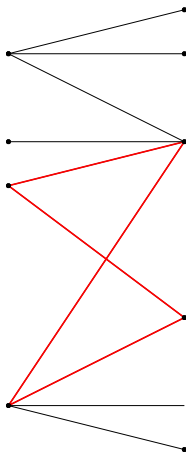


Optimum: matching of size 3



# General graphs

Bipartite graphs can have cycles, but always only of even length:



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**Output:** maximum matching  $\mathcal{M}^*$  on  $\mathcal{G}$

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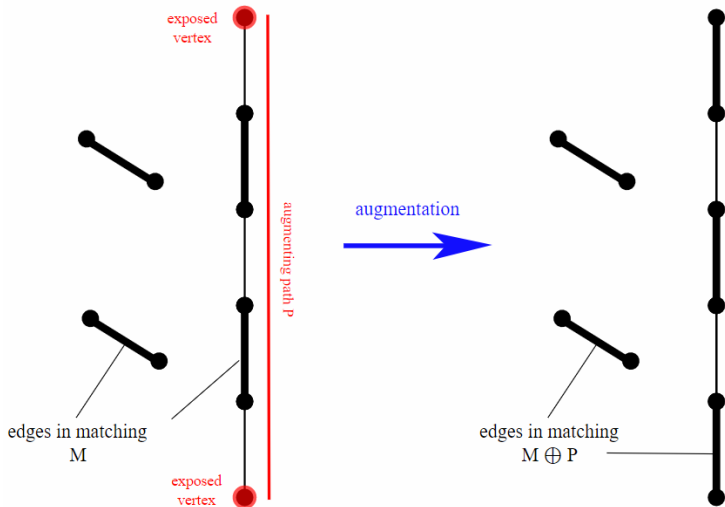
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In other words, Blossom algorithm improves existing matching  $\mathcal{M}$  in  $\mathcal{G}$  as long as augmenting paths exist, then returns.

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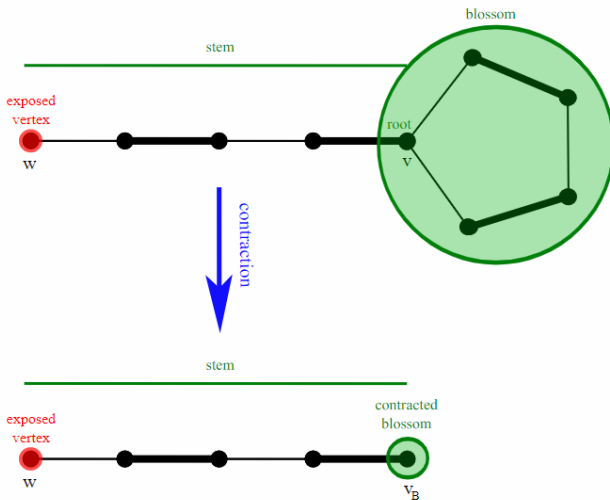
Vertex  $v$  is exposed iff no edge of  $M$  is incident with  $v$ .

Blossom algorithm examines all exposed vertices  $v$  and

- if an augmenting path is found, improve matching
- if an odd cycle ("blossom") is found, temporarily remove cycle and execute algorithm on a modified graph
- (delete vertex  $v$  from exposed vertices)

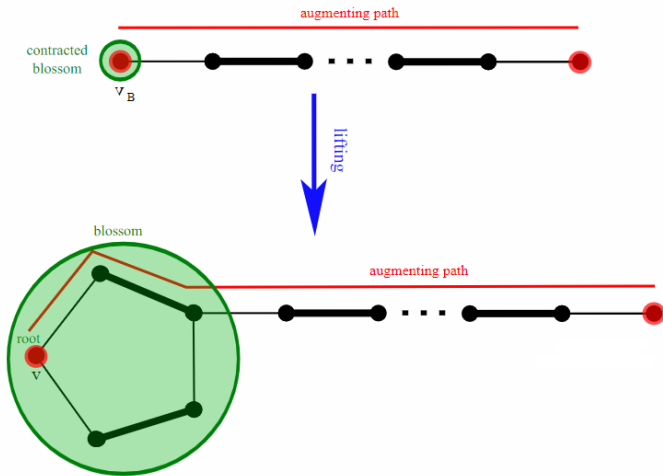
# Edmonds' Blossom algorithm (1965)

## Blossom contraction



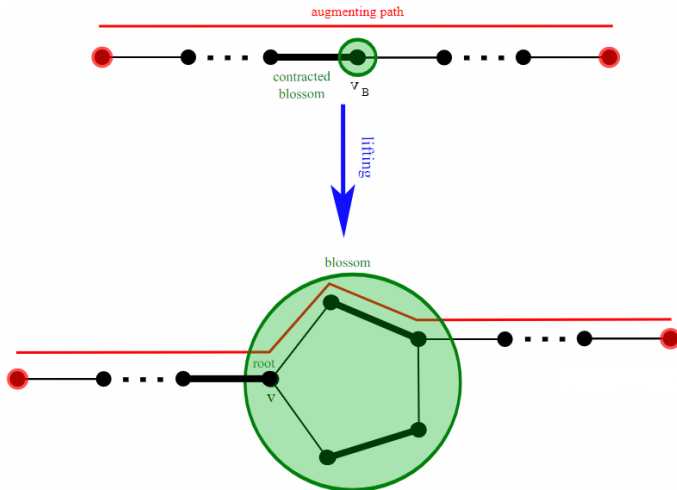
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## Blossom lift



# Edmonds' Blossom algorithm (1965)

## Blossom lift (other variant)



**Complexity:** For general graphs a straightforward implementation of the maximum matching algorithm of Edmonds (1965) runs in  $O(n^4)$  time (Papadimitriou and Steiglitz, 1982). More efficient general matching algorithms have been designed with the following running times:

- $O(n^3)$  - Gabow, 1976,
- $O(n * m)$  - Kameda and Munro, 1974,
- $O(n^{2.5})$  - Even and Kariv, 1975,
- $O(n^{1/2} * m)$  - Micali and Vazirani, 1980

# Randomized Algorithms



# Determinant and Permanent

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$i$	1	2	3	4	5
$\delta(i)$	3	2	4	1	5

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Efficient Gaussian Elimination

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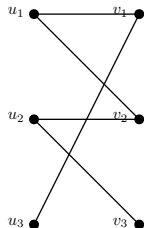
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**NP-Hard** problem.

Presumably no polynomial time algorithm to compute.

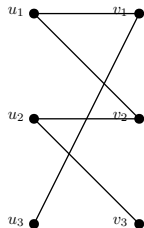
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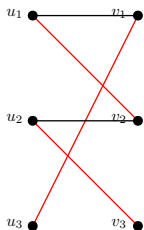
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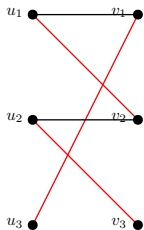
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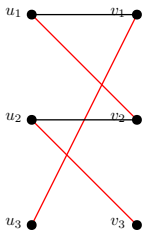
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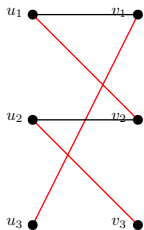
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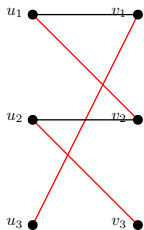
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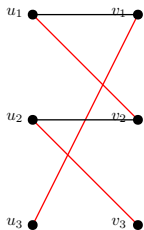
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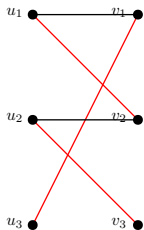
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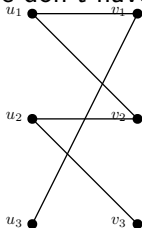
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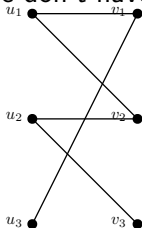
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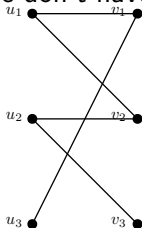
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The graph admits a perfect matching  $\iff$  the polynomial  $\det(A'^G)$  is not identical zero.

# Schwartz-Zippel lemma

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Let  $\rho$  be a non-zero polynomial of  $n$  variables and degree  $d$  over a field  $\mathbb{F}$ .  
Let  $\mathcal{S} \subseteq \mathbb{F}^n$ , then for  $x_0 \underset{\text{u.a.r}}{\in} \mathcal{S}$

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*u.a.r*

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Choosing  $|\mathcal{S}| = 2d$ , we get a method that tells with probability at least  $1/2$ , if the polynomial is identical zero.

Since one non-zero answer is enough to know the polynomial is a non-zero, we can repeat the operation a couple of times magnifying the probability.