

# Bipartite Graphs

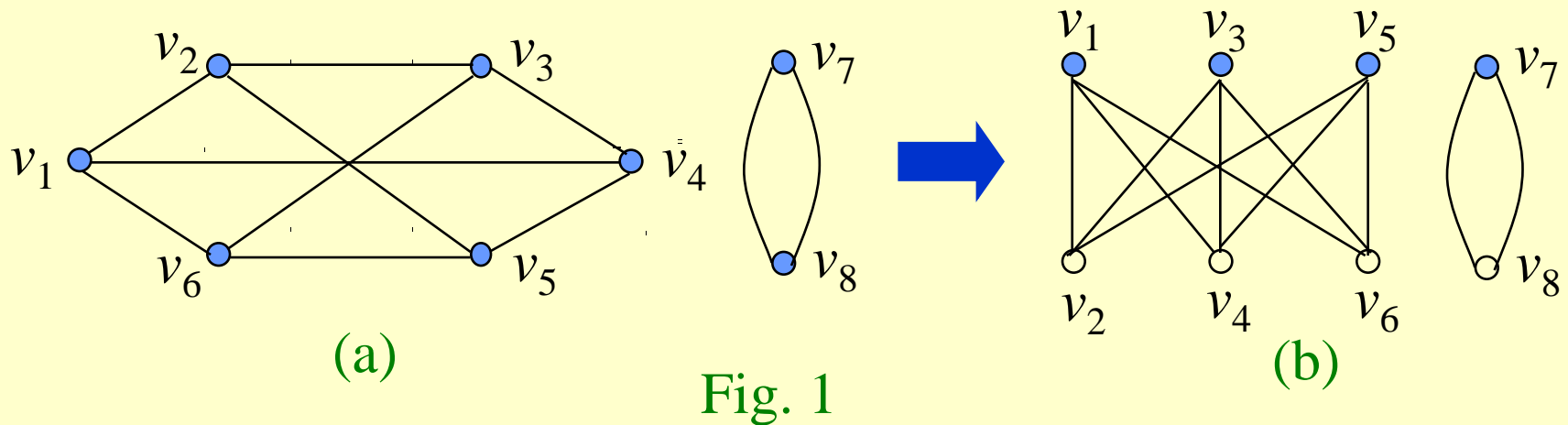
- What is a bipartite graph?
- Properties of bipartite graphs
- Matching and maximum matching
  - alternative paths
  - augmenting paths
- Hopcroft-Karp algorithm

# Bipartite Graph

- 1. A graph  $G$  is *bipartite* if the node set  $V$  can be partitioned into two sets  $V_1$  and  $V_2$  in such a way that no nodes from the same set are adjacent.
- 2. The sets  $V_1$  and  $V_2$  are called the *color classes* of  $G$  and  $(V_1, V_2)$  is a bipartition of  $G$ . In fact, a graph being bipartite means that the nodes of  $G$  can be colored with at most two colors, so that no two adjacent nodes have the same color.

# Bipartite Graph

3. We will depict bipartite graphs with their nodes colored black and white to show one possible bipartition.
4. We will call a graph *m by n bipartite*, if  $|V_1| = m$  and  $|V_2| = n$ , and a graph a *balanced bipartite graph* when  $|V_1| = |V_2|$ .



# Properties

**Property 3.1** A connected bipartite graph has a unique bipartition.

**Property 3.2** A bipartite graph with no isolated nodes and  $p$  connected components has  $2^{p-1}$  bipartitions.

For example, the bipartite graph in the above figure has two bipartitions. One is shown in the figure and the other has  $V_1 = \{v_1, v_3, v_5, v_8\}$  and  $V_2 = \{v_2, v_4, v_6, v_7\}$ .

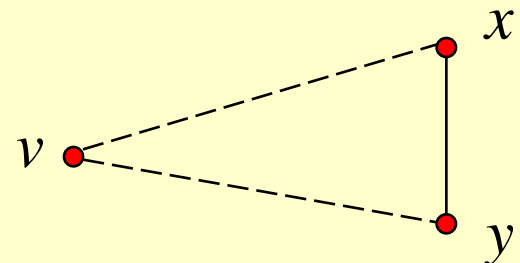
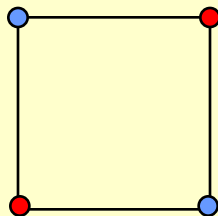
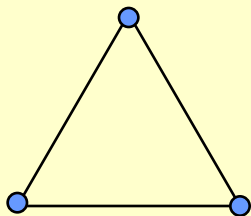
# Properties

The following theorem belongs to König (1916).

**Theorem 3.3** A graph  $G$  is bipartite if and only if  $G$  has no cycle of odd length.

**Corollary 3.4** A connected graph is bipartite if and only if for every node  $v$  there is no edge  $(x, y)$  such that  $distance(v, x) = distance(v, y)$ .

**Corollary 3.5** A graph  $G$  is bipartite if and only if it contains no closed walk of odd length.

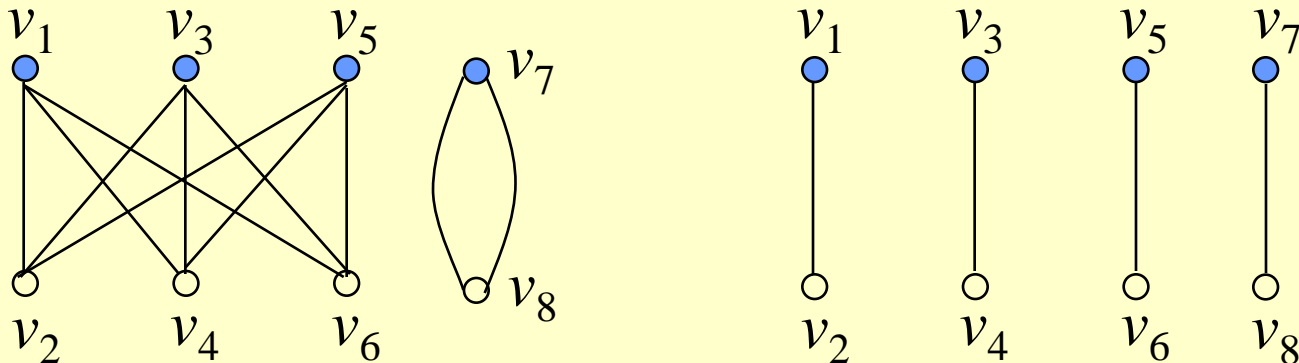


# Matching and Maximum Matching

## ■ Maximum matching

- A set of edges in a bipartite graph  $G$  is called a *matching* if no two edges have a common end node.
- A matching with the largest possible number of edges is called a *maximum matching*.

Example. A maximum matching for the bipartite graph in Fig. 1(b) is shown below.



# Maximum Matching

## ■ Maximum matching

- Many discrete problems can be formulated as problems about maximum matchings. Consider, for example, probably the most famous:

A set of boys each know several girls, is it possible for the boys each to marry a girl that he knows?

This situation has a natural representation as a bipartite graph with bipartition  $(V_1, V_2)$ , where  $V_1$  is the set of boys,  $V_2$  the set of girls, and an edge between a boy and a girl represents that they know one another. The marriage problem is then the problem: does a maximum matching of  $G$  have  $|V_1|$  edges?

# Alternative and Augmenting Path

## ■ Maximum matching

Let  $M$  be a matching of a graph  $G$ .

- A node  $v$  is said to be *covered*, or *saturated* by  $M$ , if some edge of  $M$  is incident with  $v$ . We will also call an unsaturated node *free*.
- A path or cycle is *alternative*, relative to  $M$ , if its edges are alternatively in  $M$  and  $E \setminus M$ .
- A path is an *augmenting path* if it is an alternating path with free origin and terminus.
- $P$  - a path.  $C$  - a cycle.  
 $|P|$  - the number of edges in  $P$   
 $|C|$  - the number of edges in  $C$ .

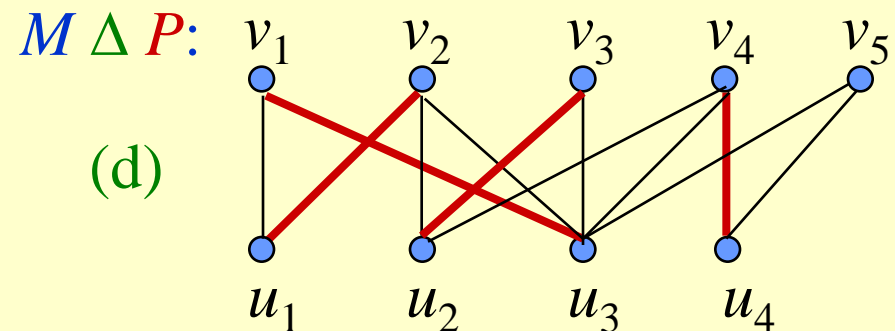
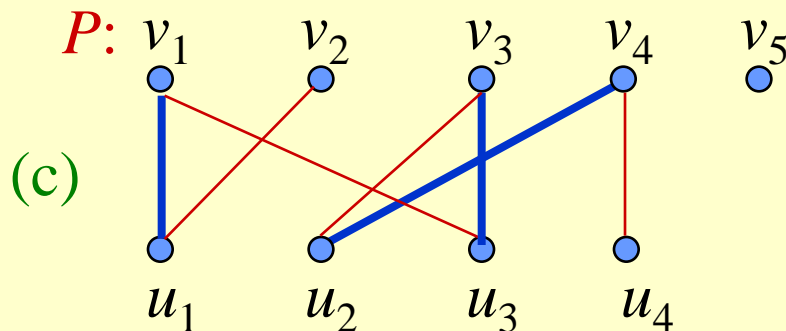
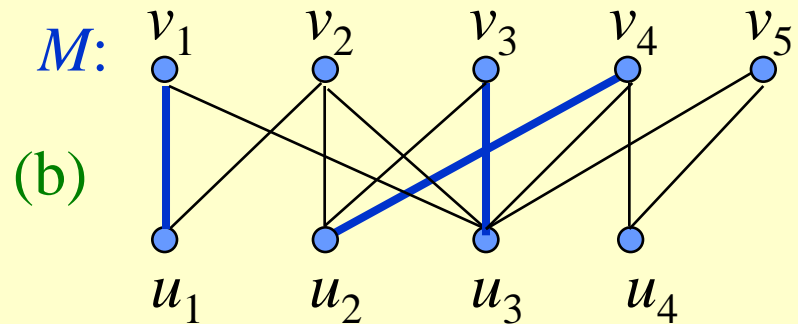
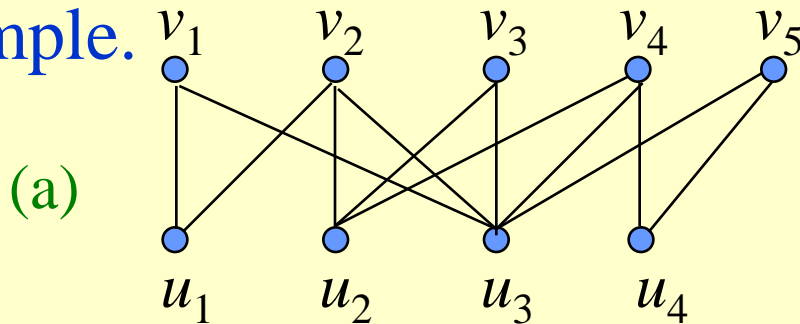


# Properties of Matchings

**Property 3.3** Let  $M$  be a matching and  $P$  an augmenting path relative to  $M$ . Then the symmetric difference of  $M$  and  $P$ , denoted  $M \Delta P$ , is also a matching of  $G$  and  $|M \Delta P| = |M| + 1$ .

$$M \Delta P = (M \setminus P) \cup (P \setminus M)$$

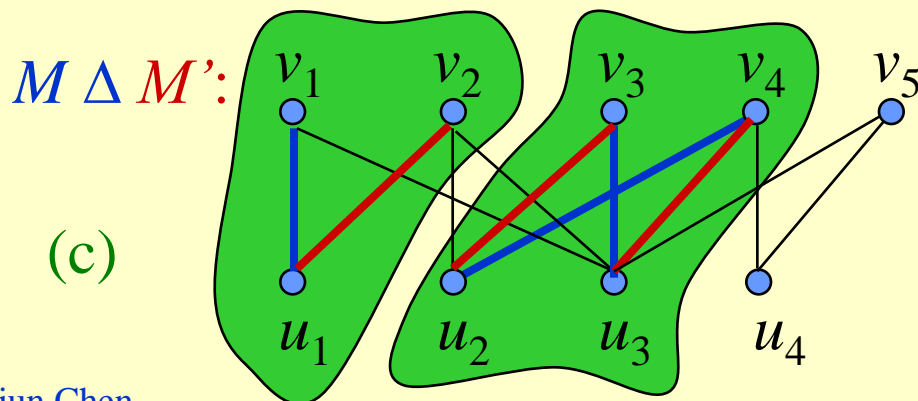
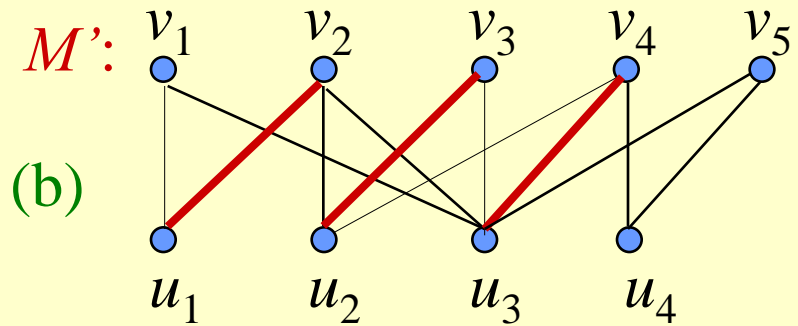
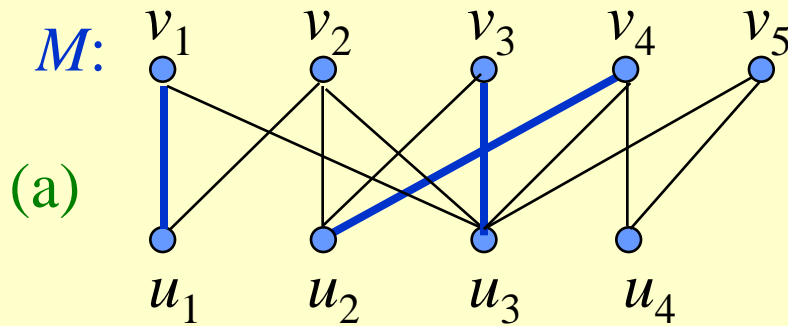
Example.

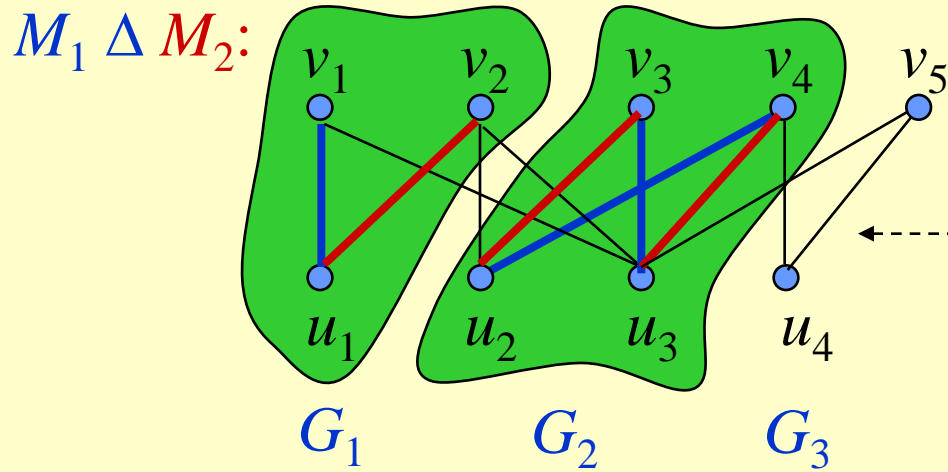
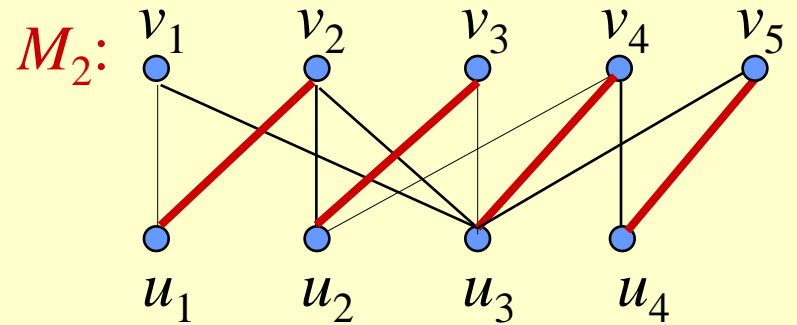
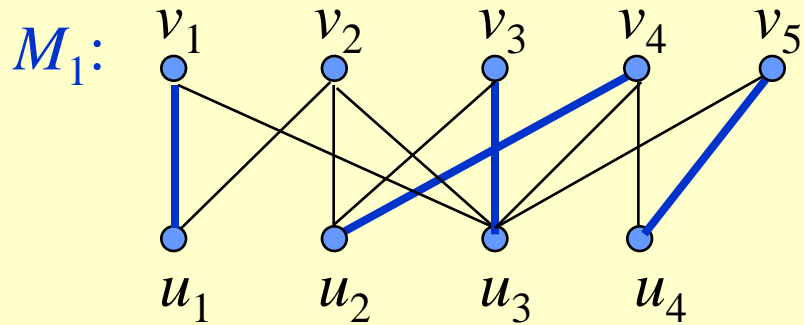


# Properties of Matchings

**Property 3.4** Let  $M$  and  $M'$  be matchings in  $G$ . Then, each connected component of  $M \Delta M'$  is one of the following:

- (1) an even cycle with edges alternatively in  $M \setminus M'$  and  $M' \setminus M$ ,  
or
- (2) a path whose edges are alternatively in  $M \setminus M'$  and  $M' \setminus M$ .





This edge is considered to be an augmenting path relative to  $M_1$ , also to  $M_2$ .

# Properties of Matchings

**Proposition 3.5** Let  $M$  and  $M'$  be matchings in  $G$ . If  $|M| = r$ ,  $|M'| = s$  and  $s > r$ , then  $M \Delta M'$  contains at least  $s - r$  node-disjoint augmenting paths relative to  $M$ .

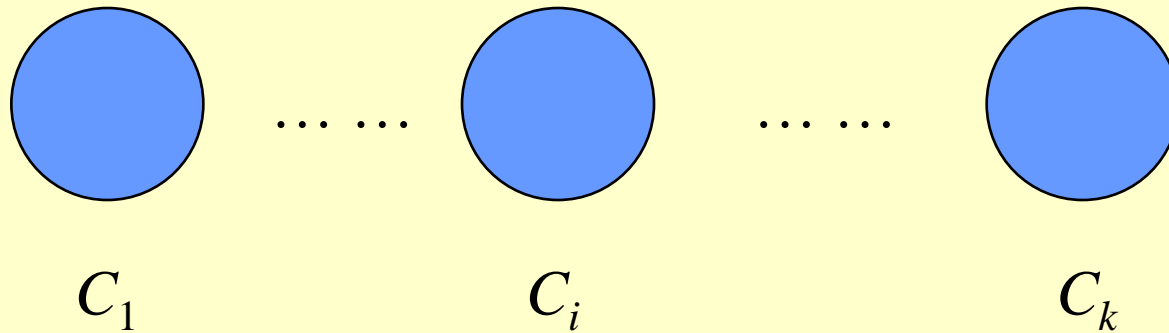
*Proof.* Let the components of  $M \Delta M'$  be  $C_1, C_2, \dots, C_k$ . Let  $f(C_i) = |C_i \cap M'| - |C_i \cap M|$  ( $1 \leq i \leq k$ ). Then it follows from Property 3.4 that  $f(C_i) \in \{-1, 0, 1\}$  for each  $1 \leq i \leq k$ .  $f(C_i) = 1$  if and only if  $C_i$  is an augmenting path relative to  $M$ . To complete the proof, we need only observe that

$$\sum_{j=1}^k f(C_j) = |M \setminus M| - |M \setminus M'| = |M'| - |M| = s - r.$$

Thus, there are at least  $s - r$  components with  $f(C_i) = 1$ , and at least  $s - r$  node-disjoint augmenting paths relative to  $M$ .

$C_i$  contains one more edge from  $M'$  or from  $M$ .

$M \Delta M'$



If  $C_i$  is a cycle or an even path,  $f(C_i) = 0$ .

If  $C_i$  is an augmenting path relative to  $M'$ ,  $f(C_i) = -1$ .

If  $C_i$  is an augmenting path relative to  $M$ ,  $f(C_i) = 1$ .

$$\sum_{j=1}^k f(C_j) = s - r$$

# Properties of Matchings

Combining Proposition 3.5 with Property 3.3, we can deduce the following theorem (Berge, 1957).

**Theorem 3.6**  $M$  is a maximum matching if and only if there is no augmenting path relative to  $M$ .

*Proof. If-part.* If there is no augmenting path,  $M$  must be a maximum matching. Otherwise, let  $M'$  be a maximum matching. According to Proposition 3.5,  $M \Delta M'$  contains at least  $|M'| - |M|$  node-disjoint augmenting paths relative to  $M$ . Contradiction.

*Only-if-part.* If  $M$  is a maximum matching, there is definitely no augmenting path relative to  $M$ . Otherwise, let  $P$  be an augmenting path relative to  $M$ . Then,  $M \Delta P$  is a larger matching than  $M$ .

# Properties of Matchings

The above theorem implies the following property.

**Property 3.7** If  $M$  is a matching of  $G$ , then there exists a maximum matching  $M$  of  $G$  such that the set of nodes covered by  $M$  is also covered by  $M$ .

*Proof.* If  $M$  is a maximum matching, the property trivially holds. Otherwise, consider an augmenting path  $P$  relative to  $M$ . Then, according to Property 3.1,  $M \Delta P$  is also a matching of  $G$  with

$$|M \Delta P| = |M| + 1.$$

Moreover, the nodes covered by  $M$  are also covered by  $M \Delta P$ . Repeating the above process will prove the property.

# Properties of Matchings

The following theorem was obtained by Dulmage and Mendelsohn (1958).

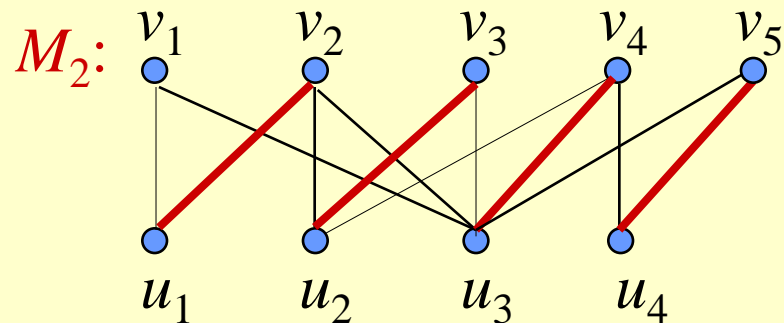
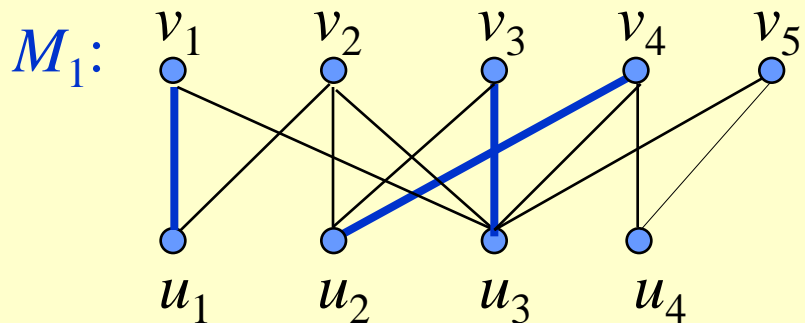
**Theorem 3.8** Let  $G$  be a bipartite graph with bipartition  $(V_1, V_2)$ . Let  $M_1$  and  $M_2$  be matchings in  $G$ . Then, there is a matching  $M \subseteq M_1 \cup M_2$  such that  $M$  covers all the nodes of  $V_1$  covered by  $M_1$  and all the nodes of  $V_2$  covered by  $M_2$ .

*Proof.* Let  $U_i$  be the nodes of  $V_1$  covered by  $M_i$  ( $i = 1, 2$ ). Let  $W_i$  be the nodes of  $V_2$  covered by  $M_i$  ( $i = 1, 2$ ). Let  $G_1, G_2, \dots, G_k$  be the connected components of  $M_1 \Delta M_2$ . By Property 3.4, each  $G_i$  ( $1 \leq i \leq k$ ) is an even cycle or a path. Let  $M_{1i} = G_i \cap M_1$  and  $M_{2i} = G_i \cap M_2$ . Define in each  $G_i$  a matching  $\pi_i$ :

$$\pi_i = \begin{cases} M_{1i} & \text{if } G_i \text{ is a cycle} \\ M_{1i} & \text{if there is a node } v \in V \cap (U_1 \setminus U_2) \text{ in } G_i \\ M_{2i} & \text{if there is a node } v \in V \cap (W_2 \setminus W_1) \text{ in } G_i \end{cases}$$

Then, it is not difficult to check that  $M = (M_1 \cap M_2) \cup \pi_1 \cup \pi_2 \dots \cup \pi_k$  is the required matching.



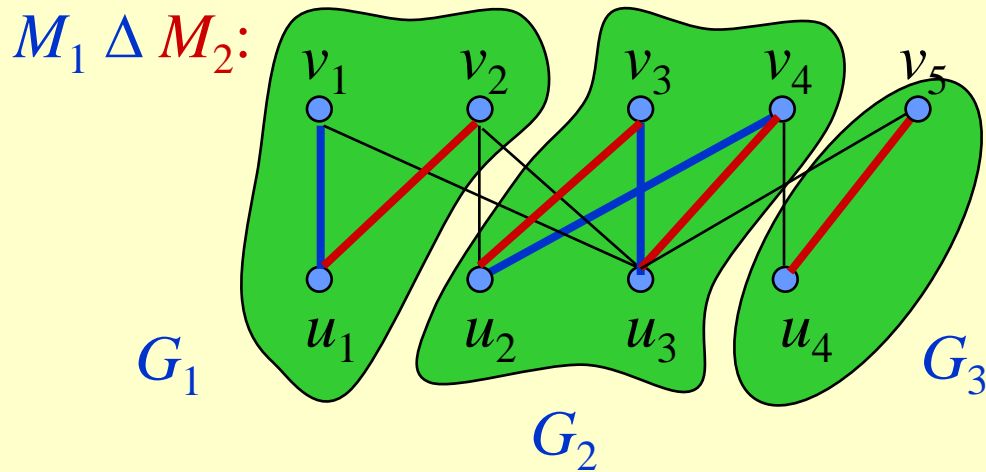


$$U_1 = \{v_1, v_3, v_4\}$$

$$U_2 = \{v_2, v_3, v_4, v_5\}$$

$$W_1 = \{u_1, u_2, u_3\}$$

$$W_2 = \{u_1, u_2, u_3, u_4\}$$



$$\pi_1 = M_{11} = G_1 \cap M_1$$

$$\pi_2 = M_{12} = G_2 \cap M_1$$

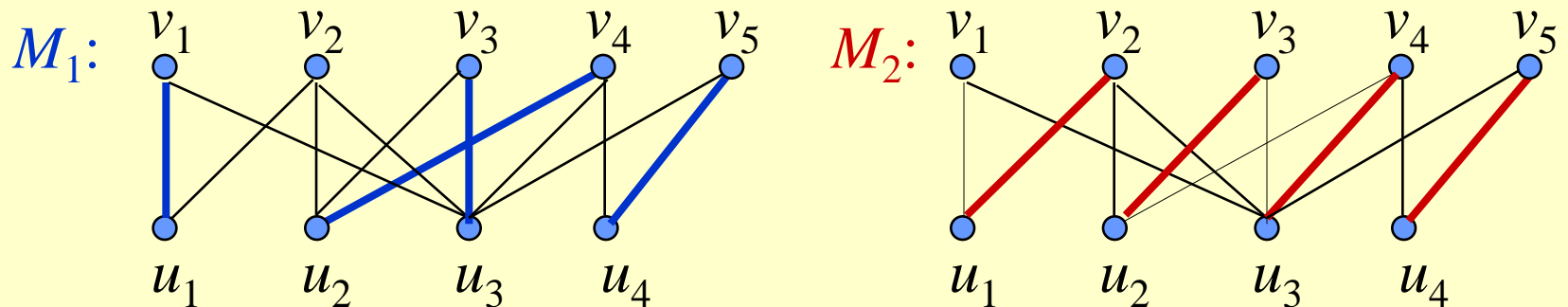
$$\pi_3 = M_{23} = G_3 \cap M_2$$

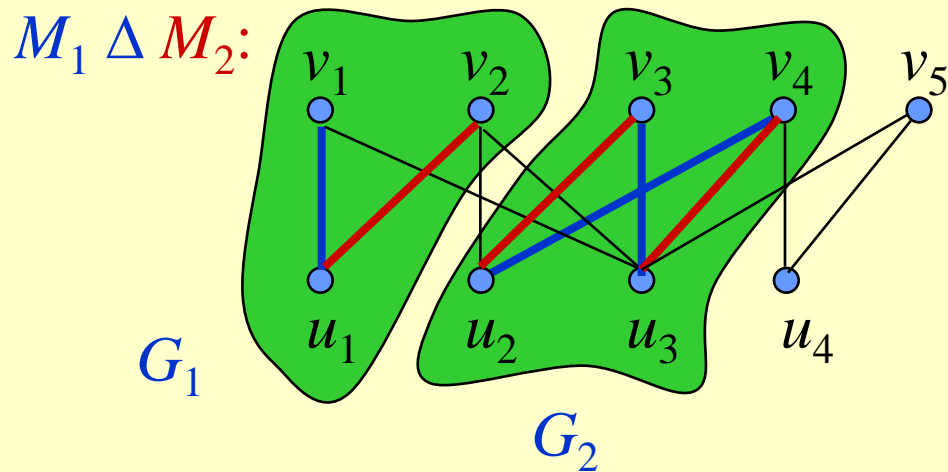
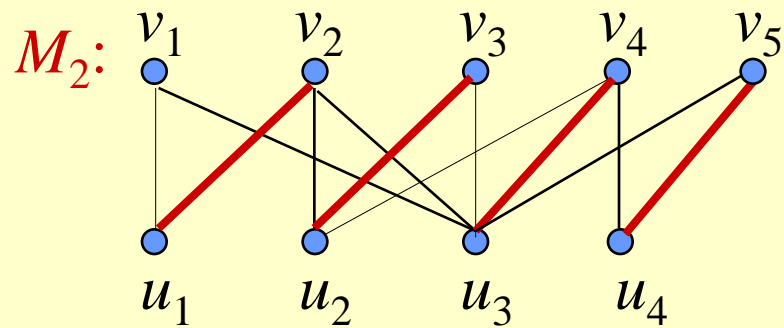
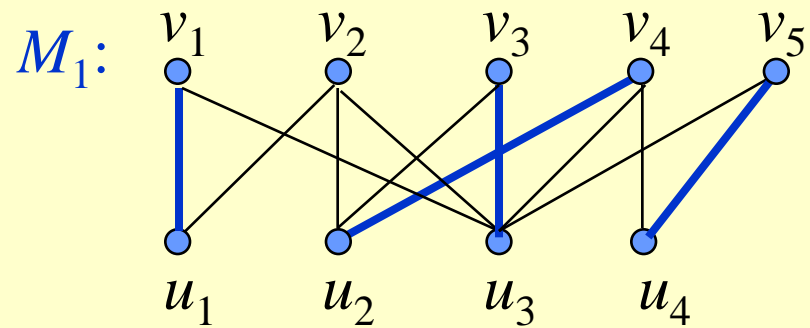
$$M = (M_1 \cap M_2) \cup \pi_1 \cup \pi_2 \cup \pi_3$$

# Properties of Matchings

**Theorem 3.9** A maximum matching  $M$  of a bipartite graph  $G$  can be obtained from any other maximum matching  $M'$  by a sequence of transfers along alternating cycles and paths of even length.

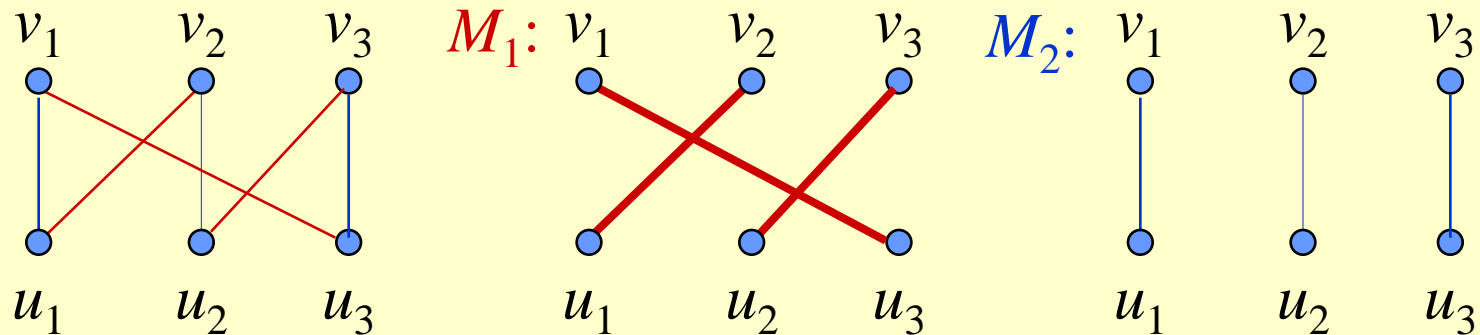
*Proof.* By Property 3.4, every component of  $M \Delta M'$  is an alternating even cycle or an alternating path relative to  $M'$ . By Property 3.3 and Theorem 3.6, a component of  $M \Delta M'$ , if it is a path, must be of even length. (Otherwise, if it is an odd path, it must be an augmenting path relative to  $M$  or to  $M'$ , contradicting the fact that both  $M$  and  $M'$  are maximum.) Then, changing  $M'$  in each component in turn will transform  $M'$  into  $M$ .





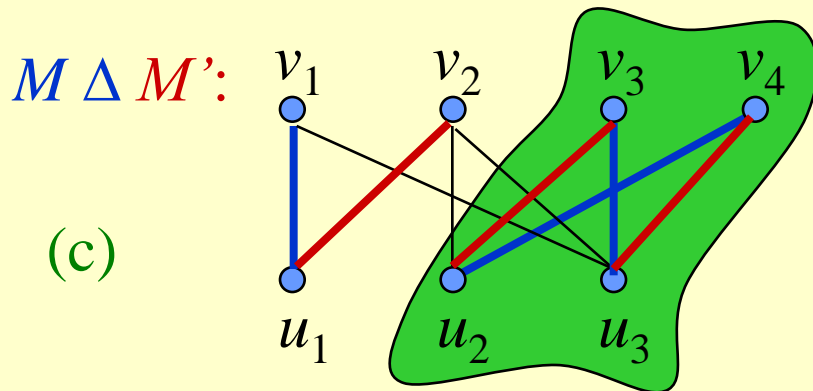
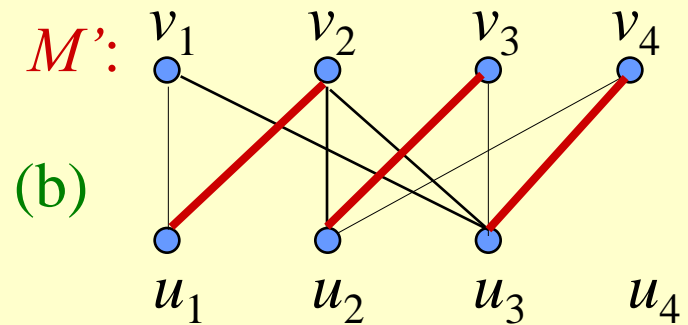
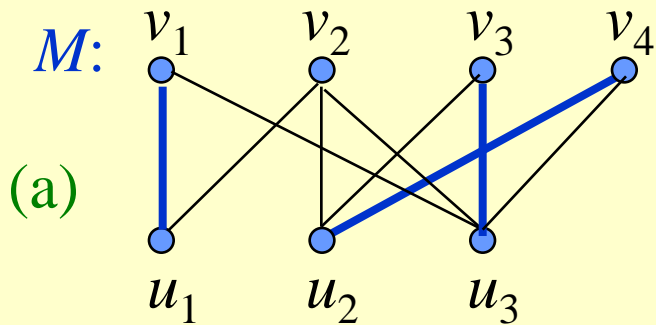
# Properties of Matchings

A perfecting matching of a graph  $G$  is a matching which covers every node of  $G$ . Clearly, if a graph has two perfect matchings  $M$  and  $M'$ , all components of  $M \Delta M'$  are even cycles. Therefore, according to Theorem 3.9, we can deduce the following result.



**Corollary 3.10** Assume that bipartite graph  $G$  has a perfect matching  $M$ . Then, any other perfect matching can be obtained from  $M$  by a sequence of transfers along alternating cycles relative to  $M$ .

Maximum, but not perfect matching



# Algorithm

## ■ Algorithms - finding a maximum matching

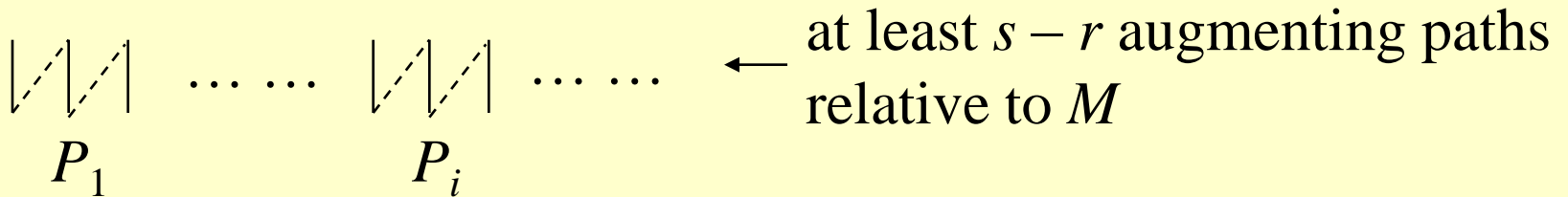
**Lemma 3.11** Let  $M$  be a matching with  $|M| = r$  and suppose that the cardinality of a maximum matching is  $s$ . Then, there exists an augmenting path relative to  $M$  of length at most  $2\lfloor r/(s - r) \rfloor + 1$ .

*Proof.* Let  $M'$  be a maximum matching. Then, by Proposition 3.5,  $M \Delta M'$  contains  $s - r$  node-disjoint augmenting paths relative to  $M$ . It is easy to see that these paths contain at most  $r$  edges from  $M$ . So one of these augmenting paths must contain at most  $\lfloor r/(s - r) \rfloor$  edges from  $M$  and so at most  $2\lfloor r/(s - r) \rfloor + 1$  edges altogether.

*Proof.* Let  $M'$  be a maximum matching. Then, by Proposition 3.5,  $M \Delta M'$  contains  $s - r$  node-disjoint augmenting paths relative to  $M$ . It is easy to see that these paths contain at most  $r$  edges from  $M$ . So one of these augmenting paths must contain at most  $\lfloor r/(s - r) \rfloor$  edges from  $M$  and so at most  $2 \lfloor r/(s - r) \rfloor + 1$  edges altogether.

Since  $s > r$ , there is at least  $s - r$  augmenting paths in  $M \Delta M'$ . On each of them the number of edges in  $M'$  is one larger than the number of edges in  $M$ .

Consider these augmenting paths:



If each  $P_i$  contains more than  $\lfloor r/(s - r) \rfloor$  edges, then  $M$  will have more than  $r$  edges. Contradiction.

# Algorithm

**Lemma 3.12** Let  $M$  be a matching and  $P$  be a shortest augmenting path relative to  $M$ . Let  $Q$  be an augmenting path relative to  $M \Delta P$ . Then,  $|Q| \geq |P| + 2|P \cap Q|$ .

*Proof.* Consider  $M' = M \Delta P \Delta Q$ . Then,  $M'$  is a matching. By Property 3.3,  $|M'| = |M \Delta P| + 1 = (|M| + 1) + 1 = |M| + 2$ . According to Proposition 3.5,  $M \Delta M'$  contains at least two node-disjoint augmenting paths relative to  $M$ . Let  $P_1, \dots, P_k$  ( $k \geq 2$ ) be such paths. Since  $P$  is a shortest augmenting path relative to  $M$ , we have

$$\begin{aligned} |M \Delta M'| &= |M \Delta (M \Delta P \Delta Q)| = |\emptyset \Delta P \Delta Q| \\ &= |P \Delta Q| \geq |P_1| + \dots + |P_k| \geq 2|P|. \end{aligned}$$

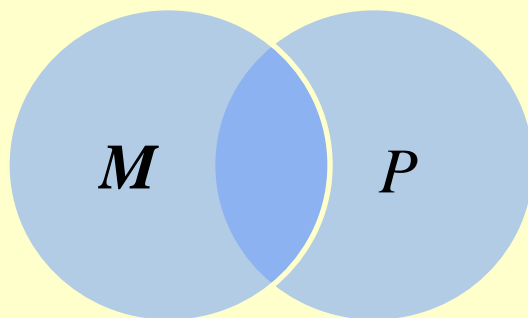
Note that  $|P \Delta Q| = |P| + |Q| - 2|P \cap Q| \geq 2|P|$ . Therefore, we have  $|Q| \geq |P| + 2|P \cap Q|$ .



## About symmetric difference

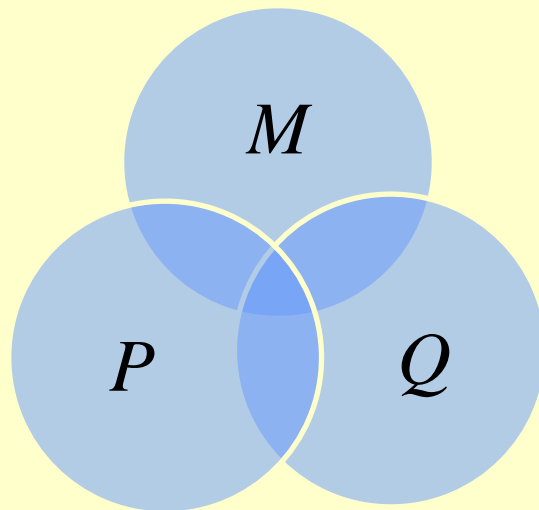
*Commutative:*

$$M \Delta P = P \Delta M$$



*Associative:*

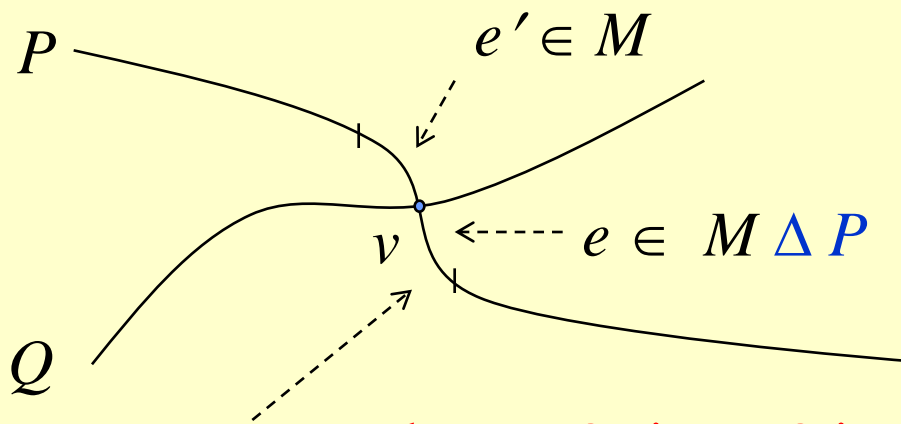
$$\begin{aligned} &M \Delta P \Delta Q \\ &= M \Delta (P \Delta Q) \\ &= (M \Delta P) \Delta Q \end{aligned}$$



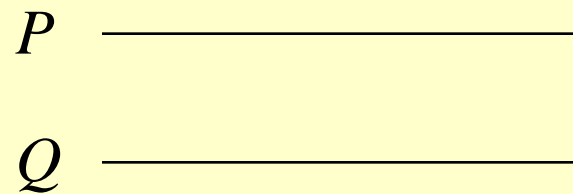
# Algorithm

**Corollary 3.13** Let  $P$  be a shortest augmenting path relative to a matching  $M$ , and  $Q$  be a shortest augmenting path relative to  $M \Delta P$ . Then, if  $|P| = |Q|$ , the paths  $P$  and  $Q$  must be node-disjoint. Moreover,  $Q$  is also a shortest augmenting path relative to  $M$ .

*Proof.* According to Lemma 3.12, we have  $|P| = |Q| \geq |P| + 2|P \cap Q|$ . So  $P \cap Q = \Phi$ . Thus,  $P$  and  $Q$  are edge-disjoint. Assume that  $P$  and  $Q$  share a common node  $v$ . Consider the edge  $e$  incident with  $v$  in  $M \Delta P$ . Then,  $P$  and  $Q$  must share  $e$ , contradicting  $P \cap Q = \Phi$ . Therefore,  $P$  and  $Q$  are also node-disjoint.



$P$  and  $Q$  are node-disjoint.



$e$  must be on  $Q$  since  $Q$  is an augmenting path of  $M \Delta P$ .

# Algorithm

## **Hopcroft-Karp algorithm (1973)**

The whole computation process is divided into a number of stages, at which some partial matching has been constructed and some way is sought to increase it. At stage  $i$ , we have the matching  $M_i$  and we search for  $\{Q_1, Q_2, \dots, Q_k\}$ , a maximal set of node-disjoint, shortest augmenting paths, relative to  $M_i$ . Then, according to Corollary 3.13,  $Q_2$  is a shortest augmenting path relative to  $M \Delta Q_1$ ,  $Q_3$  is a shortest augmenting path relative to  $(M \Delta Q_1) \Delta Q_2$ , ..., and  $Q_k$  is a shortest augmenting path relative to  $(M \Delta Q_1 \Delta Q_2 \dots \Delta Q_{k-2}) \Delta Q_{k-1}$ . Therefore, the new matching for the next stage is formed as

$$M_{i+1} = M_i \Delta Q_1 \Delta Q_2 \dots \Delta Q_k.$$

# Algorithm

**Proposition 3.14** Let  $s$  be the cardinality of a maximum matching in a bipartite graph  $G$ . Then, to construct a maximum matching by the above process requires at most  $2 \lfloor \sqrt{s} \rfloor + 2$  stages.

*Proof.* It can be derived from Property 3.3 and Lemma 3.11.

Then, the running time should be bounded by  $O(|E|\sqrt{s})$  since at each stage  $O(|E|)$  edges will be visited.

## Proof of Proposition 3.14

Let  $M_0, M_1, \dots, M_s$  be a sequence of matchings in  $G$ , where  $M_0 = \Phi$  and  $M_{i+1} = M_i \Delta P_i$  and  $P_i$  is shortest augmenting path relative to  $M_i$  for each  $i = 1, \dots, s - 1$ . Then, an upper bound for the number of stages we require can be given by bounding the number of distinct integers in the sequence  $|P_0|, |P_1|, \dots, |P_s|$ . Let  $r = \lfloor s - \sqrt{s} \rfloor$ . Then, since  $|M_r| = r$ , we have

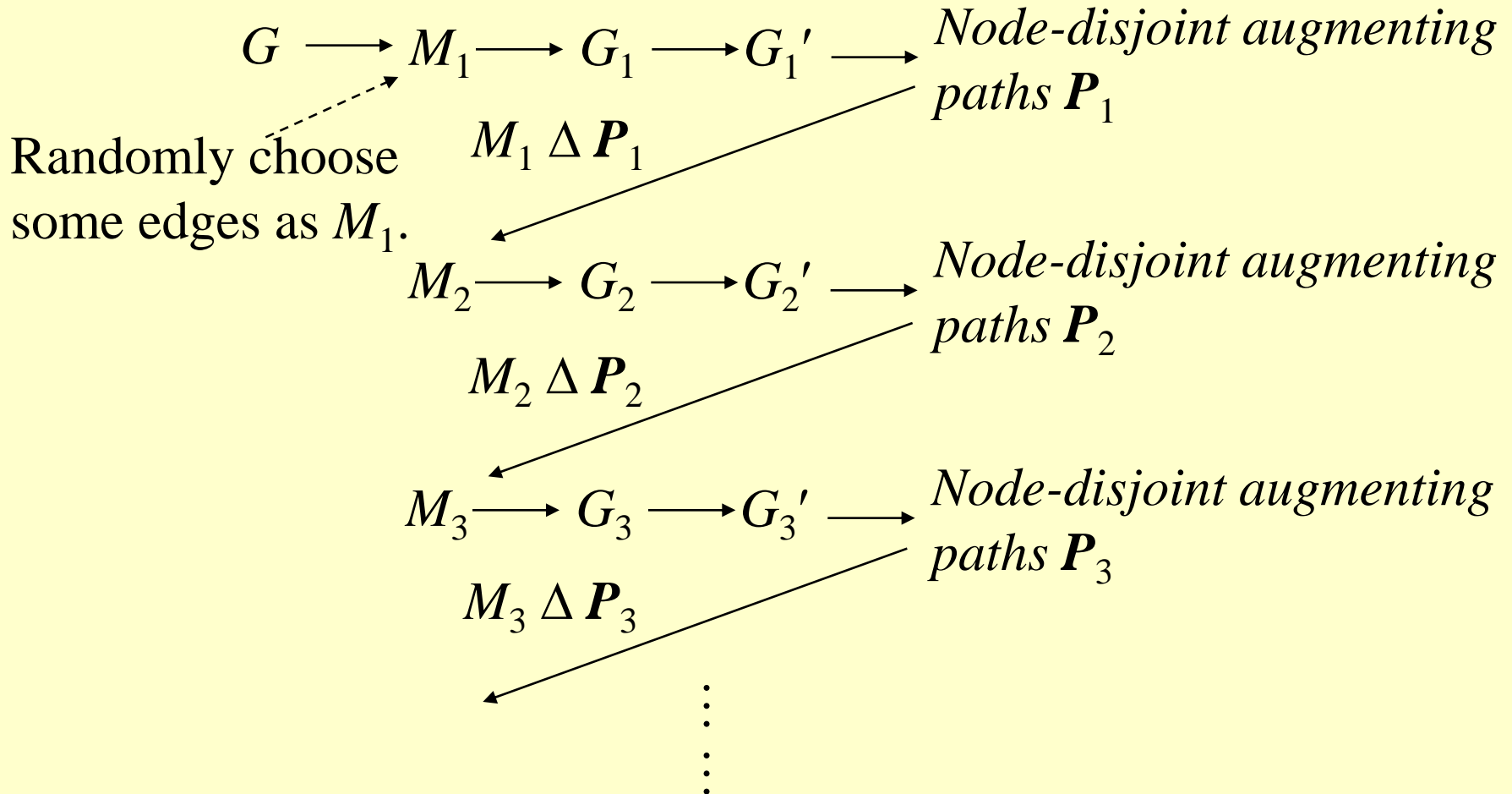
$$|P_r| \leq 2 \lfloor r/(s - r) \rfloor \leq 2 \lfloor \sqrt{s} \rfloor + 1.$$

Thus, for each  $i \leq r$ ,  $|P_i|$  is one of the  $\lfloor \sqrt{s} \rfloor + 1$  odd number less than or equal to  $2 \lfloor \sqrt{s} \rfloor + 1$ . Certainly,  $s - r = \lceil \sqrt{s} \rceil$  paths contribute at most  $\lceil \sqrt{s} \rceil$  distinct integers to the collection, and so the total number is at most

$$\lfloor \sqrt{s} \rfloor + 1 + \lceil \sqrt{s} \rceil \leq 2 \lfloor \sqrt{s} \rfloor + 2.$$

# Algorithm

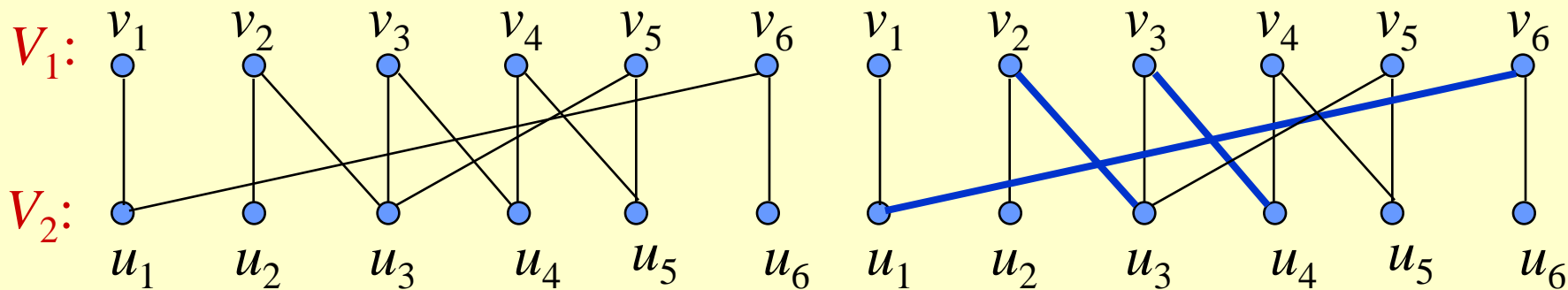
Main process:



# Algorithm

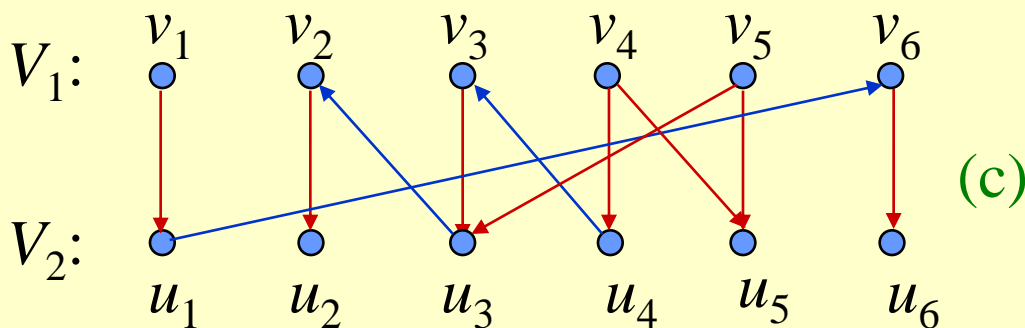
Let  $M_i$  be the matching of  $G$  produced at stage  $i$ . We define a directed graph  $G_i$  (called an *alternating graph*) with the same node set as  $G$ , but with edge set

$$E(G_i) = \{u \rightarrow v \mid u \in V_1, v \in V_2, \text{ and } (u, v) \in E \setminus M_i\} \\ \cup \{v \rightarrow u \mid u \in V_1, v \in V_2, \text{ and } (u, v) \in M_i\}.$$



(a)

(b)



(c)

# Algorithm

First step:

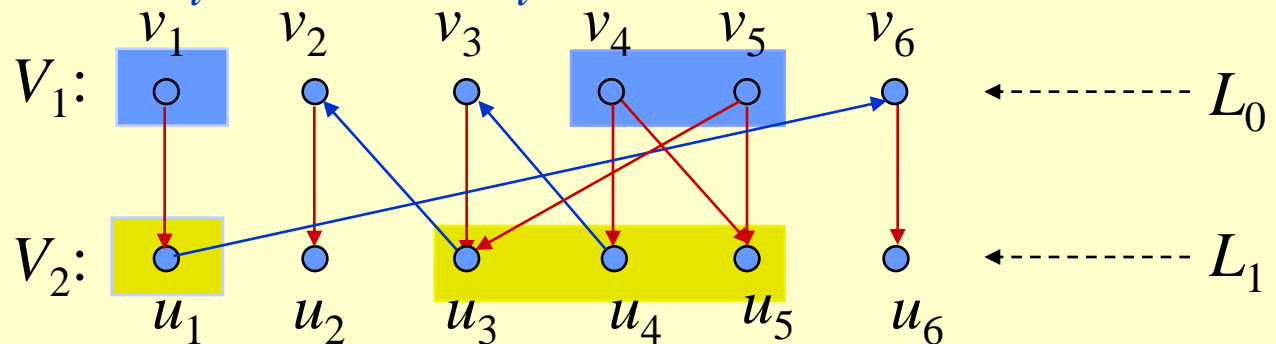
From  $G_i$ , construct a subgraph  $G_i'$  (called a *layered graph*) described below.

Let  $L_0$  be the set of free nodes (relative to  $M_i$ ) in  $V_1$  and define  $L_j$  ( $j > 0$ ) as follows:

$$E_{j-1} = \{u \rightarrow v \in E(G_i) \mid u \in L_{j-1}, v \notin L_0 \cup L_1 \cup \dots \cup L_{j-1}\},$$

$$L_j = \{v \in V(G_i) \mid \text{for some } u, u \rightarrow v \in E_{j-1}\}.$$

Define  $j^* = \min\{j \mid L_j \cap \{\text{free nodes in } V_2\} \neq \Phi\}$ .  $G_i'$  is formed with  $V(G_i')$  and  $E(G_i')$  defined below.





# Algorithm

First step:

If  $j^* = 1$ , then

$$V(G_1') = L_0 \cup (L_1 \cap \{\text{free nodes in } V_2\}),$$

$$E(G_1') = \{u \rightarrow v \mid u \in L_0 \text{ and } v \in \{\text{free nodes in } V_2\}\}.$$

If  $j^* > 1$ , then

$$V(G_i') = L_0 \cup L_1 \cup \dots \cup L_{j^*-1} \cup (L_{j^*} \cap \{\text{free nodes in } V_2\}),$$

$$E(G_i') = E_0 \cup E_1 \cup \dots \cup E_{j^*-2} \cup \{u \rightarrow v \mid u \in L_{j^*-1} \text{ and } v \in \{\text{free nodes in } V_2\}\}.$$

With this definition of the graph  $G_i'$ , directed paths from  $L_0$  to  $L_{j^*}$  are precisely in one-to-one correspondence with shortest augmenting paths relative to  $M_i$  in  $G$ .

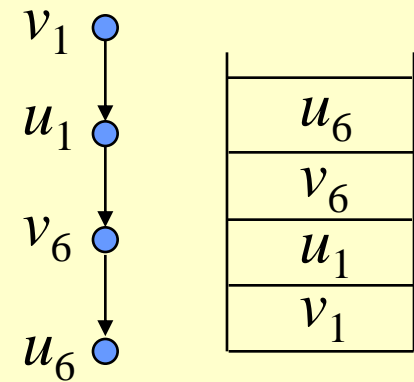
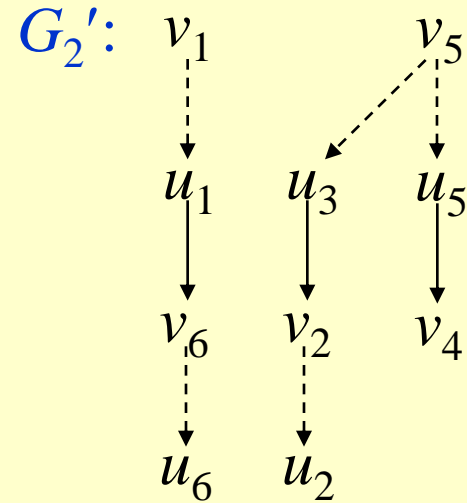
# Algorithm

Second step:

In this step, we will traverse  $G_i'$  in a depth-first searching fashion to find a maximal set of node-disjoint paths from  $L_0$  to  $L_{j^*}$ .

- For this, a stack structure *stack* is used to control the graph exploring.
- In addition, we use *c-list*( $v$ ) to represent the set of  $v$ 's child nodes.

# Algorithm



**Algorithm** *finding-augmenting-paths*( $G_i'$ )

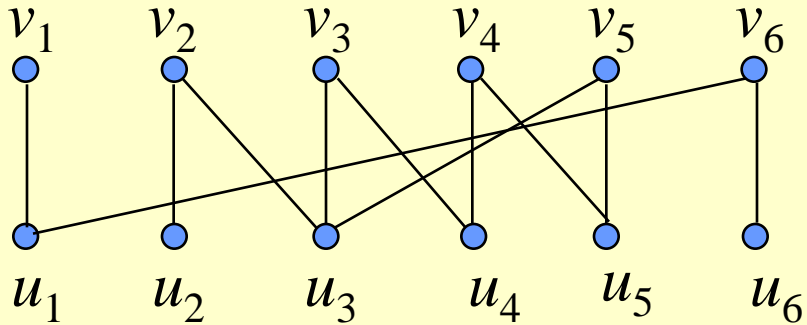
**begin**

1. let  $x$  be the first element in  $L_0$ ;
2. *push*( $x$ , *stack*); mark  $v$ ;
3. **while** *stack* is not empty **do** {
4.      $v := \text{top}(\text{stack})$ ;
5.     **while**  $c\text{-list}(v) \neq \Phi$  **do** {
6.         let  $u$  be the first element in  $c\text{-list}(v)$ ;
7.         **if**  $u$  is marked **then** remove  $u$  from  $c\text{-list}(v)$
8.         **else** {*push*( $u$ , *stack*); mark  $u$ ;  $v := u$ ;}
9.     }
10.    **if**  $v$  is neither in  $L_{j^*}$  nor in  $L_0$  **then** *pop*(*stack*)
11.    **else** {**if**  $v$  is in  $L_{j^*}$  **then** output all the elements in *stack*;
- (\*all the elements in *stack* make up an augmenting path.\*)
12.        remove all elements in *stack*;
13.        let  $v$  be the next element in  $L_0$ ;
14.        *push*( $v$ , *stack*); mark  $v$ ;
15.    }

**end**

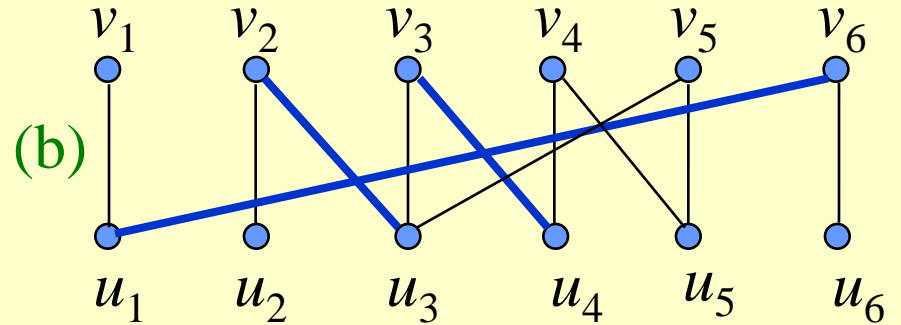
# Example Trace

Example.

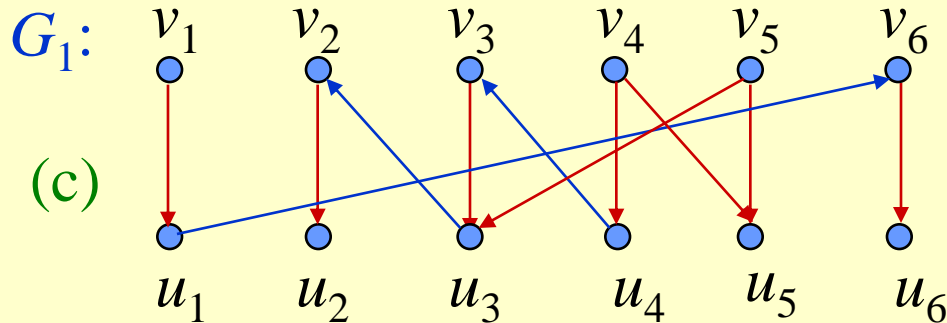


(a)

$M_1$ :



(b)



(c)

$G_1'$  will be constructed as follows:

$$L_0 = \{v_1, v_4, v_5\}$$

$$E_0 = \{(v_1, u_1), (v_4, u_4), (v_4, u_5), (v_5, u_3), (v_5, u_5)\}$$

$$L_1 = \{u_1, u_3, u_4, u_5\} \quad /*j* = 1 \text{ since } u_5 \text{ is free.}*/$$

(If  $u_5$  is not free, the following layers will be constructed.)

$$E_1 = \{(u_1, v_6), (u_3, v_2), (u_4, v_3)\}$$

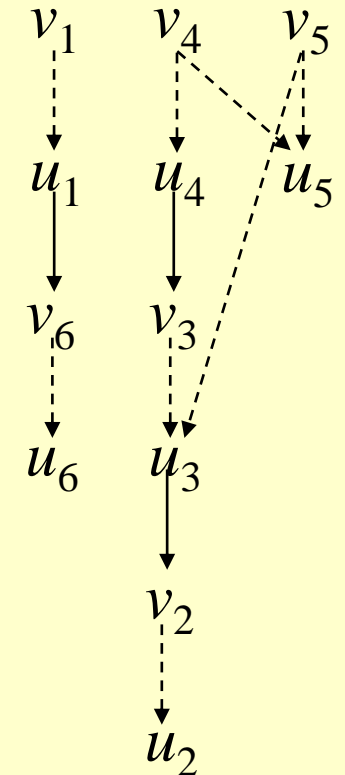
$$L_2 = \{v_6, v_2, v_3\}$$

$$E_2 = \{(v_6, u_6), (\cancel{v_3, u_3}), (v_2, u_2)\}$$

$$L_3 = \{u_6, u_2\}$$

$/*u_3$  is not in  $L_3$  since it already appears in  $L_1$  .

$G_1$ :



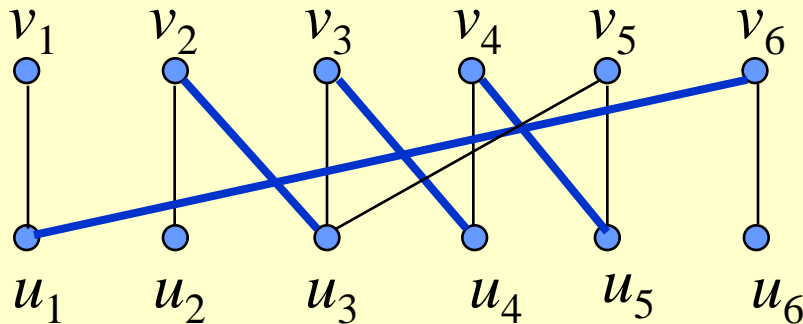
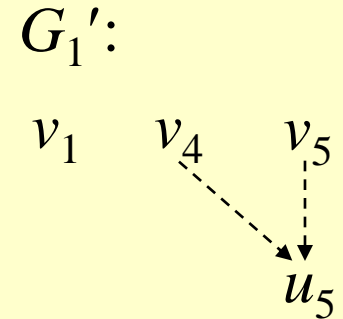
# Example Trace

Since  $L_1$  contains free node  $u_5$  in  $V_2$ ,  $j^* = 1$ .  
Therefore, we have

$$V(G_1') = \{v_1, v_4, v_5\} \cup \{u_5\}, \text{ and}$$

$$E(G_1') = \{(v_4, u_5), (v_5, u_5)\}$$

Note that  $v_4 \rightarrow u_5$  is an augmenting path relative to  $M_1$ , and  $v_5 \rightarrow u_5$  is another. By applying the second step of Hopcroft-Karp algorithm to  $G_1'$ ,  $v_4 \rightarrow u_5$  will be chosen, yielding a new matching  $M_2 = M_1 \Delta \{v_4 \rightarrow u_5\}$  as shown in the following figure.



# Example Trace

At a next stage, we will construct  $G_2$  as shown in Figure 7.  $G_2'$  will then be constructed as follows:

$$L_0 = \{v_1, v_5\},$$

$$E_0 = \{(v_1, u_1), (v_5, u_3), (v_5, u_5)\},$$

$$L_1 = \{u_1, u_3, u_5\},$$

$$E_1 = \{(u_1, v_6), (u_3, v_2), (u_5, v_4)\},$$

$$L_2 = \{v_2, v_4, v_6\},$$

$$E_2 = \{(v_2, u_2), (v_4, u_4), (v_6, u_6)\},$$

$$L_3 = \{u_2, u_4, u_6\},$$

*/\*j\* = 3 since  $u_2$  and  $u_6$  are free.\*/*

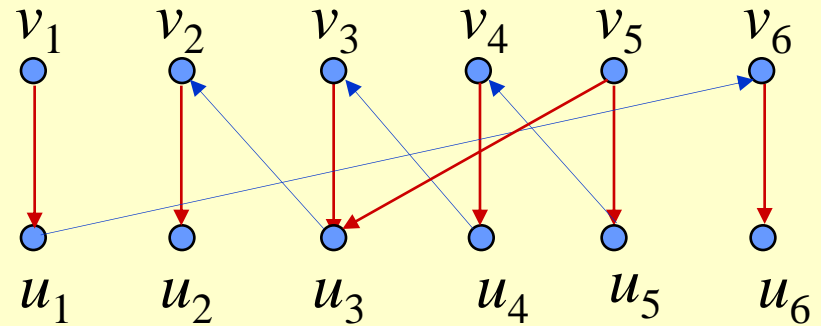
$$E_3 = \{(u_4, v_3)\},$$

$$L_4 = \{v_3\},$$

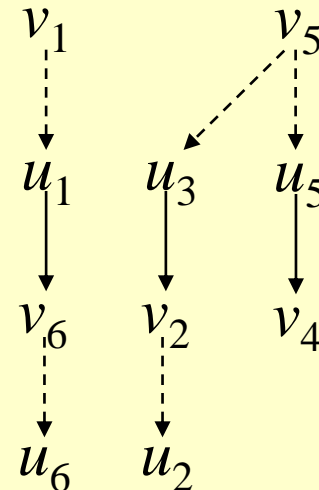
$$E_4 = \{(v_3, u_3)\}.$$

←----- This part will not be created.

$G_2$ :



$G_2'$ :



Since  $L_3$  contains two free nodes  $u_2$  and  $u_6$  in  $V_2$ ,  $j^* = 3$ .  
 So we have

$$V(G_2') = L_0 \cup L_1 \cup L_2 \cup \{u_2, u_6\}, \text{ and}$$

$$E(G_2') = E_0 \cup E_1 \cup \{(v_2, u_2), (v_6, u_6)\}.$$

$$L_0 = \{v_1, v_5\},$$

$$E_0 = \{(v_1, u_1), (v_5, u_3), (v_5, u_5)\},$$

$$L_1 = \{u_1, u_3, u_5\},$$

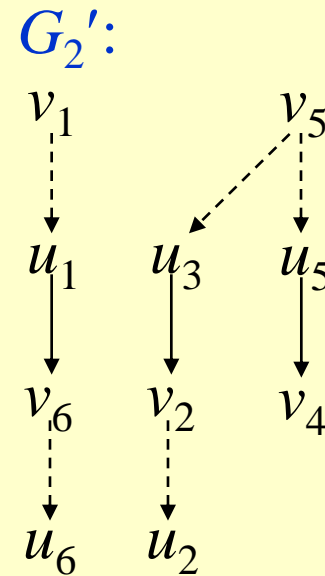
$$E_1 = \{(u_1, v_6), (u_3, v_2), (u_5, v_4)\},$$

$$L_2 = \{v_2, v_4, v_6\},$$

$$E_2 = \{(v_2, u_2), (v_6, u_6), (\cancel{v_4}, \cancel{u_4})\},$$

$$L_3 = \{u_2, u_6, \cancel{u_4}\}.$$

*/\*j\* = 3 since  $u_2$  and  $u_6$  are free.\*/\**





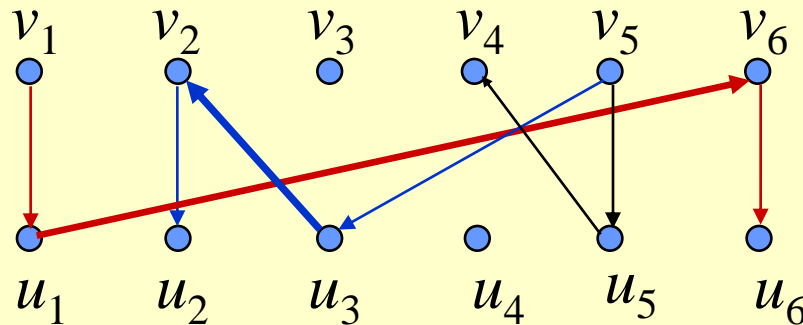
# Example Trace

Since  $L_3$  contains two free nodes  $u_2$  and  $u_6$  in  $V_2$ ,  $j^* = 3$ .  
So we have

$$V(G_2') = L_0 \cup L_1 \cup L_2 \cup \{u_2, u_6\}, \text{ and}$$

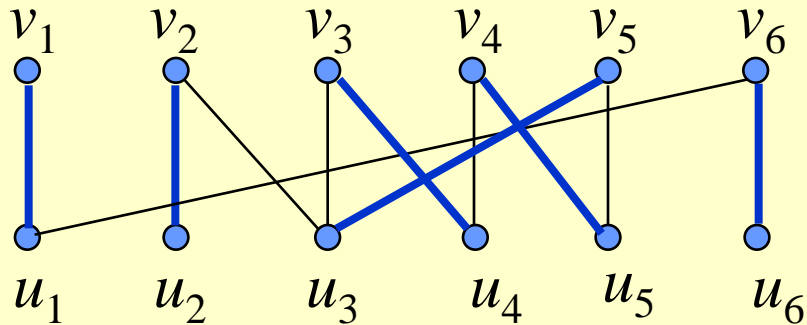
$$E(G_2') = E_0 \cup E_1 \cup \{(v_2, u_2), (v_6, u_6)\}.$$

In Fig. 8, we show  $G_2'$ , which contains two augmenting paths  $P_1$  and  $P_2$ , where  $P_1 = v_1 \rightarrow u_1 \rightarrow v_6 \rightarrow u_6$  (represented by **red** edges in Fig. 8) and  $P_2 = v_5 \rightarrow u_3 \rightarrow v_2 \rightarrow u_2$  (represented by **blue** edges in Fig. 8).

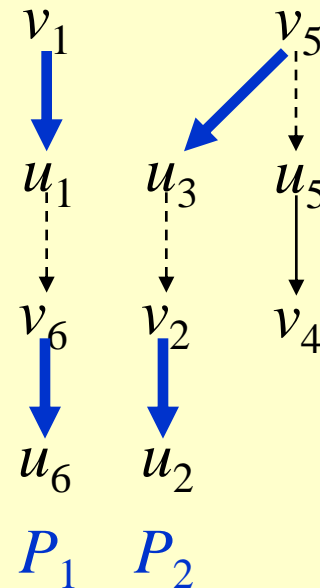


# Example Trace

In Fig. 8, we show  $G_2'$ , which contains two augmenting paths  $P_1$  and  $P_2$ , where  $P_1 = v_1 \rightarrow u_1 \rightarrow v_6 \rightarrow u_6$  and  $P_2 = v_5 \rightarrow u_3 \rightarrow v_2 \rightarrow u_2$ . By applying the second step of Hopcroft-Karp algorithm, these two augmenting paths will be found. The maximum matching  $M_3 = M_2 \Delta P_1 \Delta P_2$  is shown in Fig. 9.



$G_2'$ :

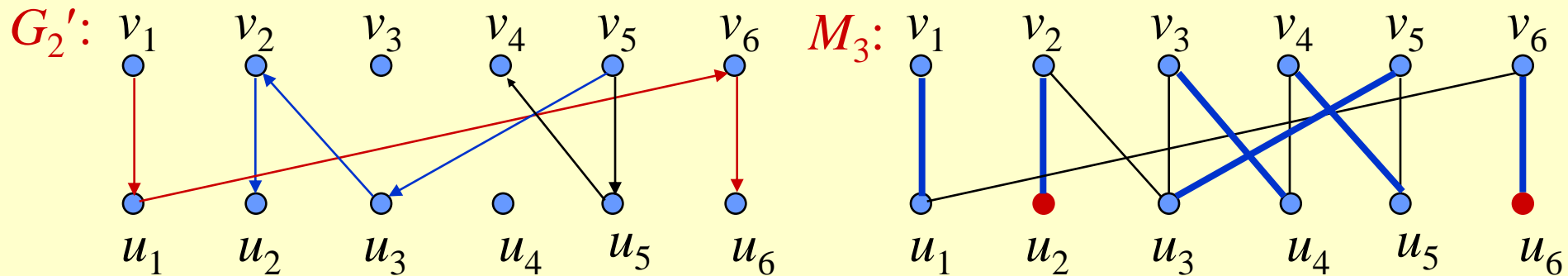


$P_1$

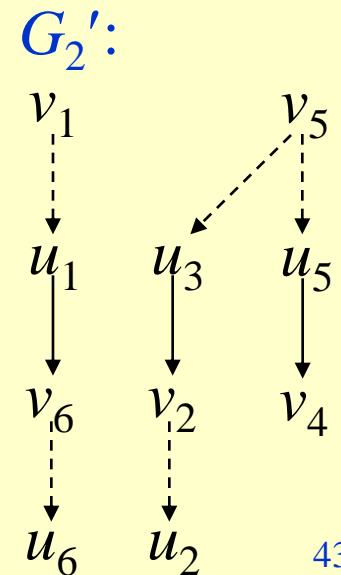
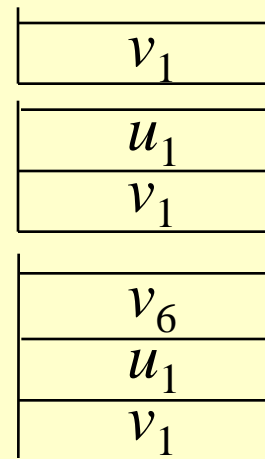
$P_2$

# Example Trace

In order to have a better understanding of the second step of Hopcroft-Karp algorithm, we trace the execution steps when applying it to the graph shown in Figure 8.

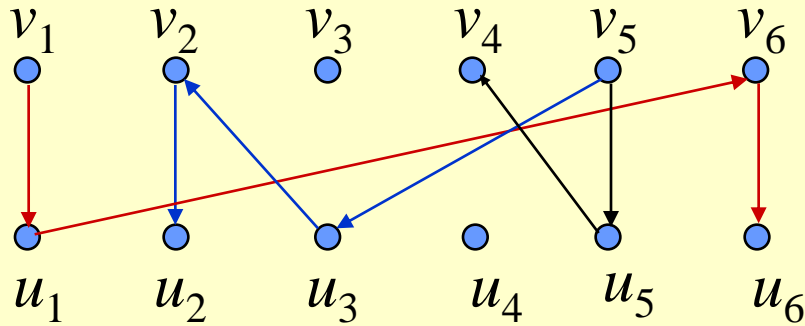


- Step 1:  $L_0 = \{v_1, v_5\}$   
 $push(v_1, stack)$ ; mark  $v_1$ ;
- Step 2:  $c-List(v_1) = \{u_1\}$   
 $push(u_1, stack)$ ; mark  $u_1$ ;
- Step 3:  $c-List(u_1) = \{v_6\}$   
 $push(v_6, stack)$ ; mark  $v_6$ ;



$$L_3 = \{u_2, u_4, u_6\}$$

$$L_0 = \{v_1, v_5\}$$



Step 4:  $c\text{-List}(v_6) = \{u_6\}$ ;  
 $push(u_6, stack)$ ; mark  $u_6$ ;

Step 5:  $c\text{-List}(u_6) = \phi$ ;  
 $u_6$  is in  $L_3$ ; /\*  $j^* = 3$ .\*/;

$u_6$  is a free node.

Output all the nodes in  $stack$ , which make up an augmenting path:

$$v_1 \rightarrow u_1 \rightarrow v_6 \rightarrow u_6;$$

empty stack;

$push(v_5, stack)$ ; mark  $v_5$ ;

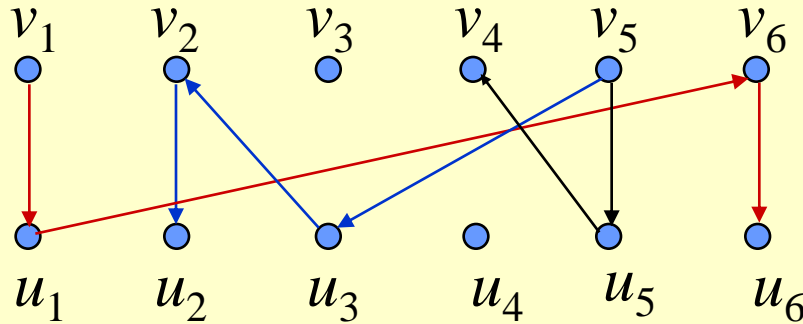
/\*  $v_5$  is the next element in  $L_0$ .\*/

$u_6$
$v_6$
$u_1$
$v_1$

$v_5$
-------

$$L_3 = \{u_2, u_4, u_6\}$$

$$L_0 = \{v_1, v_5\}$$



Step 6:  $c\text{-List}(v_5) = \{u_5, u_3\}$ ;  
 $push(u_5, stack)$ ; mark  $u_5$ ;

$u_5$
$v_5$

Step 7:  $c\text{-List}(u_5) = \{v_4\}$ ;  
 $push(v_4, stack)$ ; mark  $v_4$ ;

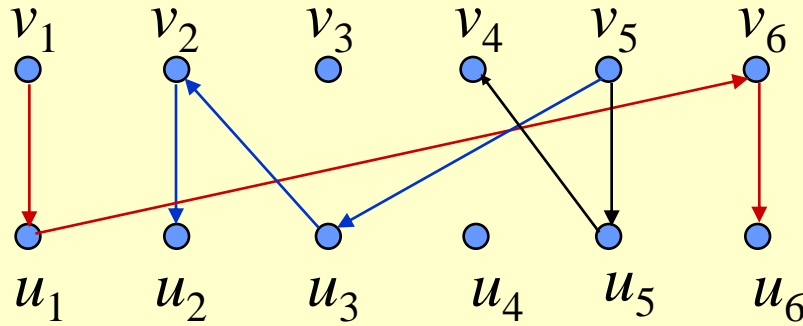
$v_4$
$u_5$
$v_5$

Step 8:  $c\text{-List}(v_4) = \phi$ ;  
 $v_4$  is neither in  $L_3$  nor in  $L_0$ ;  
 $/*j* = 3. */$   
 $pop(stack)$ ;

$u_5$
$v_5$

$$L_3 = \{u_2, u_4, u_6\}$$

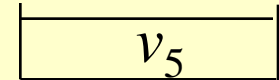
$$L_0 = \{v_1, v_5\}$$



Step 9:  $c\text{-List}(u_5) = \{v_4\}$ ;

$v_4$  is marked; remove  $v_4$  from the list;

$c\text{-List}(u_5) = \phi$ ;



Step 10:  $u_5$  is neither in  $L_3$  nor in  $L_0$ ;

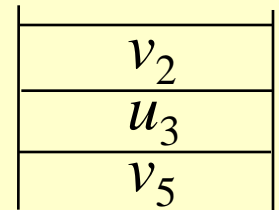
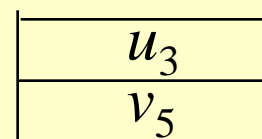
$\text{pop}(\text{stack})$ ;

$c\text{-List}(v_5) = \{u_3\}$ ; /\*  $u_5$  is removed from  $c\text{-List}(v_5)$ . \*/

Step 11:  $\text{push}(u_3, \text{stack})$ ; mark  $u_3$ ;

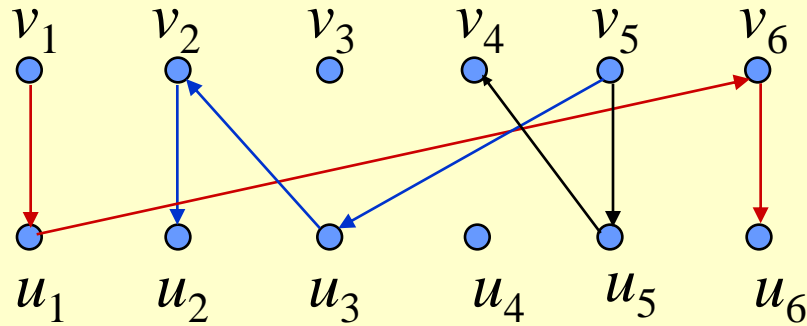
$c\text{-List}(u_3) = \{v_2\}$ ;

$\text{push}(v_2, \text{stack})$ ; mark  $v_2$ ;



$$L_3 = \{u_2, u_4, u_6\}$$

$$L_0 = \{v_1, v_5\}$$



Step 12:  $c\text{-List}(v_2) = \{u_2\}$ ;  
 $push(u_2, stack)$ ; mark  $u_2$ ;

Step 13:  $c\text{-List}(u_2) = \phi$ ;  
 $u_2$  is in  $L_3$ ; /\*  $j^* = 3$ . \*/;

$u_2$
$v_2$
$u_3$
$v_5$

Output all the nodes in  $stack$ , which make up an augmenting path:

$$v_5 \rightarrow u_3 \rightarrow v_2 \rightarrow u_2;$$

empty stack;

Now  $stack$  is empty and no element in  $L_0$  will be pushed into  $stack$ .

Stop.

# Bipartite Graph

## **Hopcroft-Karp algorithm (1973)**

In the above example, we choose the matching shown in Figure 5(b) as an initial matching for ease of explanation. In fact, we can choose any edge in the bipartite graph as an initial matching and then apply Hopcroft-Karp algorithm. Of course, the final matching found may be different from that shown in Figure 9.



# Project Requirement

1. Implementation of the algorithm in C++.
2. Documentation.