

# Hopcroft-Karp Algorithm

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1 Introduction

2 Definitions

3 Algorithm

4 Time Complexity

5 Applications

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# Introduction

## Hopcroft-Karp Algorithm

The Hopcroft-Karp algorithm is a graph algorithm that finds the maximum cardinality matching in a bipartite graph.

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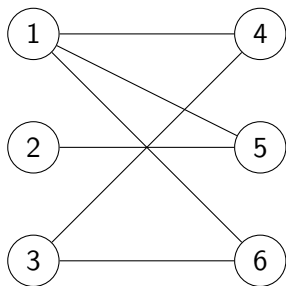


Figure: Bipartite graph

# Introduction

## Hopcroft-Karp Algorithm

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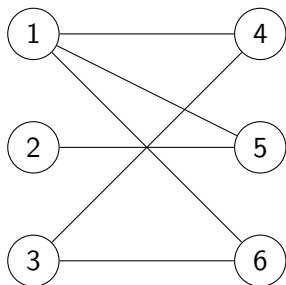


Figure: Bipartite graph

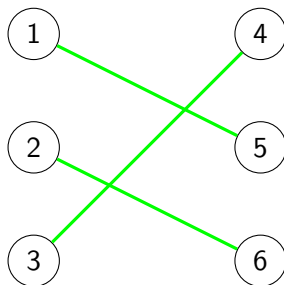


Figure: Maximum cardinality matching

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## Bipartite Graph

A graph is bipartite if its vertices can be divided into two disjoint sets such that every edge connects a vertex in one set to a vertex in the other set.



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A matching in a graph is a set of edges such that no two edges share a common vertex. Maximum cardinality means maximum possible matching.

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## Augmenting Path

- Starts and ends both at free vertex
- Edges in the path alternate between being in the matching and not in the matching

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## Hopcroft-Karp( $G$ )

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- 2: **repeat**
- 3:      $P = \{p_1, p_2, \dots, p_k\}$  //maximal set of vertex-disjoint shortest augmenting paths

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- 5: **until**  $P = \phi$



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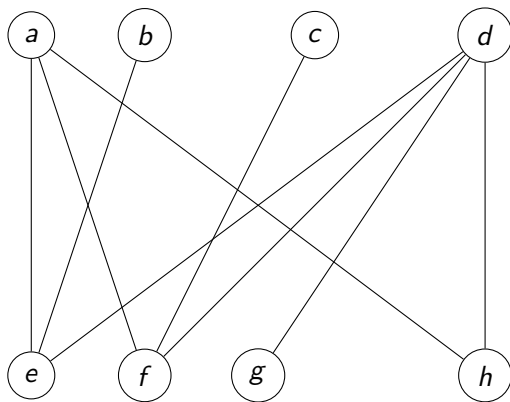
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  - 5: **until**  $P = \phi$
  - 6: Output the matching  $M$  as the maximum cardinality matching.
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# Example

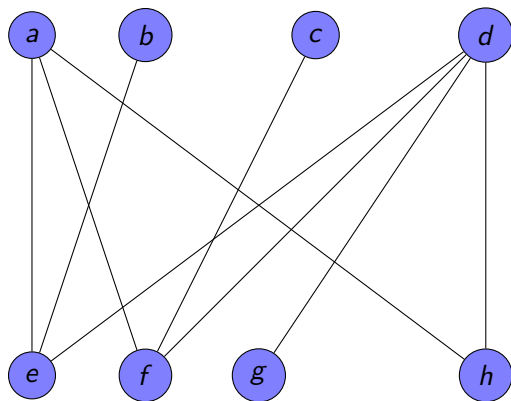
A bipartite graph is given below where we will find its maximum cardinality matching:



# First iteration: BFS

BFS sources =  $\{a, b, c, d\}$

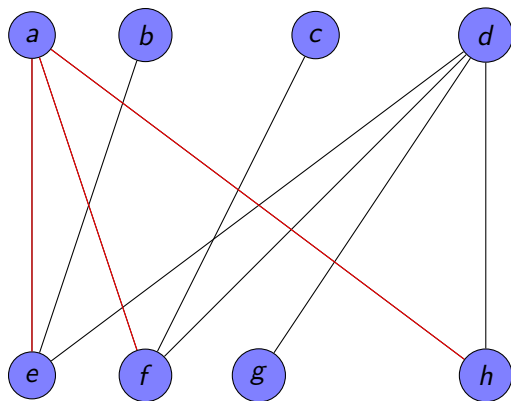
$F = \{\}$



# First iteration: BFS

BFS sources =  $\{a, b, c, d\}$

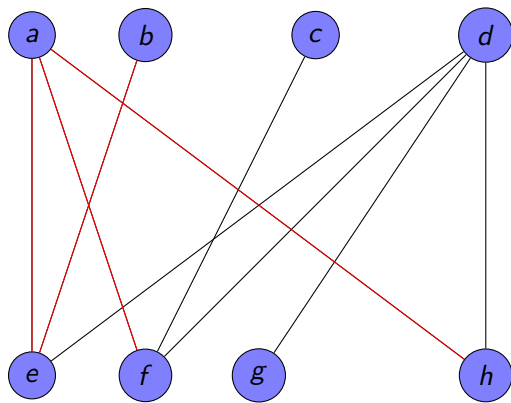
$F = \{e, f, h\}$



# First iteration: BFS

BFS sources =  $\{a, \textcolor{red}{b}, c, d\}$

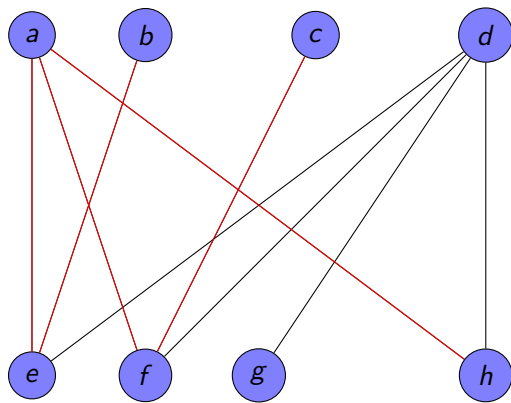
$F = \{e, f, h\}$



# First iteration: BFS

BFS sources =  $\{a, b, \textcolor{red}{c}, d\}$

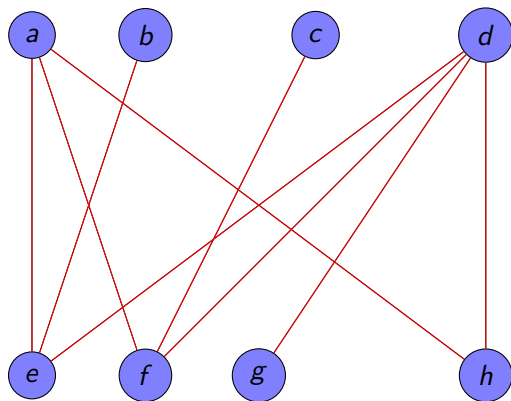
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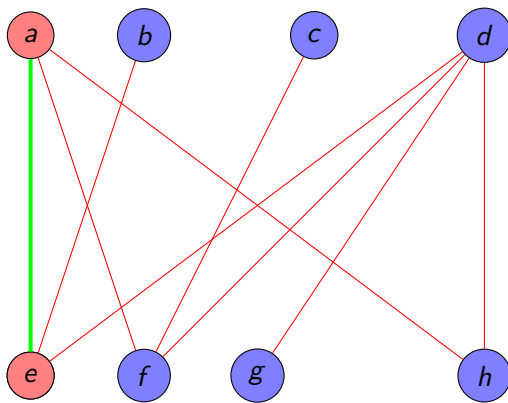
$F = \{e, f, h, g\}$



# First iteration:DFS

$$F = \{e, f, h, g\}$$

$$P = \{(e - a)\}$$

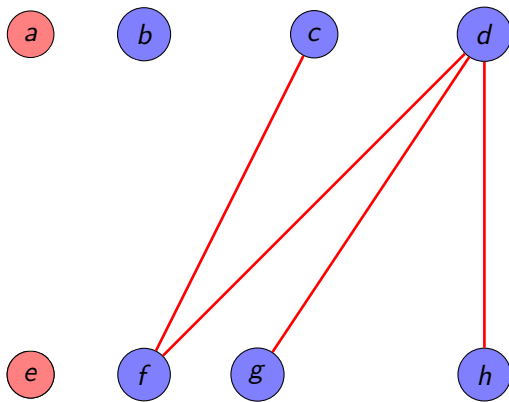




# First iteration:DFS

$$F = \{e, f, h, g\}$$

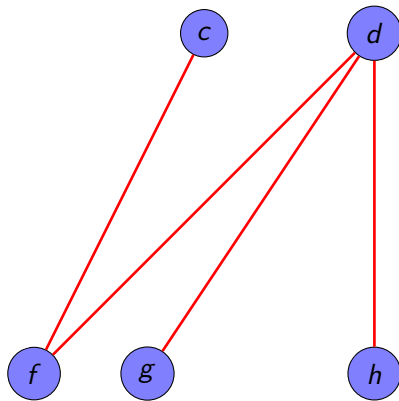
$$P = \{(e - a)\}$$



# First iteration:DFS

$$F = \{f, h, g\}$$

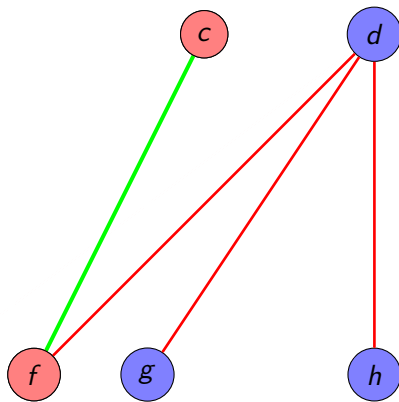
$$P = \{(e - a)\}$$



# First iteration:DFS

$$F = \{f, h, g\}$$

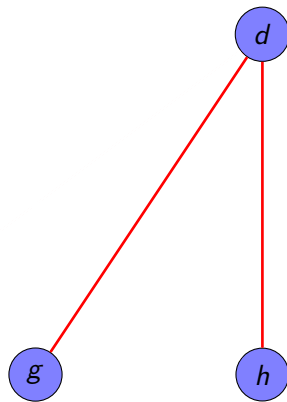
$$P = \{(e - a), (f - c)\}$$



# First iteration:DFS

$$F = \{h, g\}$$

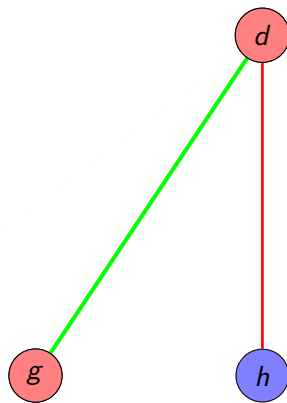
$$P = \{(e - a), (f - c)\}$$



# First iteration:DFS

$$F = \{h, \textcolor{red}{g}\}$$

$$P = \{(e - a), (f - c), (\textcolor{red}{g} - d)\}$$



# First iteration:DFS

$$F = \{h\}$$

$$P = \{(e - a), (f - c), (g - d)\}$$

$$\text{Before } M = \phi$$

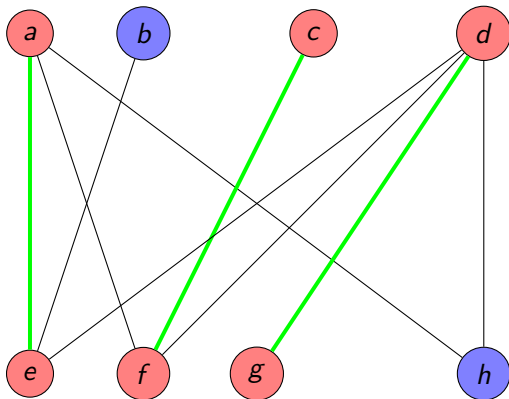
$$\text{After } M = M \oplus \{(e - a) \cup (f - c) \cup (g - d)\}$$

$$M = \{(e - a), (f - c), (g - d)\}$$

# After first iteration

After running the DFS,

$$M = \{(e - a), (f - c), (g - d)\}$$

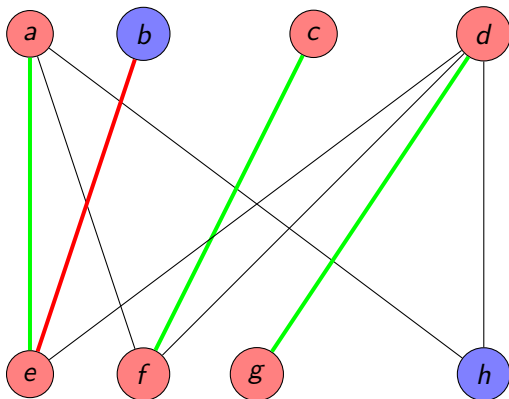


## Second iteration: BFS

BFS sources =  $\{b\}$

$F = \{\}$

BFS tree =  $b -> e$



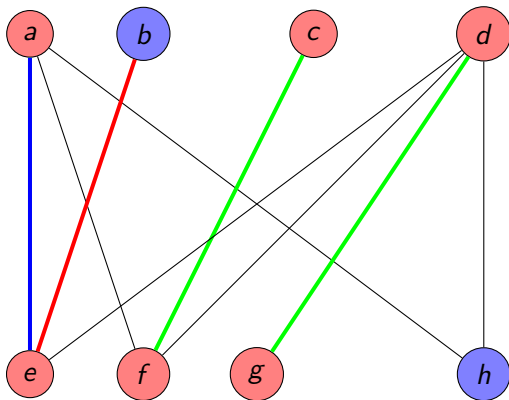


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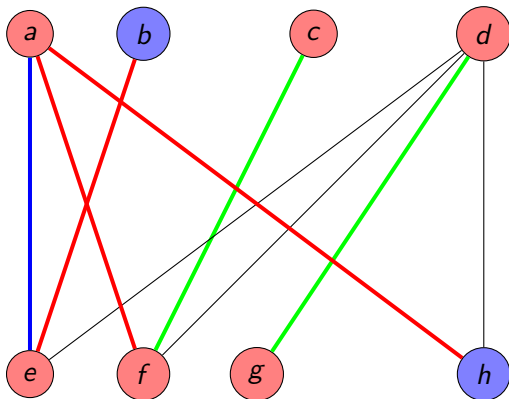


## Second iteration: BFS

BFS sources =  $\{b\}$

$F = \{h\}$

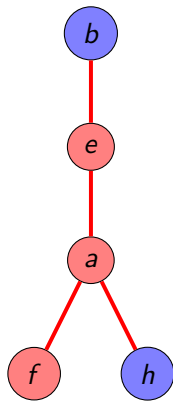
BFS tree =  $b \rightarrow e \rightarrow a \rightarrow f, h$



## Second iteration: DFS

DFS sources =  $\{h\}$

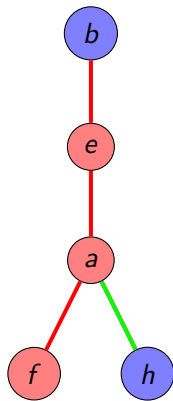
$P = \{\}$



## Second iteration: DFS

DFS sources =  $\{h\}$

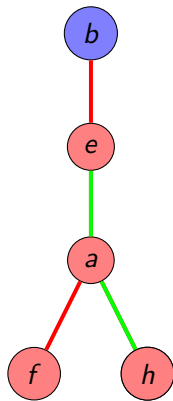
$P = \{(h - a -$



## Second iteration: DFS

DFS sources =  $\{h\}$

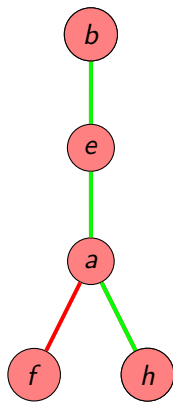
$P = \{(h - a - e - )\}$



## Second iteration:DFS

DFS sources =  $\{h\}$

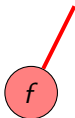
$P = \{(h - a - e - b)\}$



## Second iteration:DFS

DFS sources =  $\{\}$

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## Second iteration:DFS

$$P = \{(h - a - e - b)\}$$

$$\text{Before } M = \{(a - e), (f - c), (g - d)\}$$

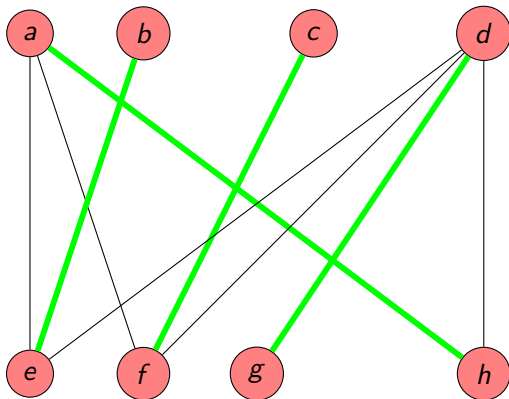
$$\text{After } M = M \oplus \{(h - a), (a - e), (e - b)\}$$

$$M = \{(h - a), (e - b), (f - c), (g - d)\}$$



# Algorithm termination

As no more free vertex is available in upper set, the algorithm terminates and finally we get the following graph with maximum cardinality 4 where  $M = \{(h - a), (e - b), (f - c), (g - d)\}$



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- 1:  $M = \phi$  //  $O(1)$
  - 2: **repeat**
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  - 4:      $M := M \oplus \{p_1 \cup p_2 \cup \dots \cup p_k\}$
  - 5: **until**  $P = \phi$
  - 6: Return  $M$  //  $O(1)$
-

## Per Loop Iteration

**Breadth First Search:** Uses all edges at most once so time complexity is  $O(|E|)$

**Depth First Search:** Since vertices and edges are deleted once used, all edges are also used at most once, so time complexity is  $O(|E|)$

**So over all each iteration is linear with the number of edges. i.e**  
 $O(|E|)$

# Complexity Analysis

- ▲ Breadth first search terminates when it reaches the free vertex. Therefore, there is no shorter path to a free vertex. Therefore, in subsequent iterations a shorter path cannot be found.
- ▲ Once it terminates it collects all free vertex on that level so all possible paths of that length are found. Therefore, in subsequent iterations the paths found must be longer.
- ▲ Since the paths alternate between matched and unmatched edges, and free vertices cannot be connected to a matched edge, then in subsequent iterations the paths must be at least two edges longer.

# Complexity Analysis

- ▲ After  $\sqrt{|V|}$  iterations the minimum path length would therefore be  $2\sqrt{|V|}$
- ▲ Since in  $P = \{p_1, p_2, \dots, p_k\}$  the paths are vertex disjoint and there are only  $|V|$  vertices in the graph, then there can only be  $\frac{|V|}{2\sqrt{|V|}} = \frac{1}{2}\sqrt{|V|}$
- ▲ Therefore after  $\sqrt{|V|}$  iterations only  $\frac{1}{2}\sqrt{|V|}$  more are needed so the loop will terminate after  $\frac{3}{2}\sqrt{|V|}$  repeats.

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# Applications

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- **Job scheduling** - matching workers with tasks based on their skills and availability.
- **Online advertising** - matching ads with potential viewers based on demographic and behavioral data.

# Thank You!

Gracias por todo



- Thank you for your attention.
- We appreciate your support.