

# **The Graph Data Structure**

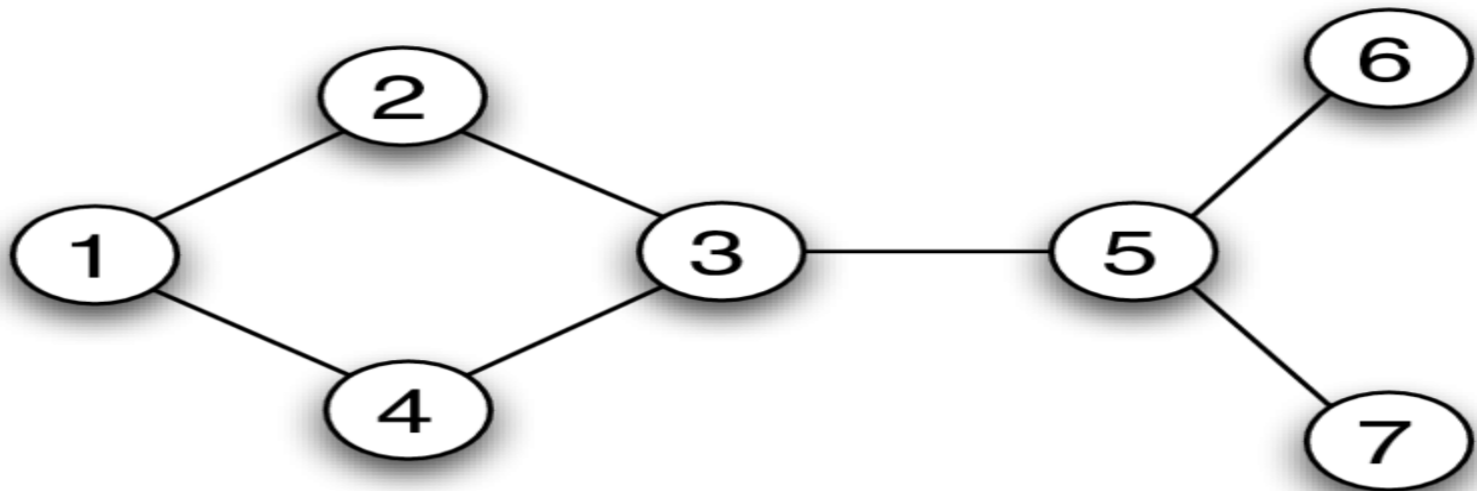
Dr. Amit Praseed

# Graphs

- A graph is a data structure that is commonly represented as  $G = (V, E)$ , where  $V$  is a set of vertices and  $E$  is a set of edges
- Graphs are commonly used to represent a large number of real world problems
  - Railways, roadways, airline routes, transmission towers etc.
  - Routing traffic over the Internet
  - Representing game outcomes
  - Representing a problem search space

# Graph Terminology

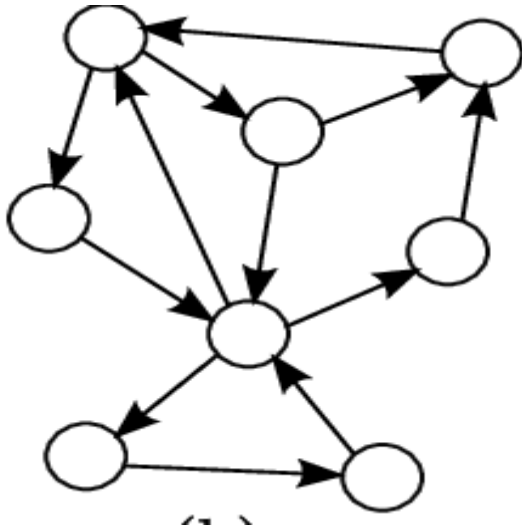
- A graph is a collection of vertices,  $V$  and a collection of edges,  $E$
- Every edge  $e \in E$  can be represented as  $\{u, v\}$  where  $u, v \in V$



# Graph Terminology

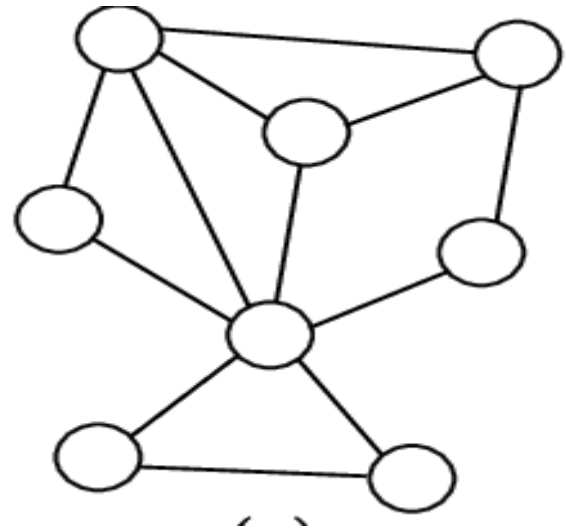
## Directed Graphs

- A graph in which the edges are represented as **ordered pairs**  $(u, v)$  are called directed graphs
- Edges have direction



## Undirected Graphs

- A graph in which the edges are represented as **unordered pairs**  $(u, v)$  are called undirected graphs
- Edges do not have direction



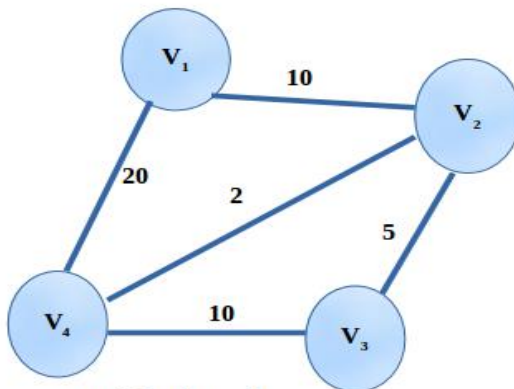
# Graph Terminology

- Graphs without loops and parallel edges are often called simple graphs; non-simple graphs are sometimes called multigraphs
- For any edge  $u-v$  in an undirected graph, we call  $u$  a neighbor of  $v$  and vice versa, and we say that  $u$  and  $v$  are adjacent.
- The degree of a node is its number of neighbors.
- For any directed edge  $u-v$ , we call  $u$  a predecessor of  $v$ , and we call  $v$  a successor of  $u$ . The in-degree of a vertex is its number of predecessors; the out-degree is its number of successors

# Graph Terminology

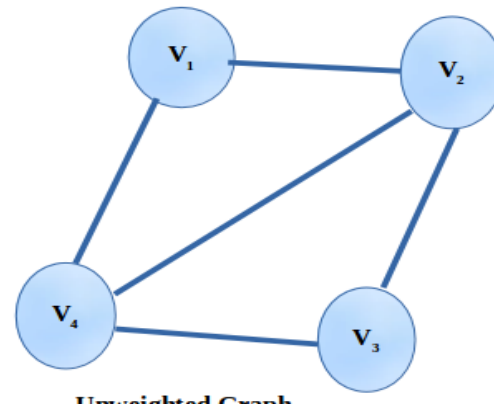
## Weighted Graphs

- A graph in which the edges are assigned a numeric value (called weight) are called weighted graphs
- Can represent path length, delay etc.



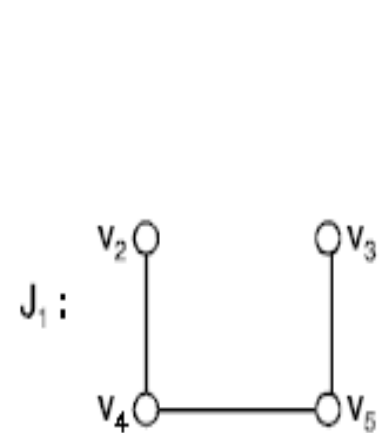
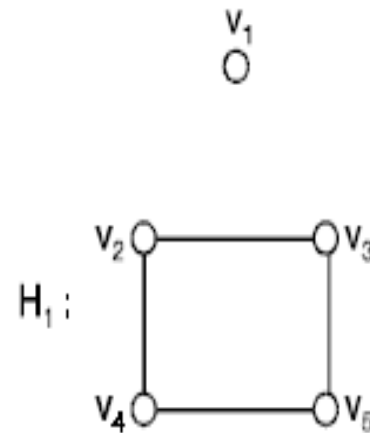
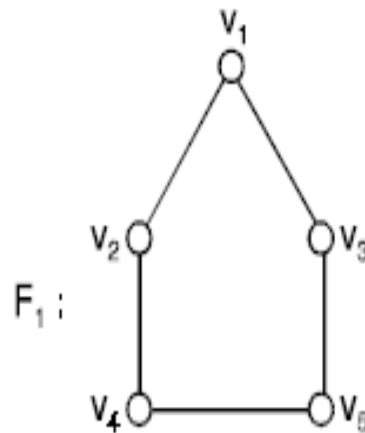
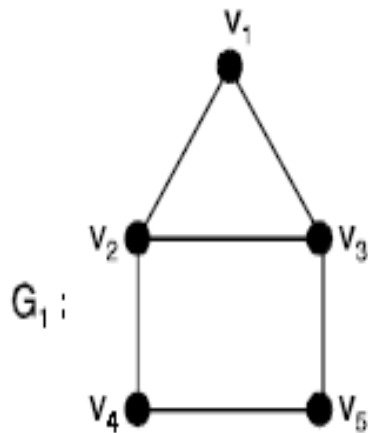
## Undirected Graphs

- A graph in which the edges do not have any numeric value associated with them are called unweighted graphs



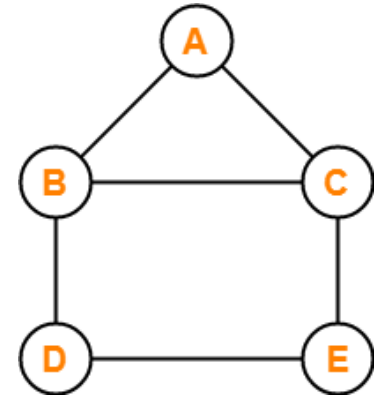
# Graph Terminology

- A graph  $G' = (V', E')$  is a subgraph of  $G = (V, E)$  if  $V' \subseteq V$  and  $E' \subseteq E$ 
  - By definition,  $G$  is a subgraph of itself
- A proper subgraph of  $G$  is any subgraph other than  $G$  itself.



# Walks, Trails, Paths

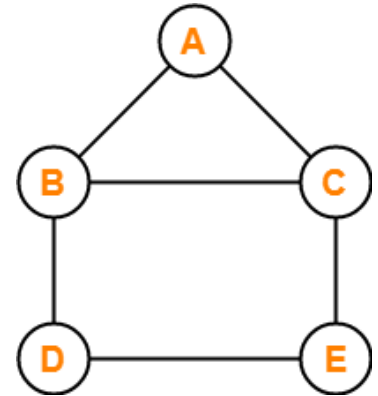
- A walk is a sequence of vertices, where each adjacent pair of vertices are adjacent in  $G$ 
  - A vertex can be traversed more than once
  - An edge can be used more than once
- Eg:  $d\ b\ a\ c\ e\ d\ e\ c$  is a walk of length 7
- If a walk starts and ends at the same vertex it is called a closed walk
  - Otherwise it is called an open walk





# Walks, Trails, Paths

- A path is a walk in which each vertex is visited at most once
  - A vertex cannot be traversed twice
  - An edge can be used more than once
- Eg: a b c e d is a path of length 4
- A trail is a walk in which an edge can be traversed at most once
  - A vertex can be visited more than once
  - An edge can only be visited once

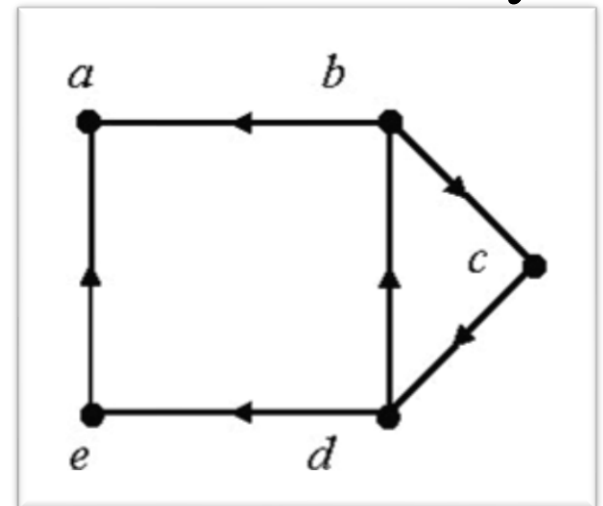


# Connected Graphs

- Two vertices  $u$  and  $v$  in a graph  $G$ ,  $v$  is said to be reachable from  $u$  if there exists a path between  $u$  and  $v$ .
- An undirected graph is connected if every vertex is reachable from every other vertex.
- Every undirected graph consists of one or more components, which are its maximal connected subgraphs; two vertices are in the same component if and only if there is a path between them

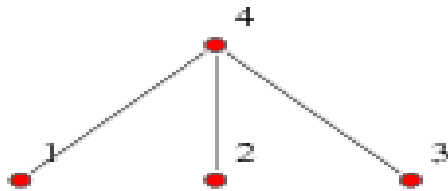
# Connected Graphs

- A directed graph  $G$  is said to be strongly connected if there exists a path between any two vertices  $u$  and  $v$ 
  - Note that the path needs to be a directed path
- The given graph is not strongly connected (why?)
- A directed graph  $G$  is said to be weakly connected if the graph is not strongly connected, but the underlying undirected graph is connected

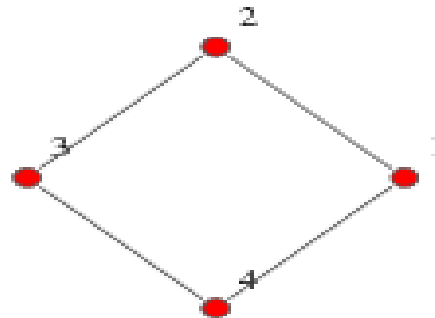


# Adjacency Matrix

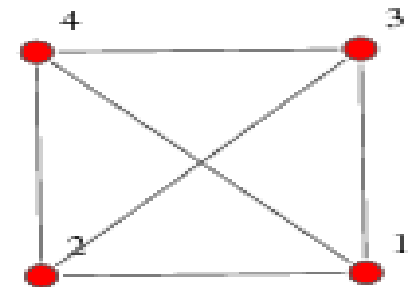
- An adjacency matrix is a  $|V| \times |V|$  matrix
  - $A[i, j] = 1$  if there is an edge from vertex  $u$  to vertex  $v$
  - Otherwise,  $A[i, j] = 0$



$$\begin{pmatrix} 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 \\ 1 & 1 & 1 & 0 \end{pmatrix}$$



$$\begin{pmatrix} 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \\ 0 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \end{pmatrix}$$



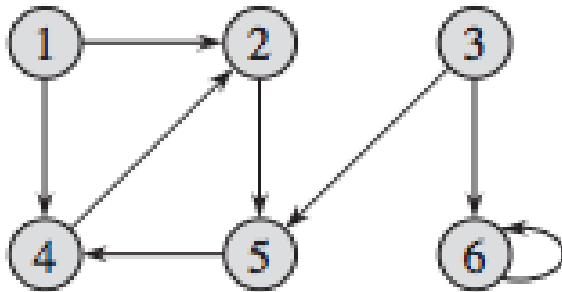
$$\begin{pmatrix} 0 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 1 \\ 1 & 1 & 1 & 0 \end{pmatrix}$$

# Adjacency Matrix

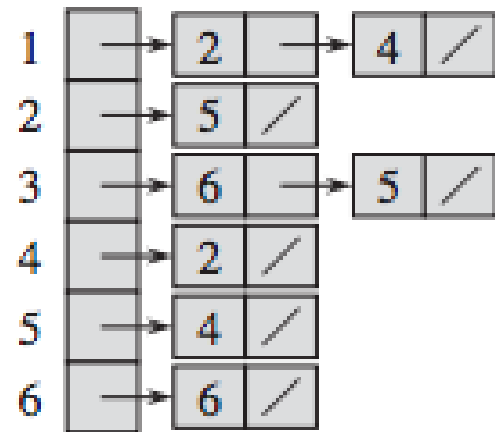
- For undirected graphs, the adjacency matrix is always symmetric, meaning  $A[u, v] = A[v, u]$  for all vertices  $u$  and  $v$  and the diagonal entries  $A[u, u]$  are all zeros.
- For directed graphs, the adjacency matrix may or may not be symmetric, and the diagonal entries may or may not be zero.
- Given an adjacency matrix
  - We can decide in  $\theta(1)$  time whether two vertices are connected by an edge
  - We can also list all the neighbors of a vertex in  $\theta(V)$  time.
  - Adjacency Matrices require  $\theta(V^2)$  space, regardless of how many edges the graph actually has
- Wastage of space and time for sparse graphs

# Adjacency List

- An adjacency list is an array of lists, each containing the neighbors of one of the vertices
  - For undirected graphs, each edge  $u-v$  is stored twice; for directed graphs, each edge  $u-v$  is stored only once



(a)



(b)

# Adjacency List

- Given an adjacency list
  - We can list the neighbors of a node  $v$  in  $O(1 + \deg(v))$  time
  - We can determine whether  $u-v$  is an edge in  $O(1 + \deg(u))$  time
  - We can traverse the graph in  $O(V + E)$  time
  - Adjacency lists require a space complexity of  $O(V + E)$
- Adjacency lists are usually preferred for representing sparse graphs, but dense graphs can be more efficiently represented using adjacency matrices

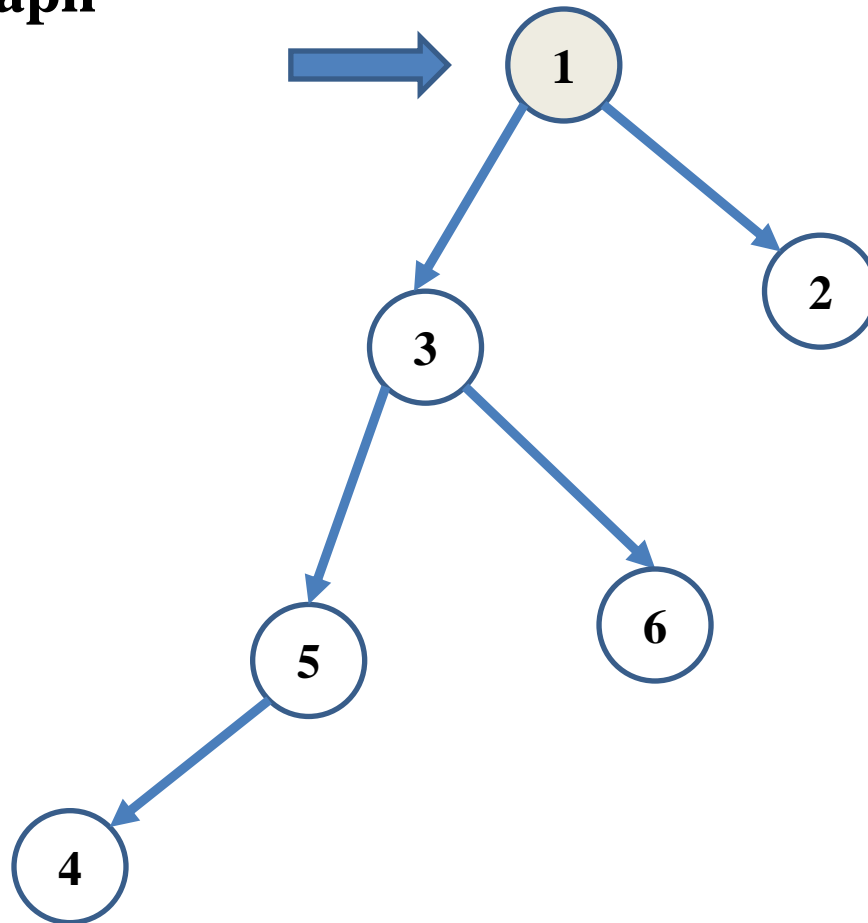
# Depth First Search

- Depth First Search (DFS) is a graph traversal algorithm
- As the name suggests, DFS explores one path in a graph completely before exploring a new one
- When DFS hits a dead end on a path, it backtracks and starts exploring a new path from the previous node
- This behaviour is suggestive of a LIFO data structure – STACK
- DFS can be implemented easily with recursion which uses an implicit stack



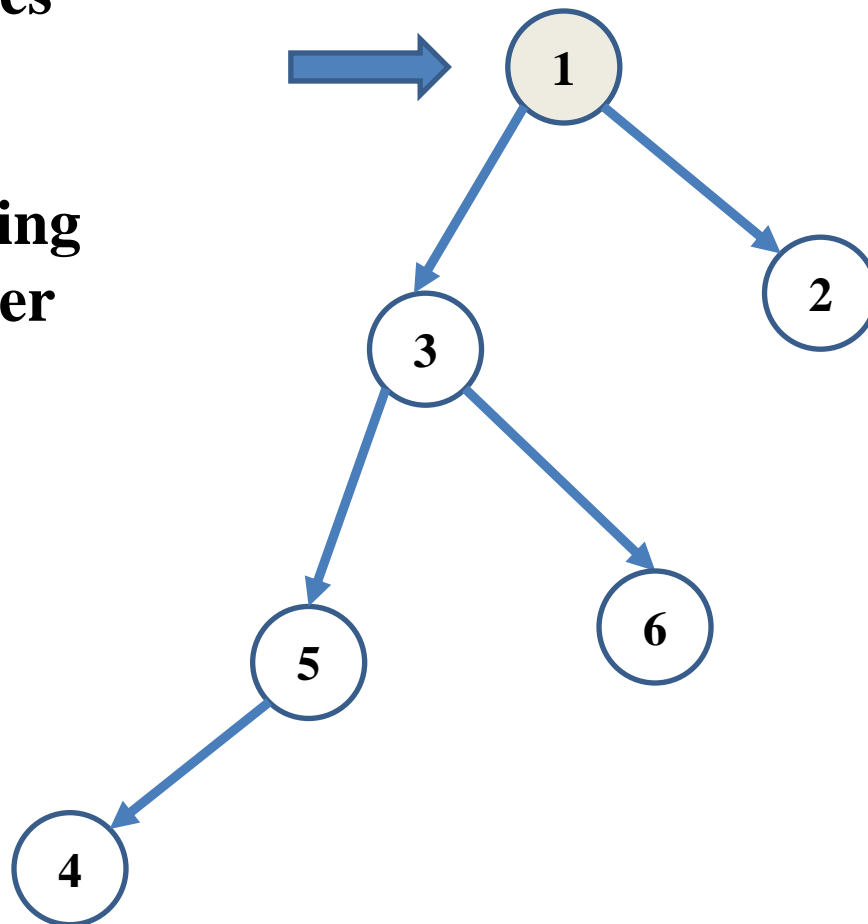
# Depth First Search

**DFS is a graph traversal algorithm**



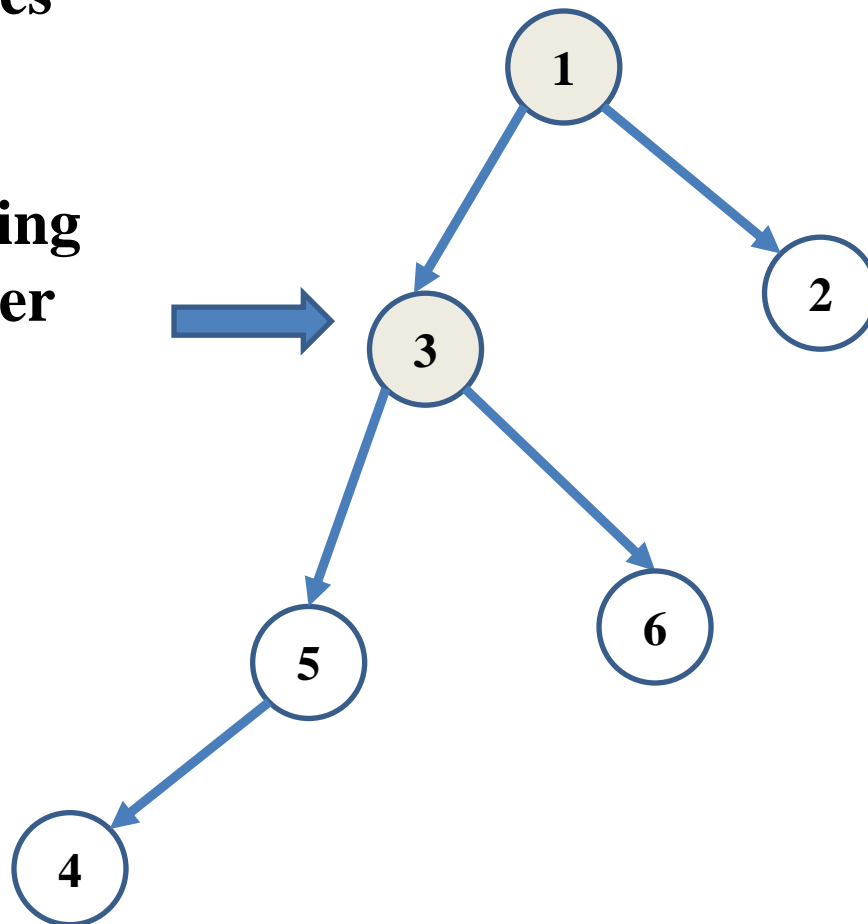
# Depth First Search

**DFS explores  
one path  
completely  
before moving  
on to another**



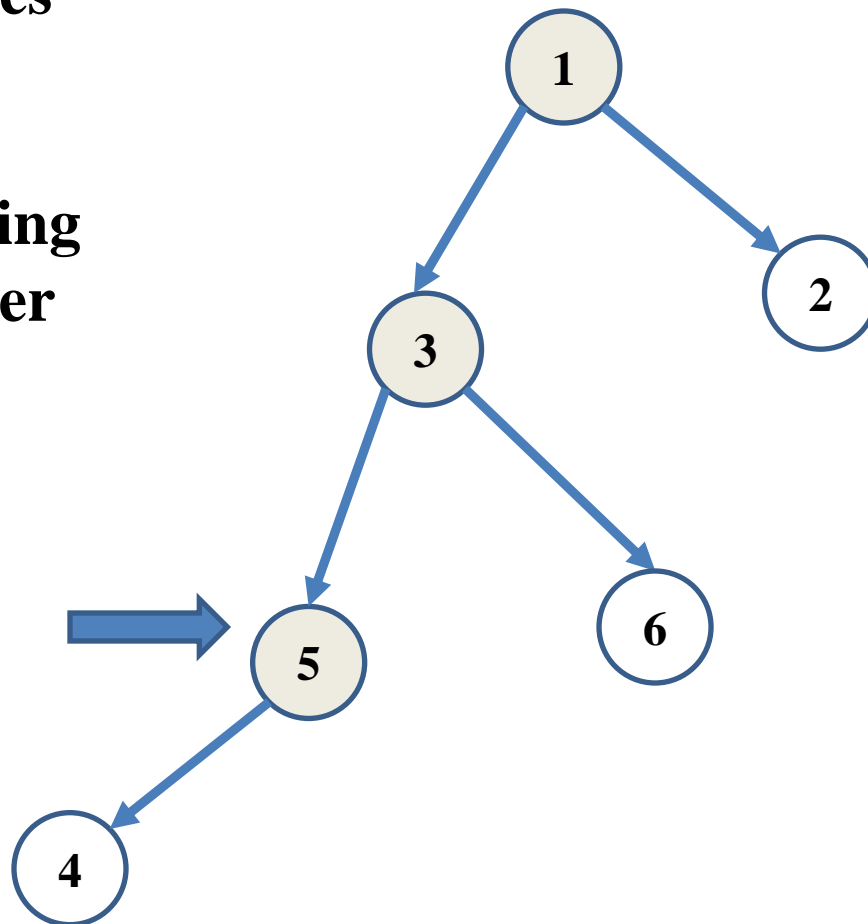
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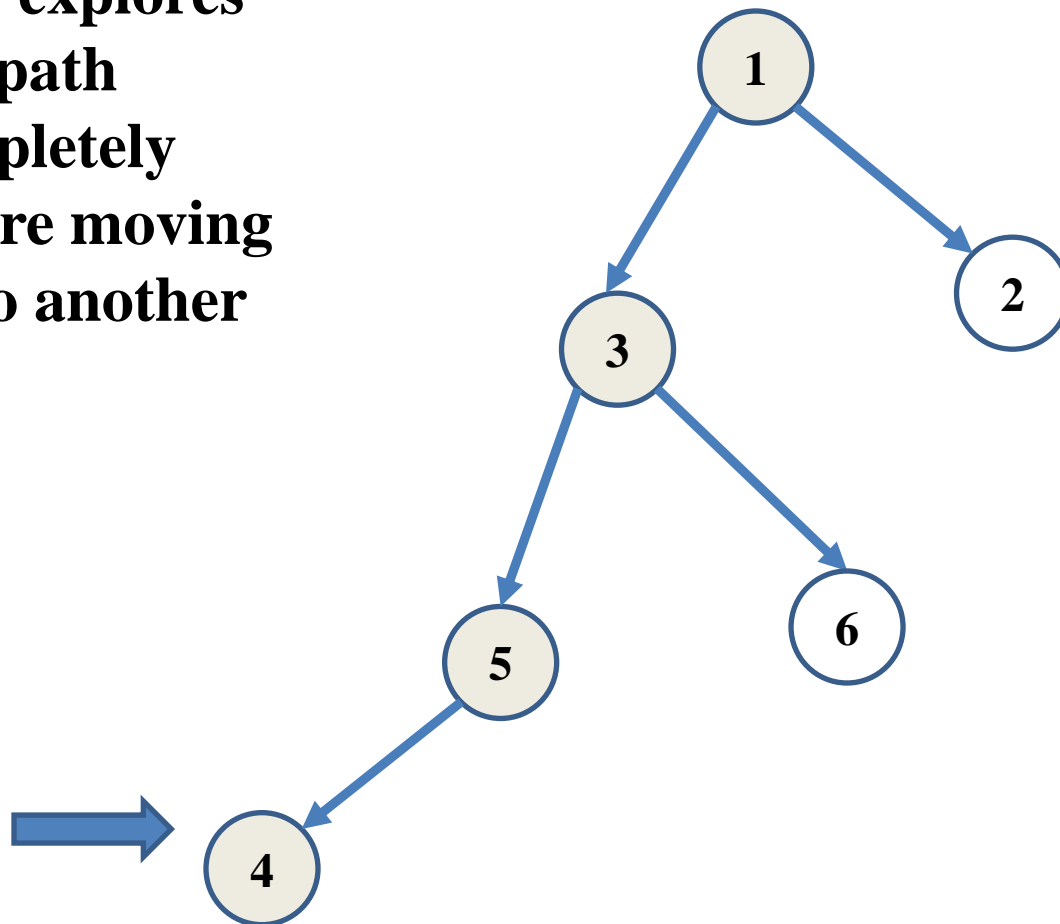
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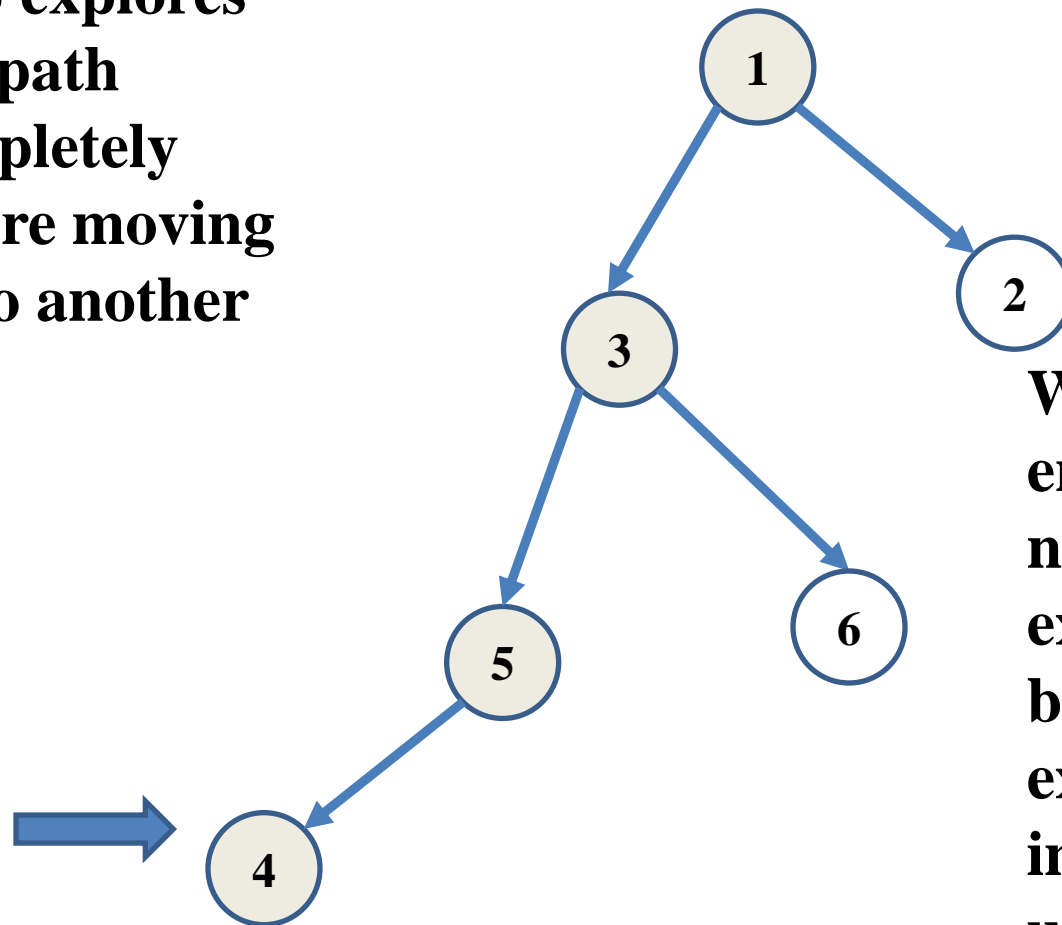
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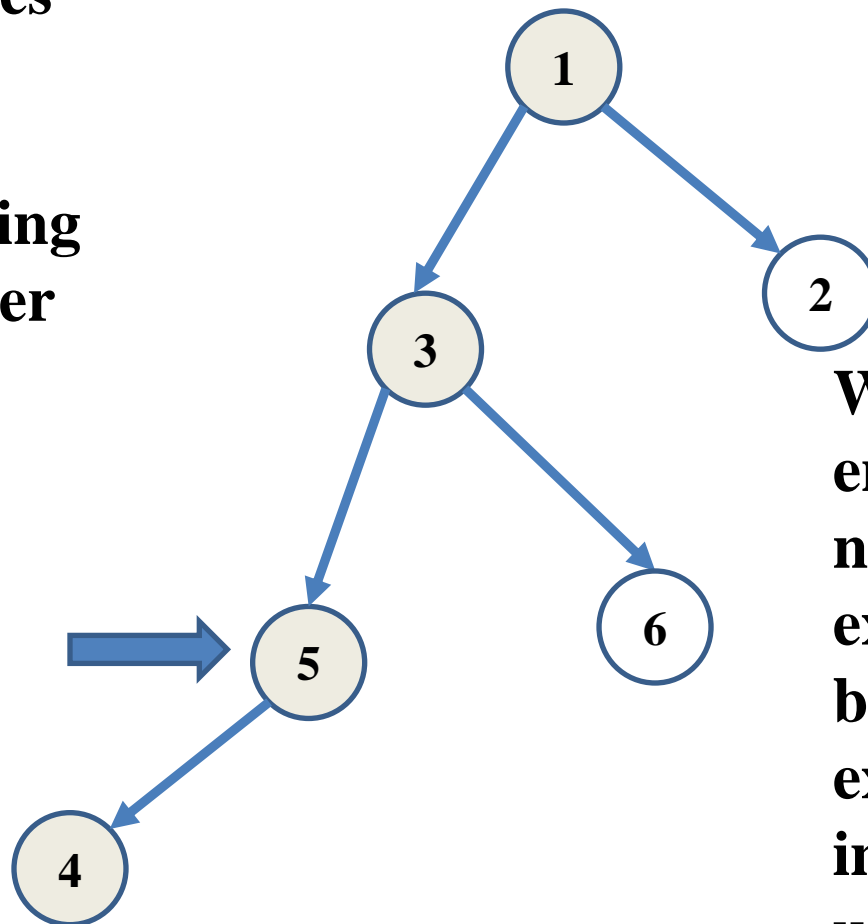


**DEAD END**

**When DFS hits a dead  
end (no further  
neighbours to be  
explored), it  
backtracks and  
explores the  
immediately previous  
unexplored path**

# Depth First Search

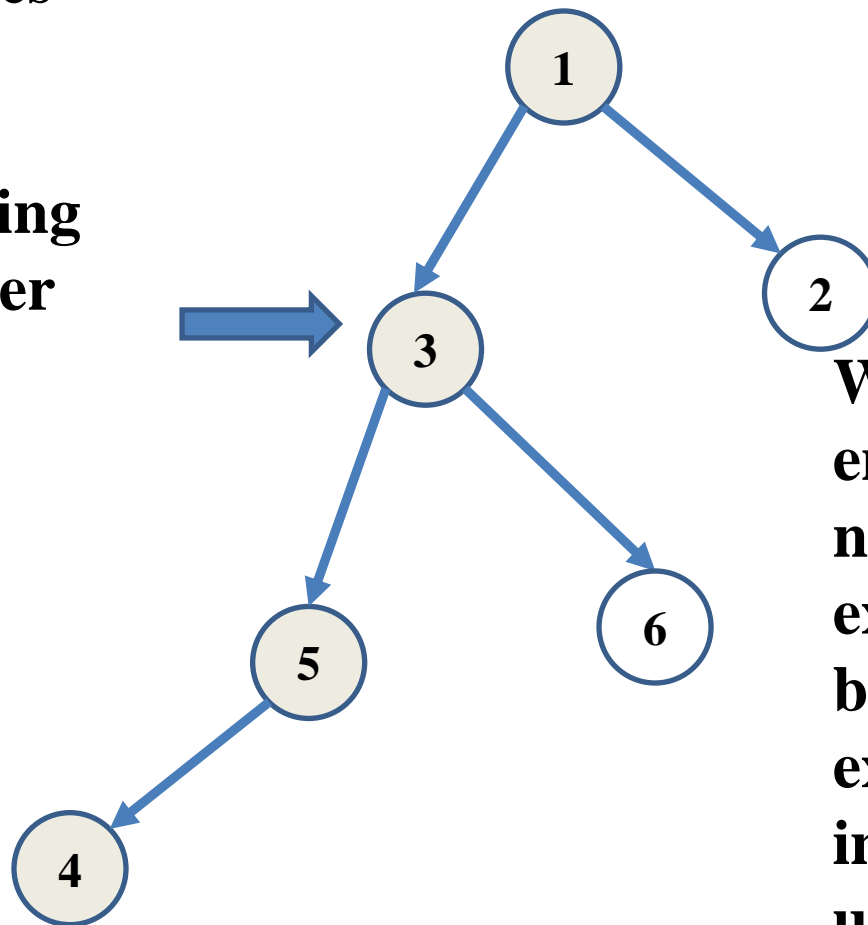
**DFS explores one path completely before moving on to another**



**When DFS hits a dead end (no further neighbours to be explored), it backtracks and explores the immediately previous unexplored path**

# Depth First Search

**DFS explores one path completely before moving on to another**

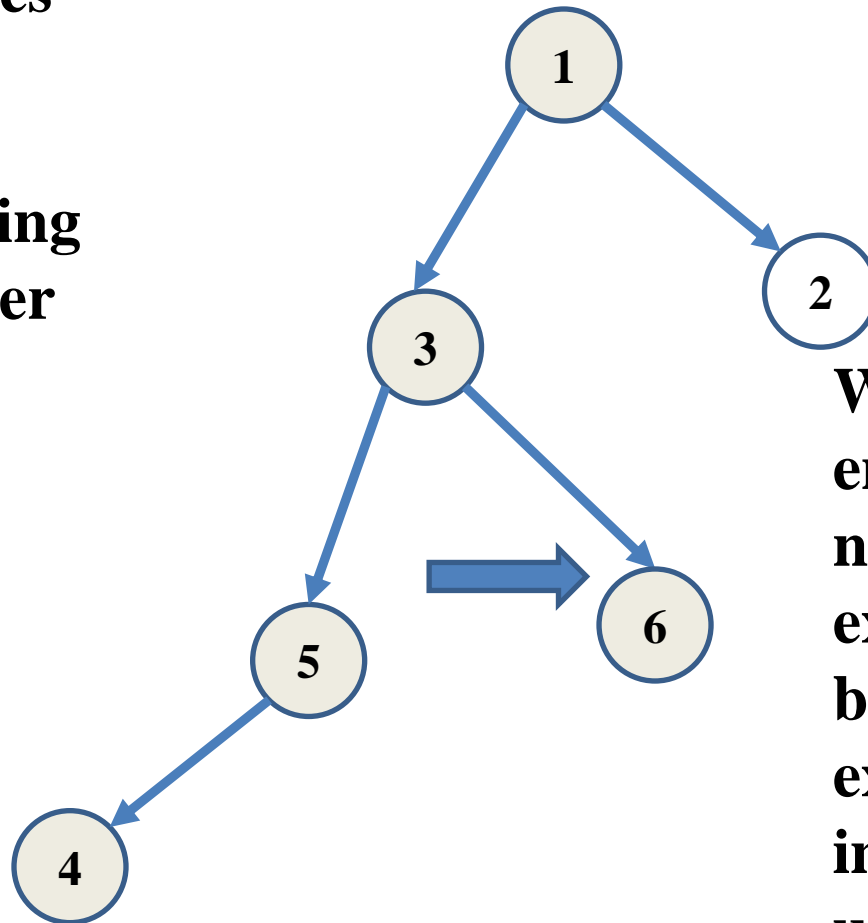


**When DFS hits a dead end (no further neighbours to be explored), it backtracks and explores the immediately previous unexplored path**



# Depth First Search

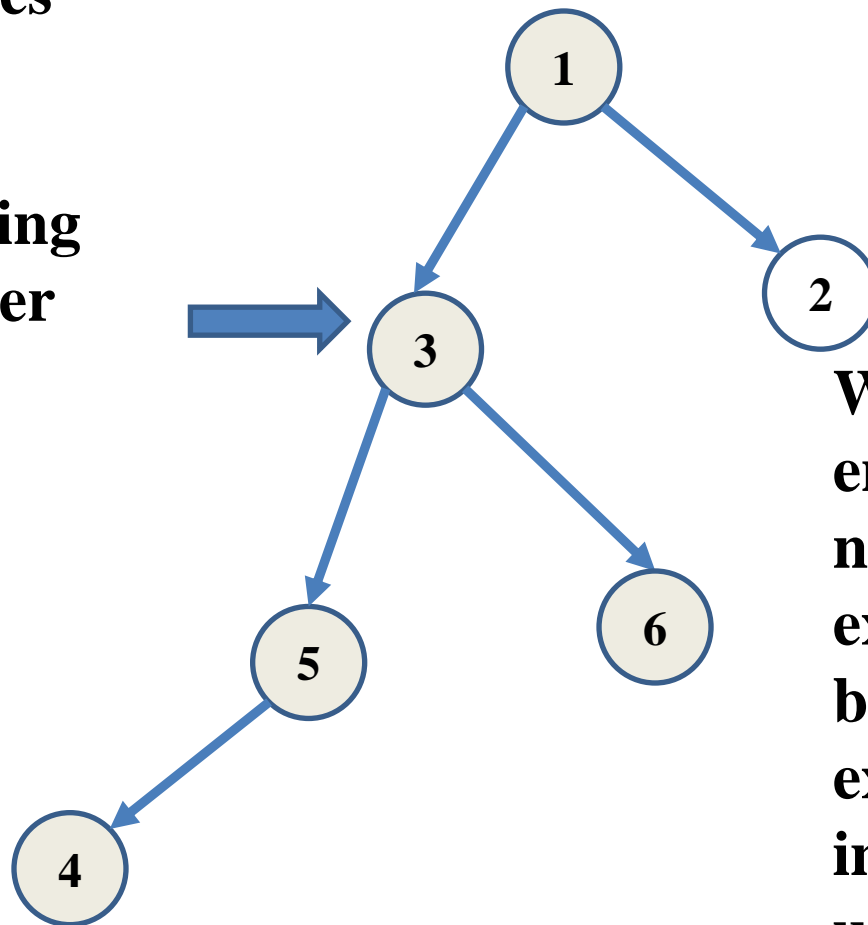
**DFS explores one path completely before moving on to another**



**When DFS hits a dead end (no further neighbours to be explored), it backtracks and explores the immediately previous unexplored path**

# Depth First Search

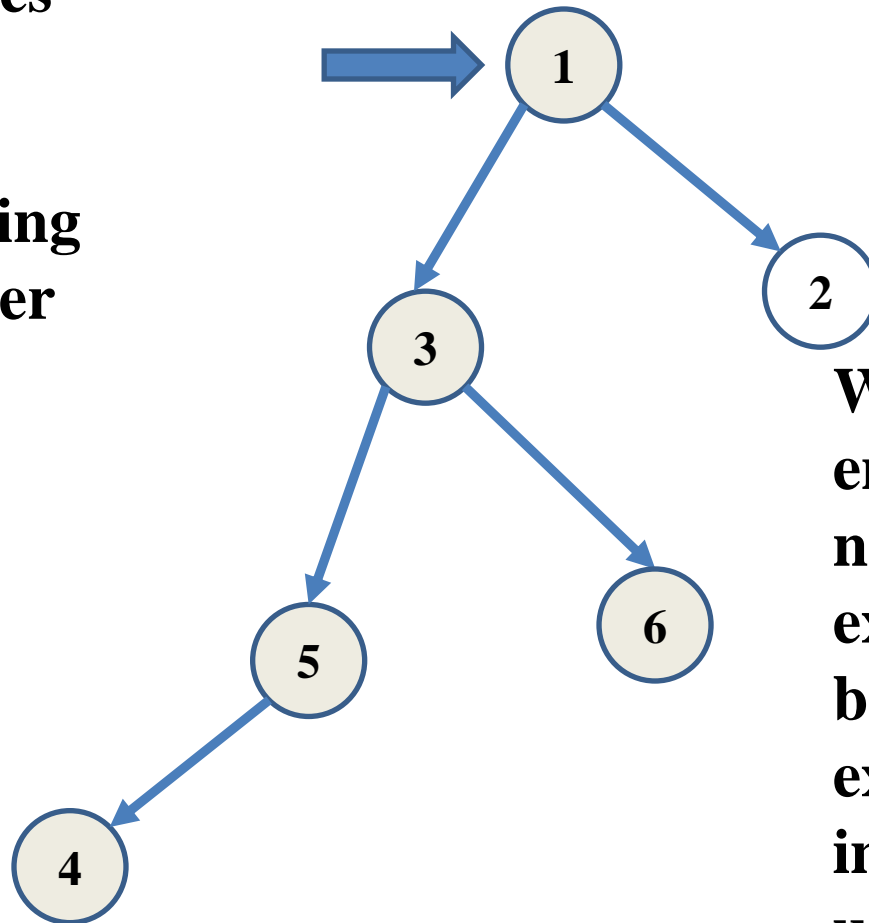
**DFS explores one path completely before moving on to another**



**When DFS hits a dead end (no further neighbours to be explored), it backtracks and explores the immediately previous unexplored path**

# Depth First Search

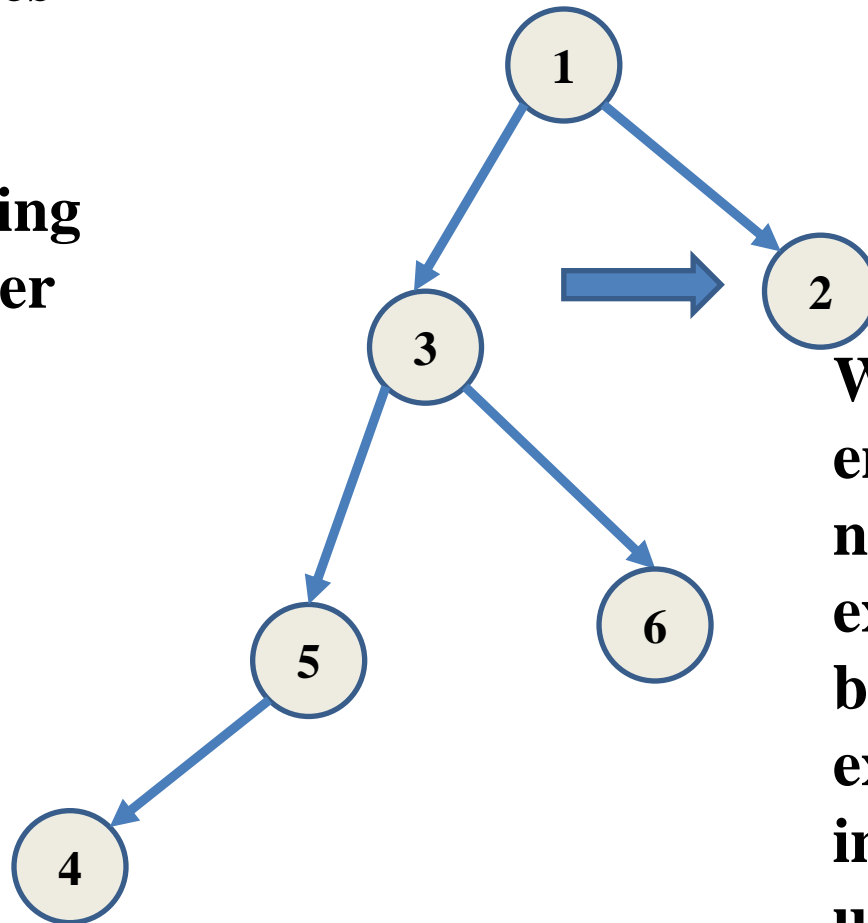
**DFS explores one path completely before moving on to another**



**When DFS hits a dead end (no further neighbours to be explored), it backtracks and explores the immediately previous unexplored path**

# Depth First Search

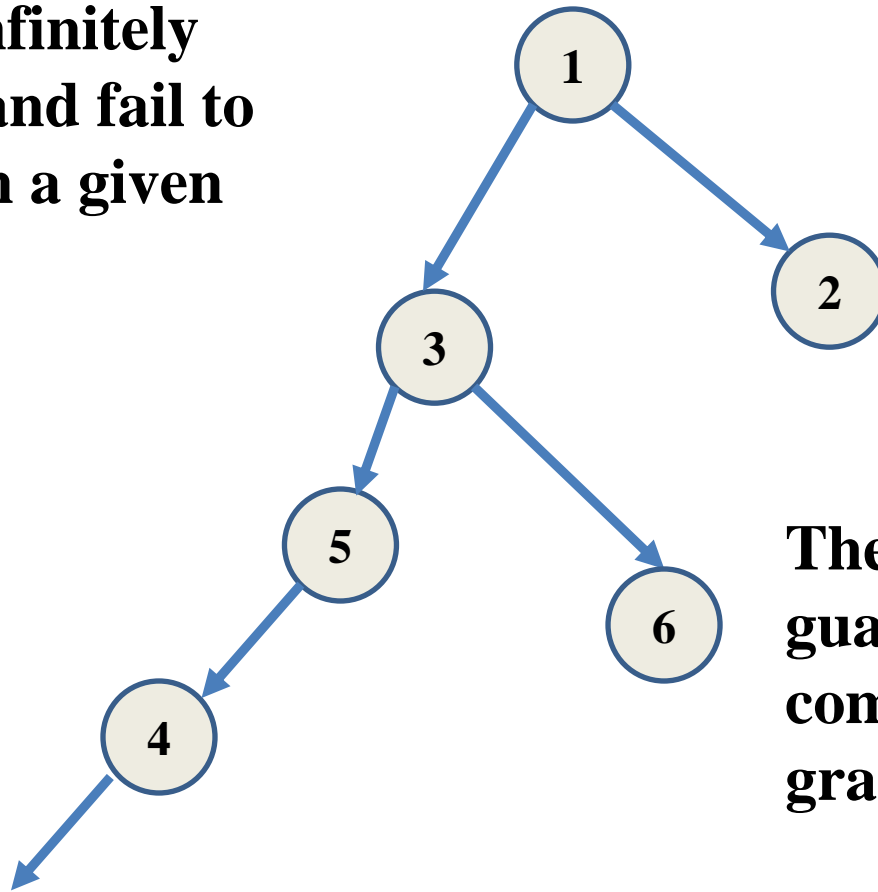
**DFS explores one path completely before moving on to another**



**When DFS hits a dead end (no further neighbours to be explored), it backtracks and explores the immediately previous unexplored path**

# Completeness of DFS

**DFS could get stuck  
exploring infinitely  
long paths and fail to  
terminate in a given  
time**

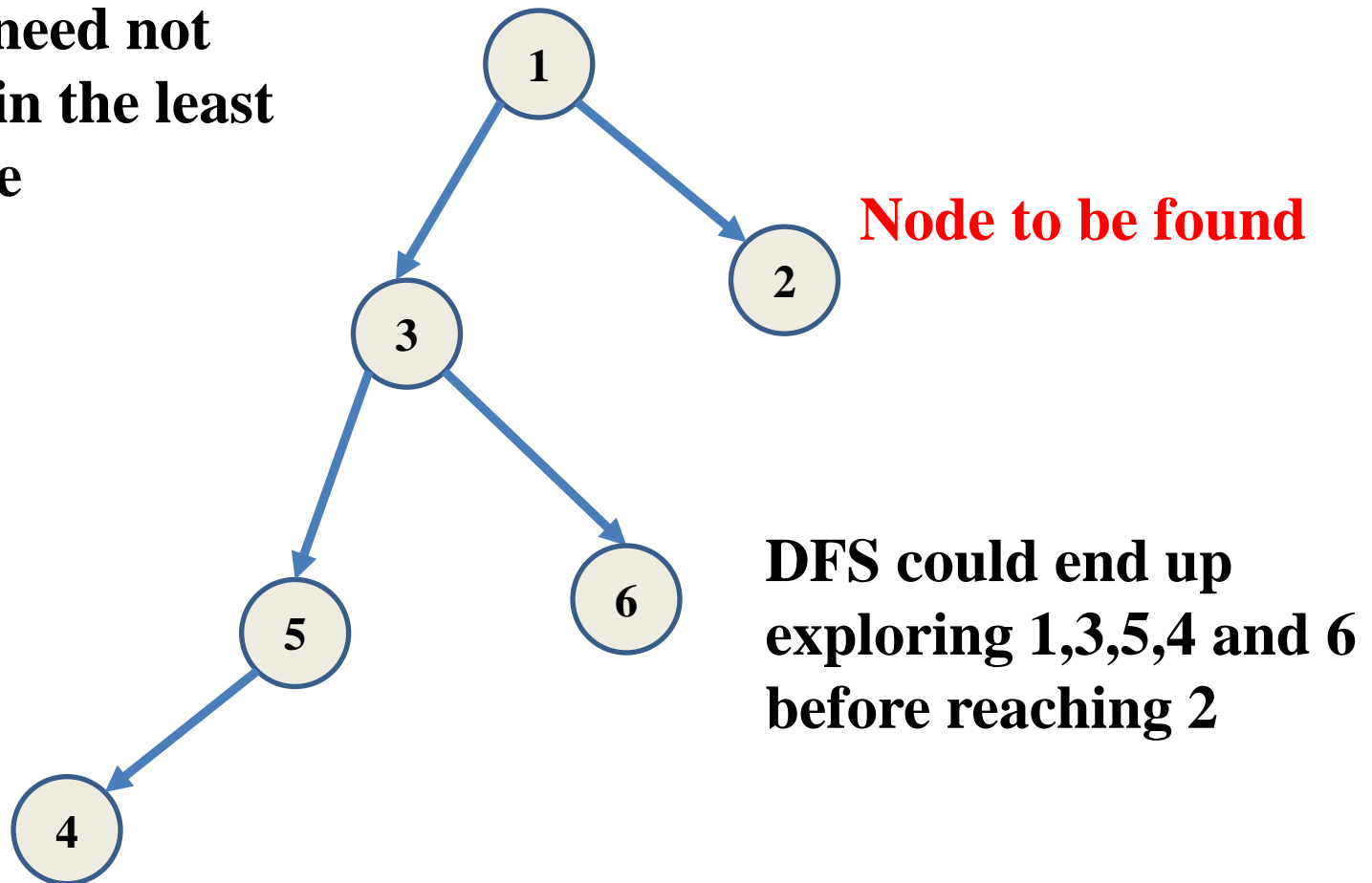


**The DFS algorithm is  
guaranteed to be  
complete for finite  
graphs**

**Potentially continues to  $\infty$**

# Optimality of DFS

**DFS algorithm is not optimal - it need not find a node in the least possible time**



# Applications of DFS

- Cycle detection
- Path Detection
- Finding strongly connected components
- Job Scheduling with dependencies (Topological Sort)

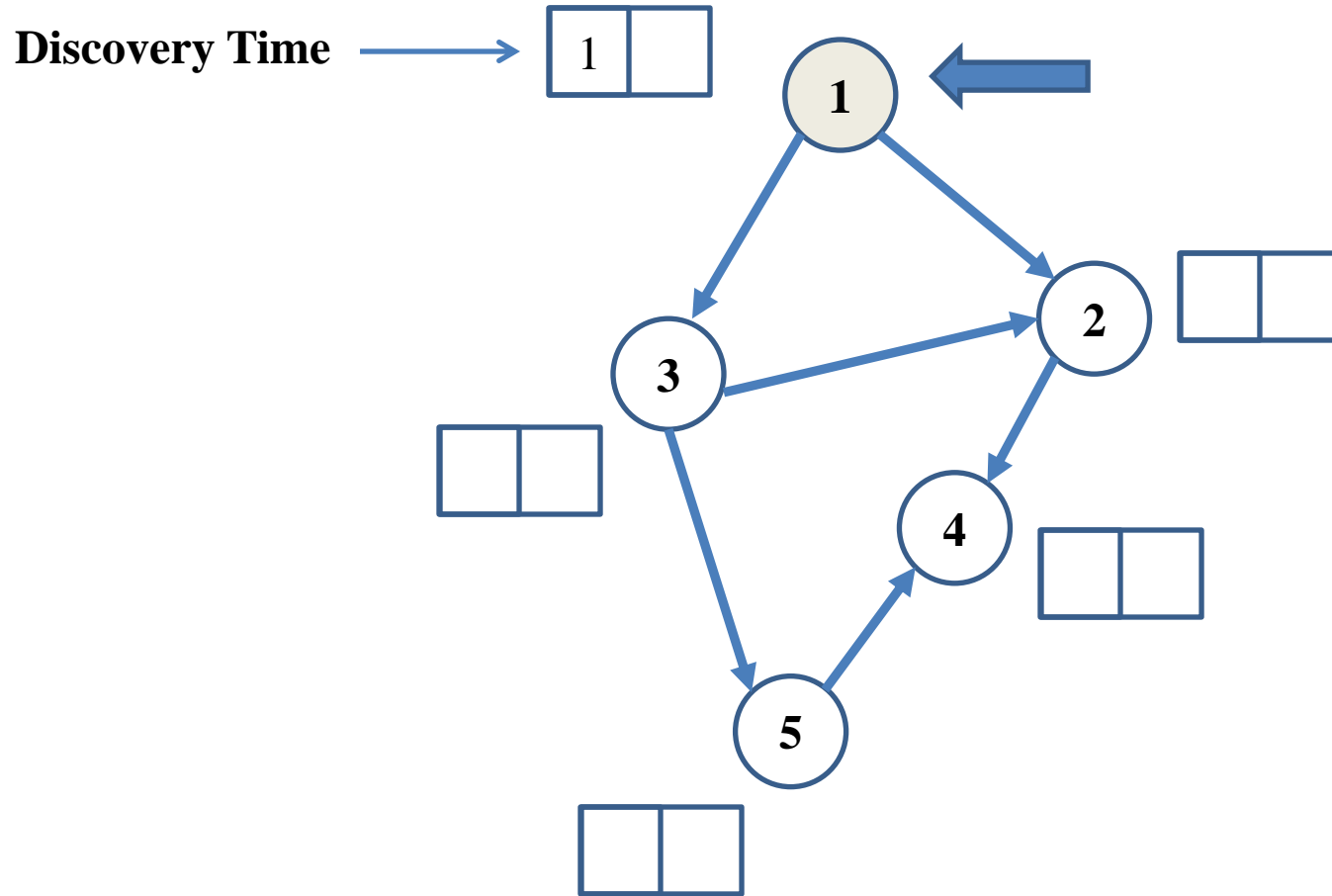
# Associated Notations

- **Nodes are assigned colours**
  - WHITE : The node has not been visited
  - GRAY : The node has been visited, but all of its branches have not been visited completely
  - BLACK : A node and its branches have been explored completely
- Often, two other values are associated with a node
  - **Discovery Time (d)** : The time at which the node becomes gray
  - **Finishing Time (f)** : The time at which the node becomes black
- **Predecessor Subgraph** : Each time a node is visited, its parent is noted to construct the **DFS tree / forest**



# DFS in Action

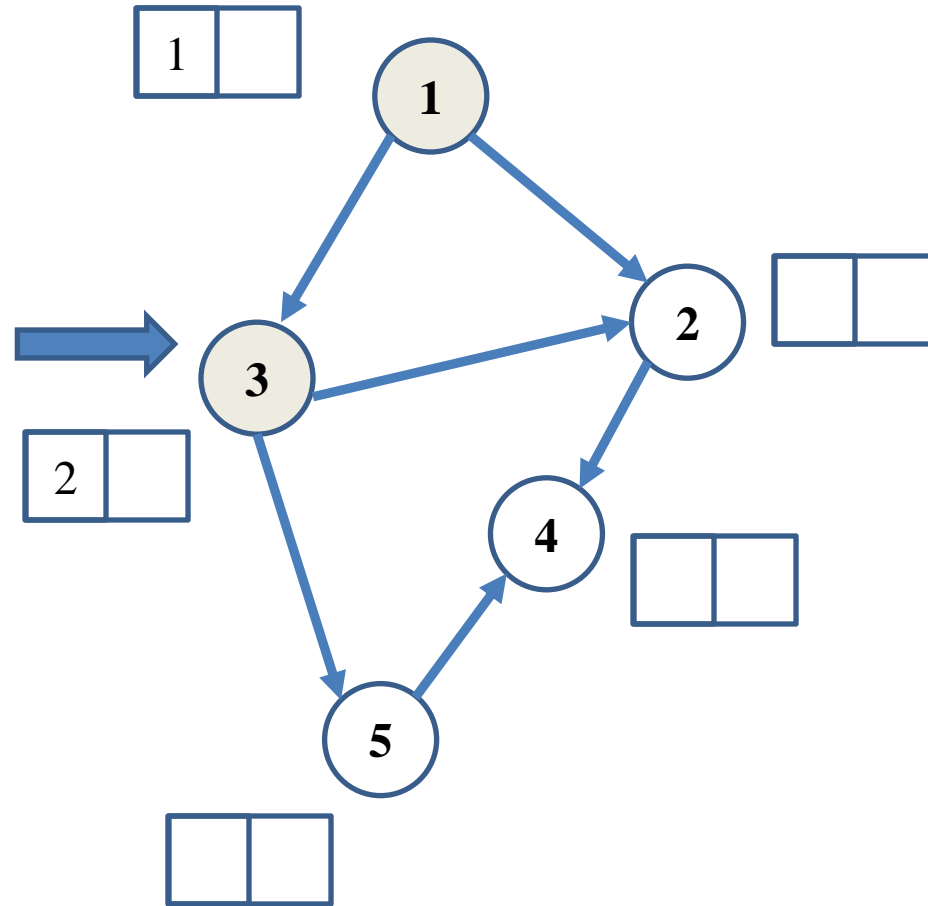
**Time : 1**



# DFS in Action

**Time : 2**

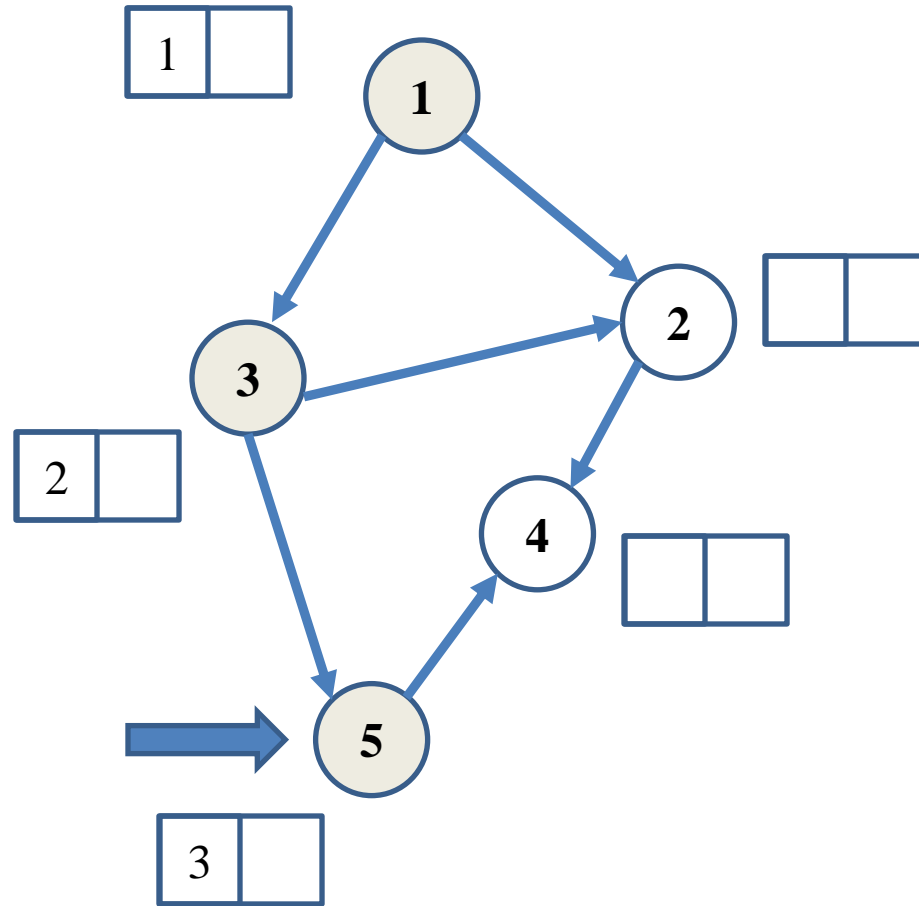
**$\Pi(3) = 1$**



# DFS in Action

**Time : 3**

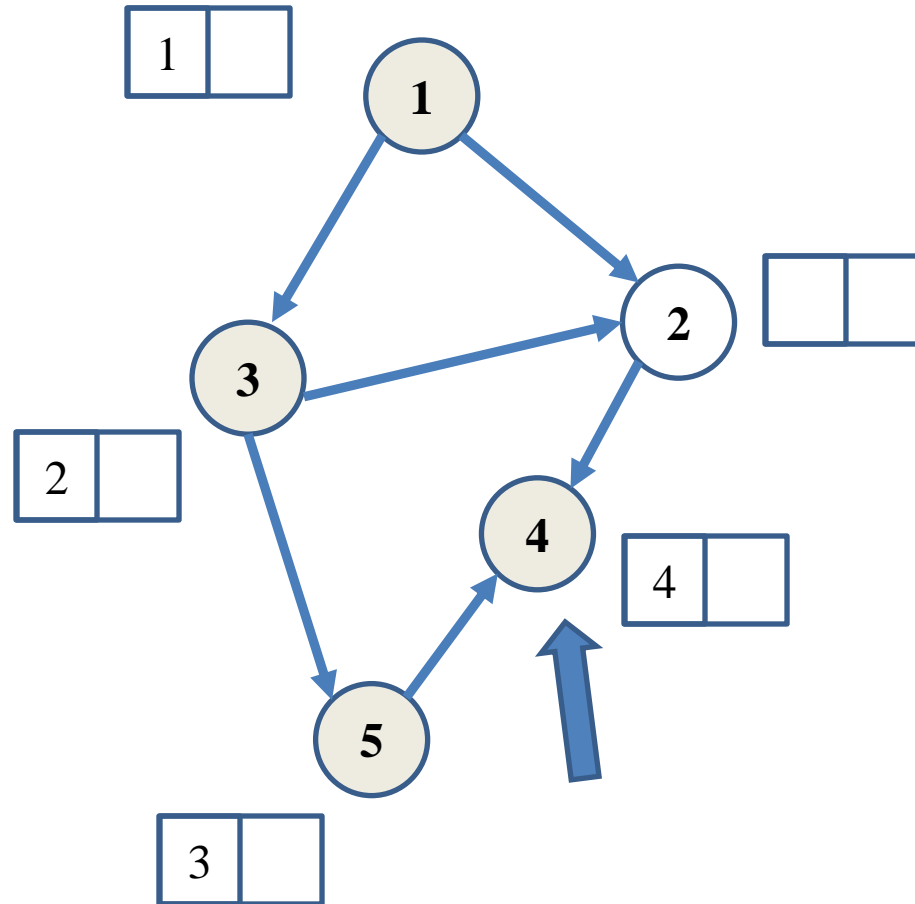
$\Pi(3) = 1$   
 $\Pi(5) = 3$



# DFS in Action

Time : 4

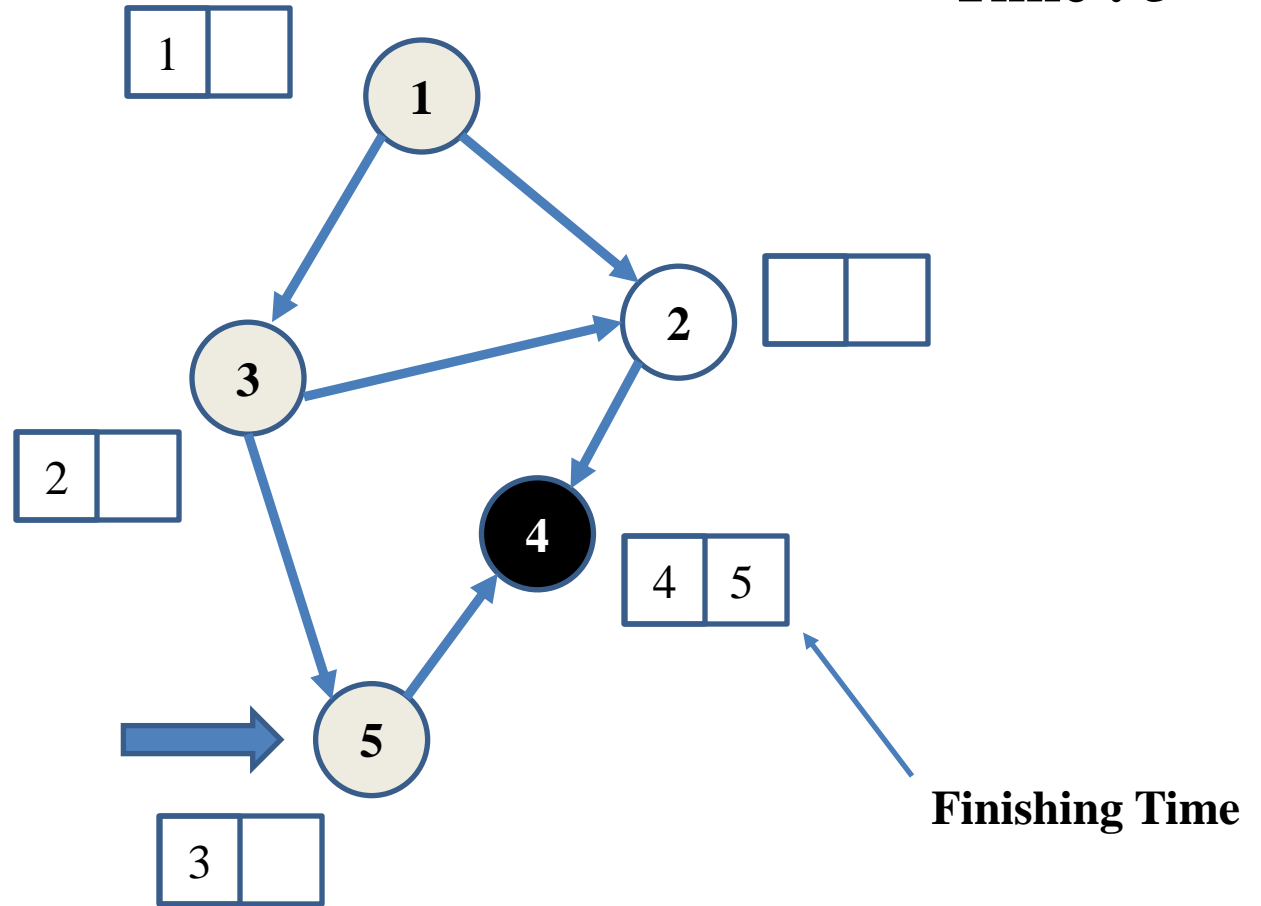
$\Pi(3) = 1$   
 $\Pi(5) = 3$   
 $\Pi(4) = 5$



# DFS in Action

**Time : 5**

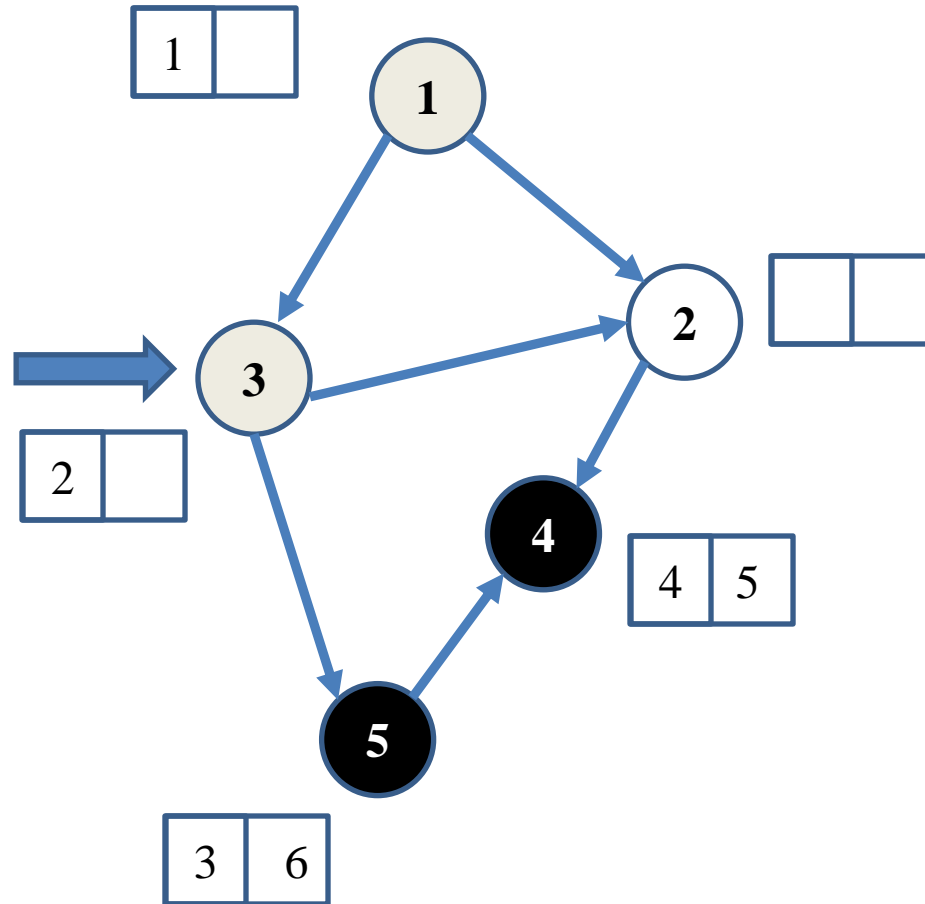
$\Pi(3) = 1$   
 $\Pi(5) = 3$   
 $\Pi(4) = 5$



# DFS in Action

Time : 6

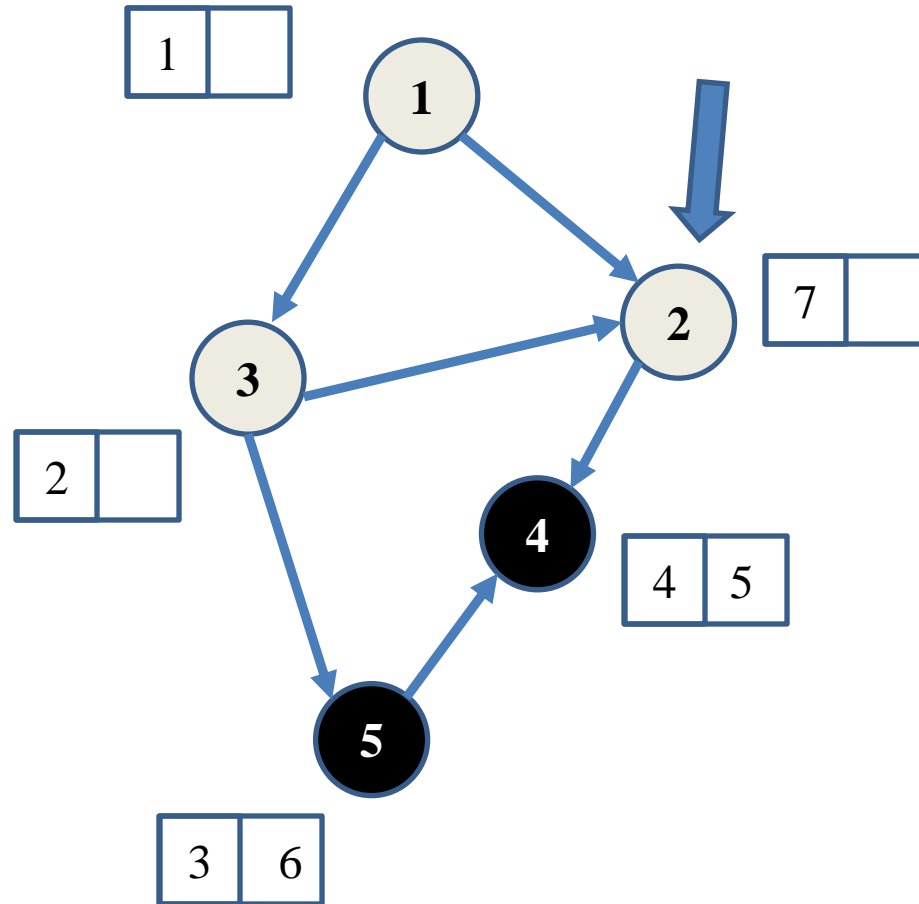
$\Pi(3) = 1$   
 $\Pi(5) = 3$   
 $\Pi(4) = 5$



# DFS in Action

Time : 7

$\Pi(3) = 1$   
 $\Pi(5) = 3$   
 $\Pi(4) = 5$   
 $\Pi(2) = 3$



# DFS in Action

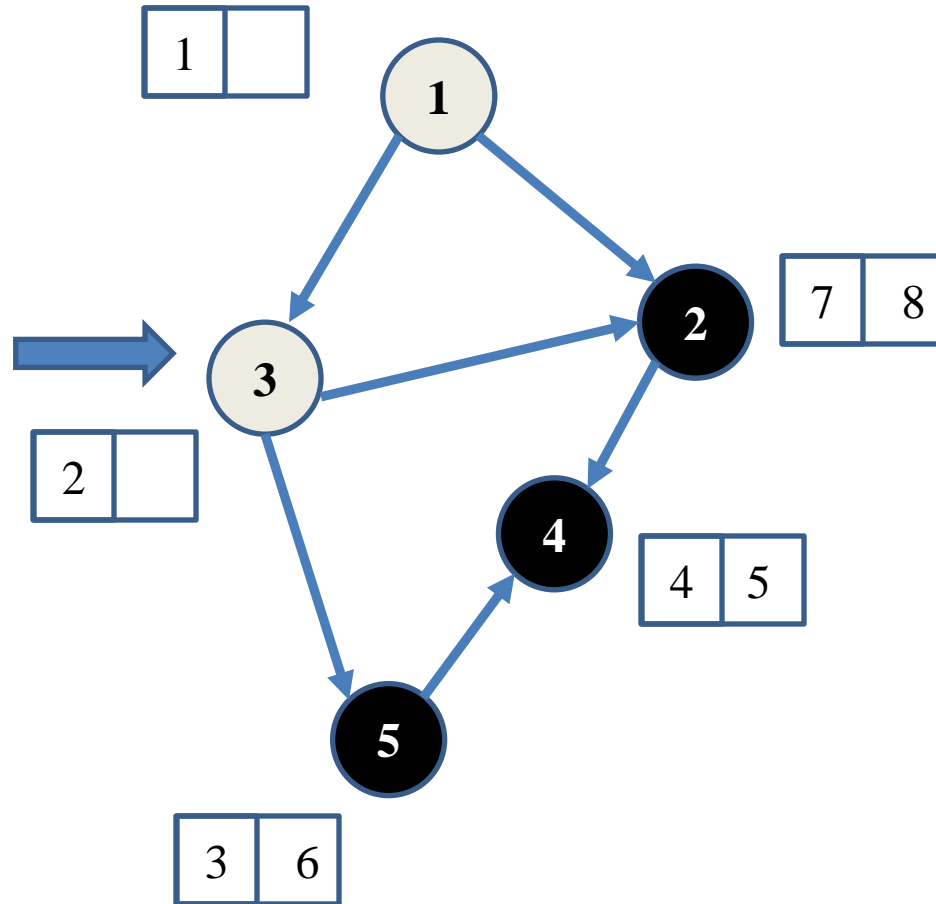
Time : 8

$$\Pi(3) = 1$$

$$\Pi(5) = 3$$

$$\Pi(4) = 5$$

$$\Pi(2) = 3$$

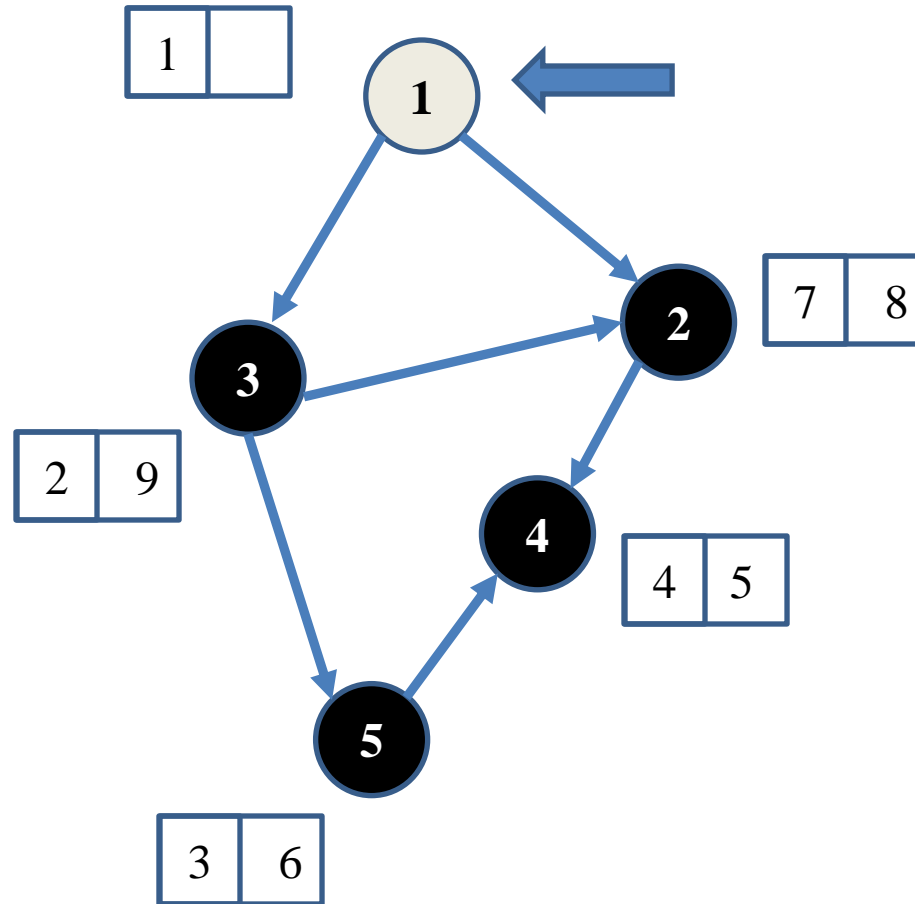




# DFS in Action

Time : 9

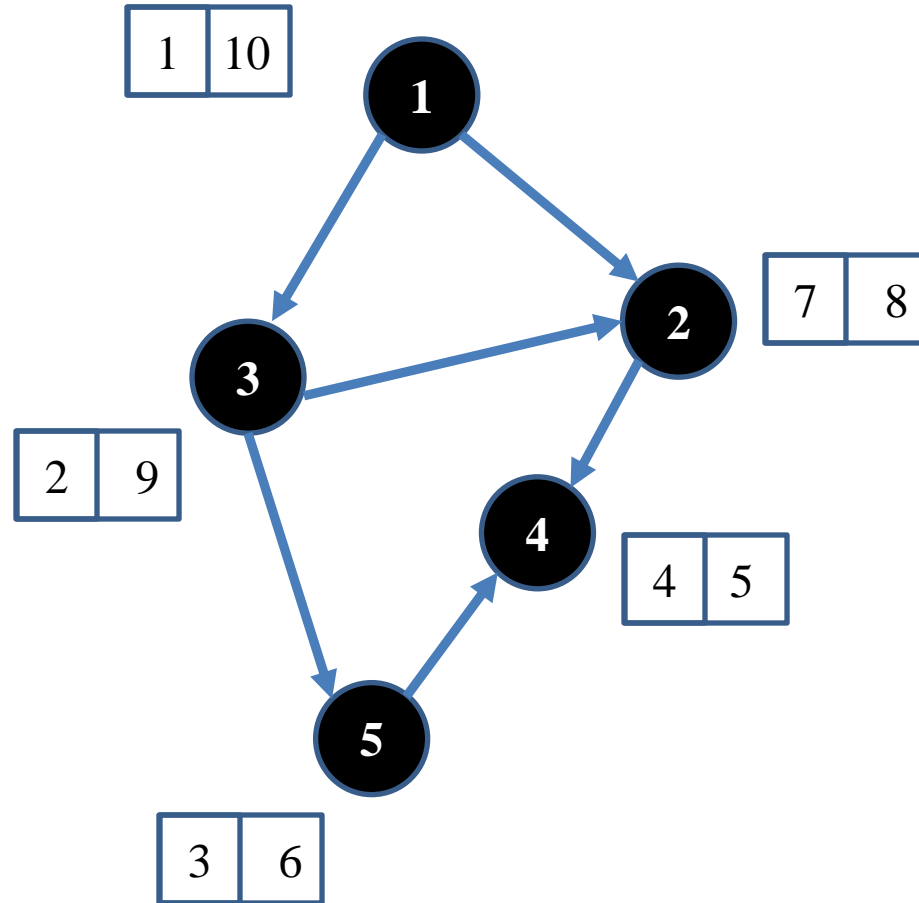
$\Pi(3) = 1$   
 $\Pi(5) = 3$   
 $\Pi(4) = 5$   
 $\Pi(2) = 3$



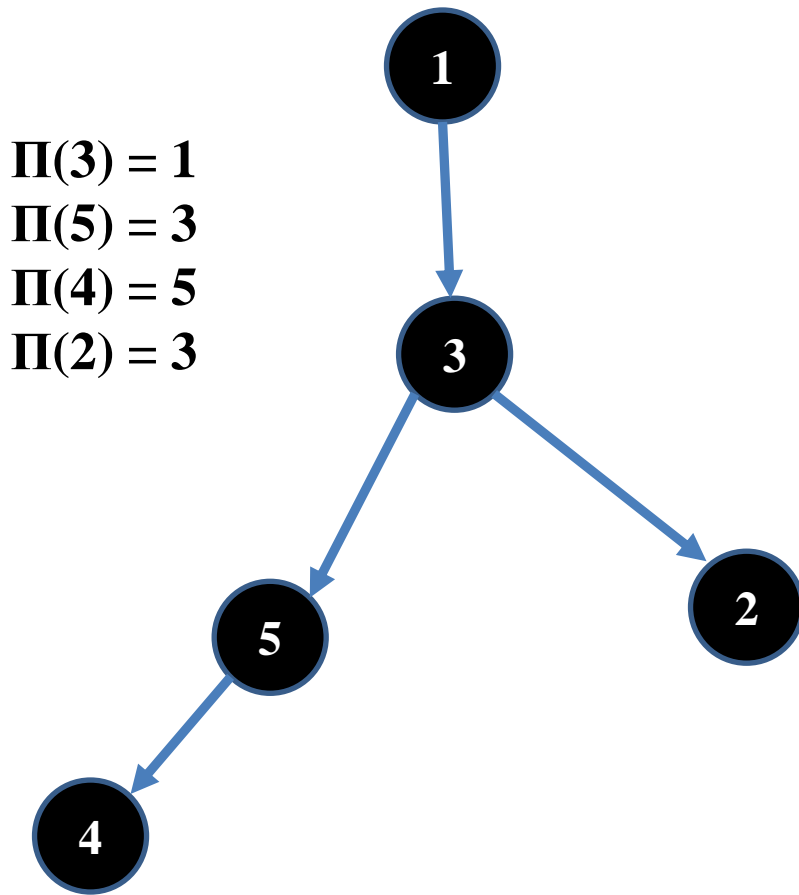
# DFS in Action

**Time : 10**

$\Pi(3) = 1$   
 $\Pi(5) = 3$   
 $\Pi(4) = 5$   
 $\Pi(2) = 3$

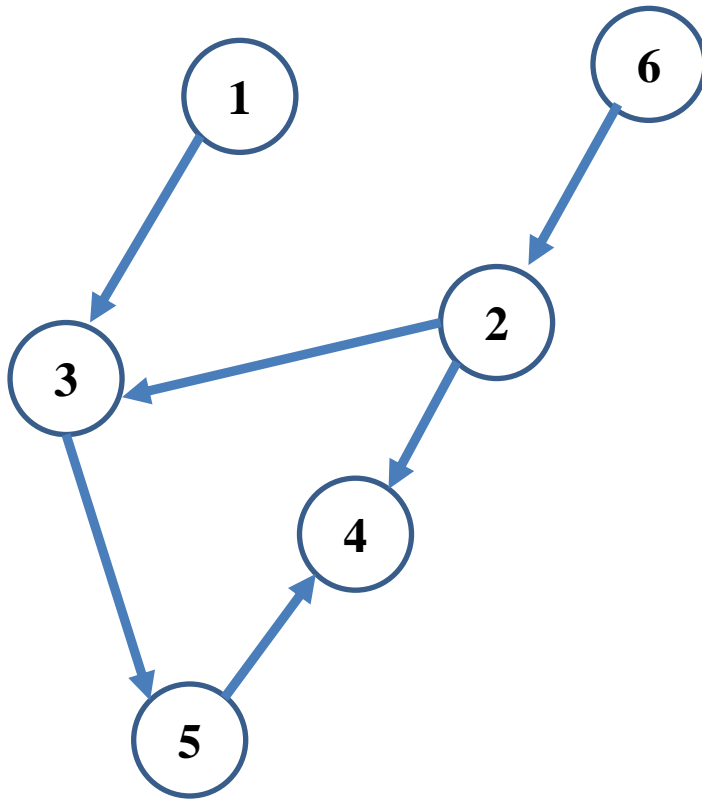


# DFS Tree



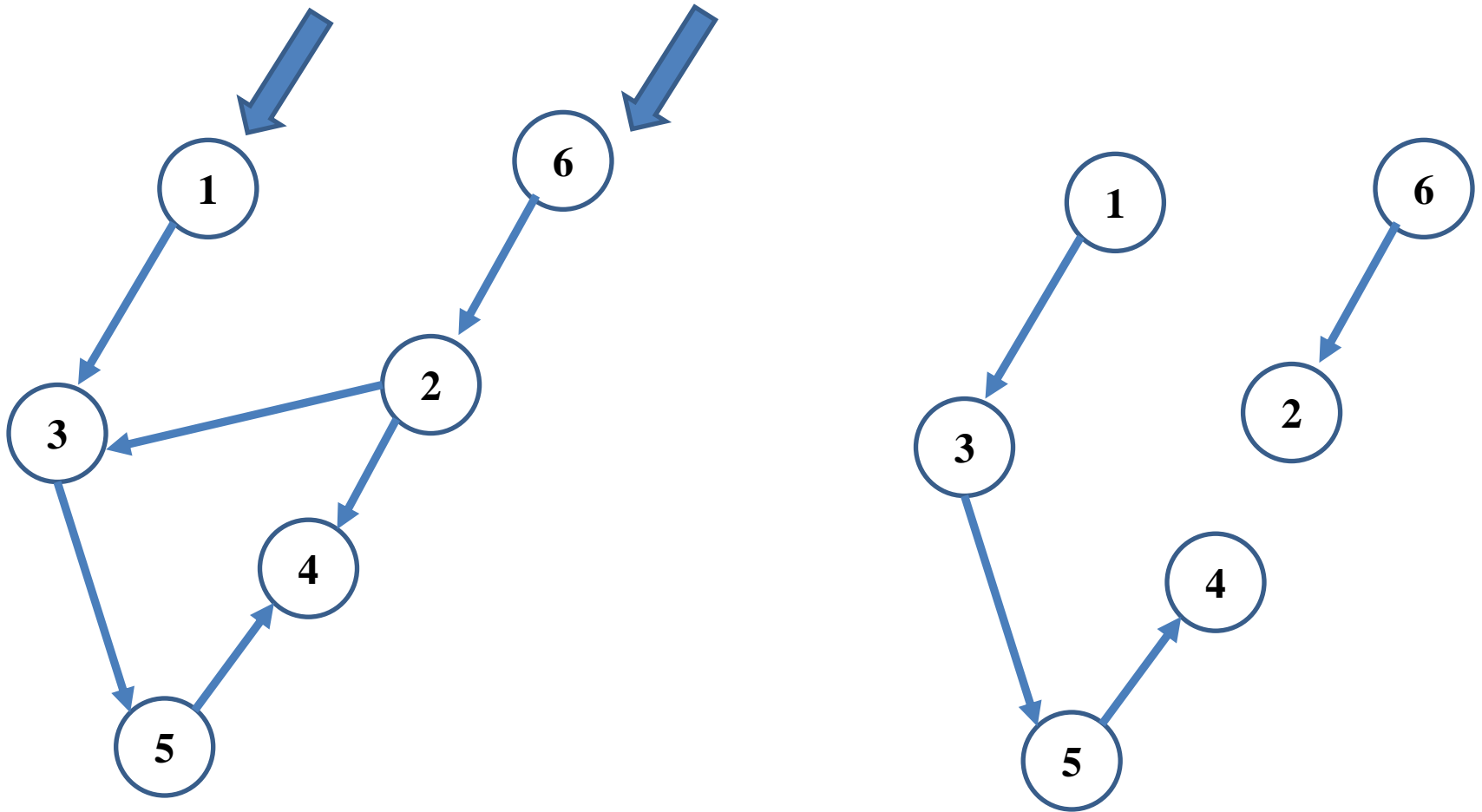
- DFS need not always form a single tree for a graph
- Sometimes, DFS may need to be performed multiple times from different starting nodes to discover all the nodes in the graph
- This gives rise to a number of DFS trees, collectively called a **DFS Forest**

# Need for Multiple DFS Runs



- Nodes 2 and 6 are not reachable from 1
- Node 1 is not reachable from 6
- There is no starting node such that all the vertices in the graph are reachable
- We need to run DFS from at least two vertices – say 1 and 6

# DFS Forest



**If DFS is run from node 1 and then from node 6, we get two DFS trees ( a DFS forest)**

# DFS Algorithm

***DFS(G)***

***for each vertex  $u$  in  $V[G]$***

***$colour[u] = WHITE$***

***$\Pi[u] = NIL$***

***time=0***

***for each vertex  $u$  in  $V[G]$***

***if  $colour[u] = WHITE$***

***DFS\_VISIT( $u$ )***

# DFS\_VISIT Algorithm

***DFS\_VISIT( $u$ )***

*colour*[ $u$ ] = *GRAY*

*time* = *time* + 1

*d*[ $u$ ] = *time*

***for each*  $v$  *in* *Adj*[ $u$ ]**

***if* *colour*[ $v$ ] = *WHITE***

*$\Pi$* [ $v$ ] =  $u$

*DFS\_VISIT*( $v$ )

*colour*[ $u$ ] = *BLACK*

*time* = *time* + 1

*f*[ $u$ ] = *time*

# Time Complexity of DFS

- Every node is explored EXACTLY ONCE ---  $\Theta(|V|)$
- For every node  $u$ , DFS explores all the edges in  $Adj[u]$
- When summed over all the nodes in the graph, this amounts to the number of edges in the graph

$$\sum_{u \in V} Adj[u] = \Theta(|V|)$$

- Thus, **total complexity of DFS is  $\Theta(|V| + |E|)$**

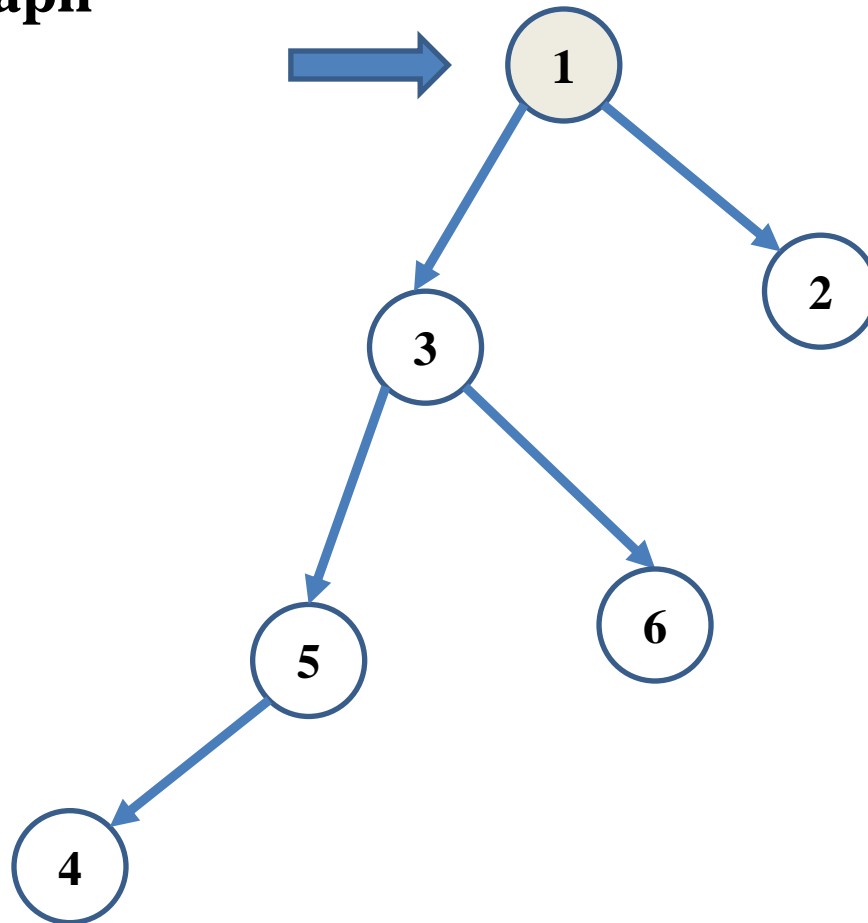


# Breadth First Search

- Breadth First Search (BFS) is another graph traversal algorithm
- As the name suggests, BFS explores all the neighbours of a node before exploring any other node
- When BFS hits a dead end on a path, it backtracks and starts exploring a new path from the previous node
- This behaviour is suggestive of a FIFO data structure
  - QUEUE

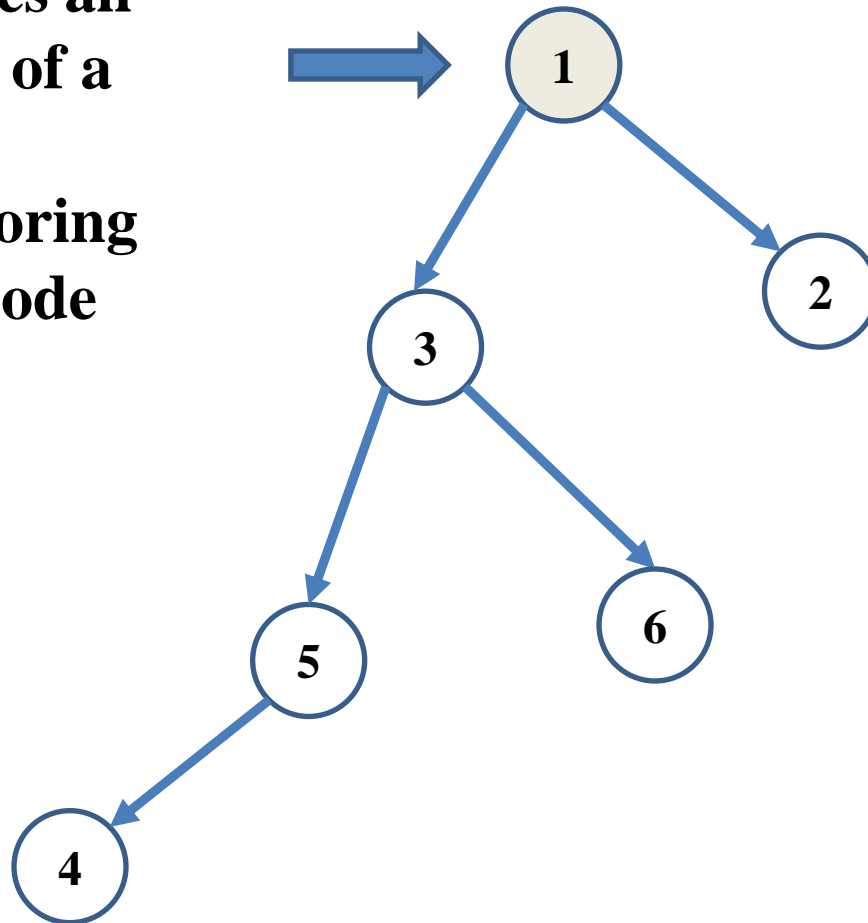
# Breadth First Search

**BFS is a graph traversal algorithm**



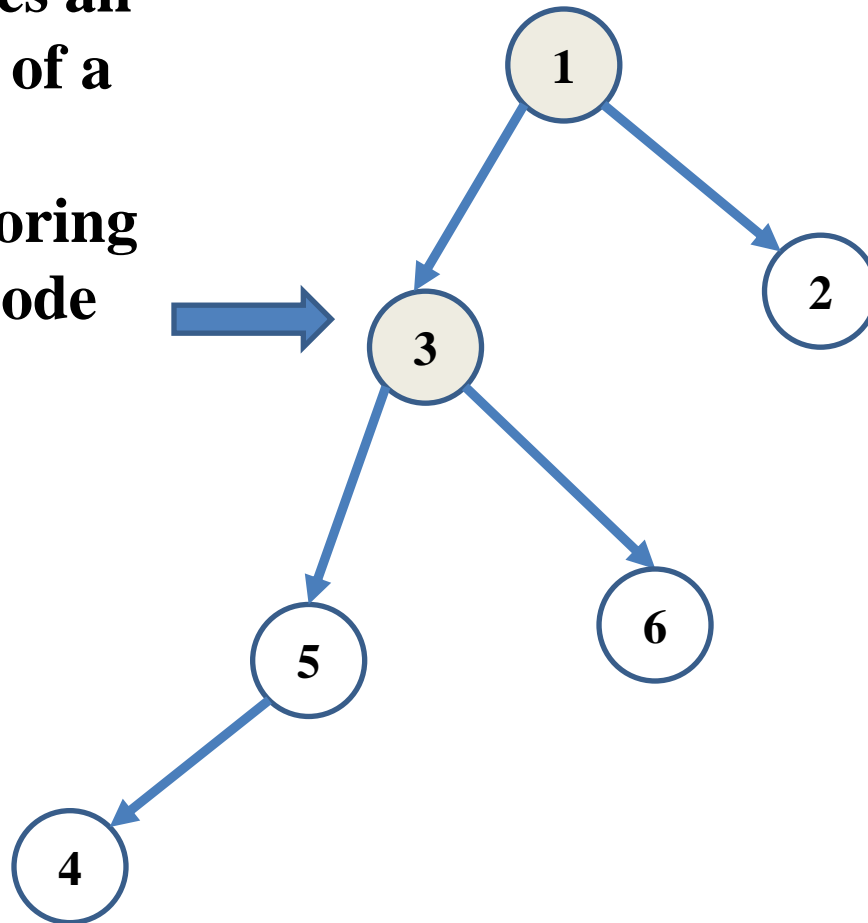
# Breadth First Search

**BFS explores all  
neighbours of a  
given node  
before exploring  
any other node**



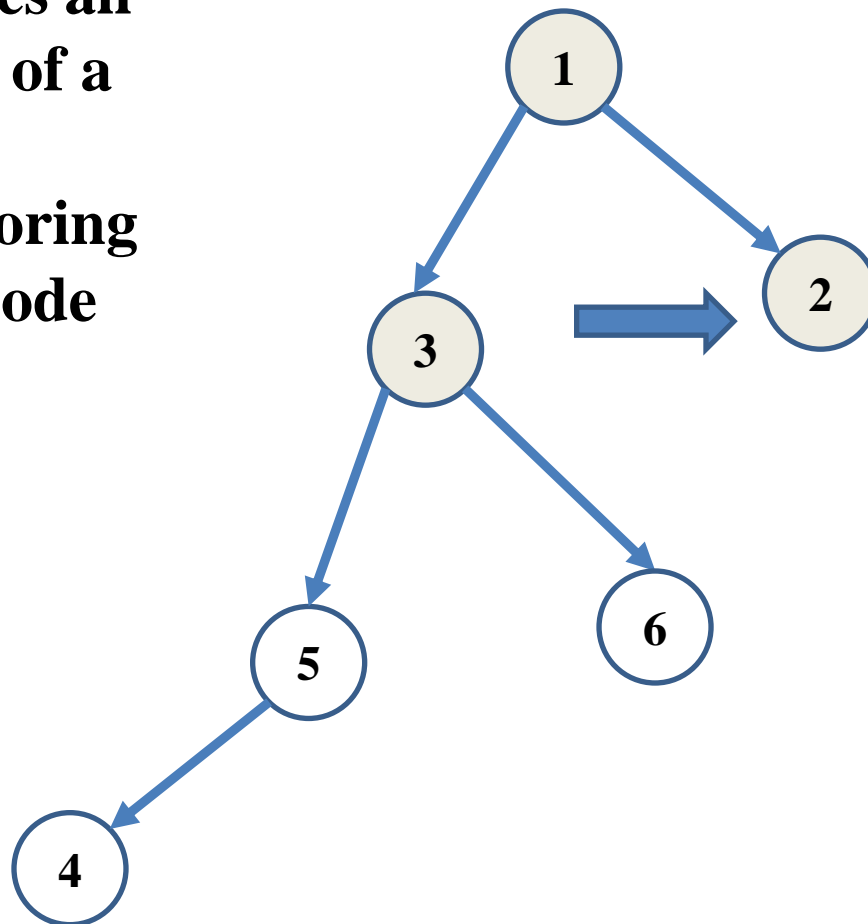
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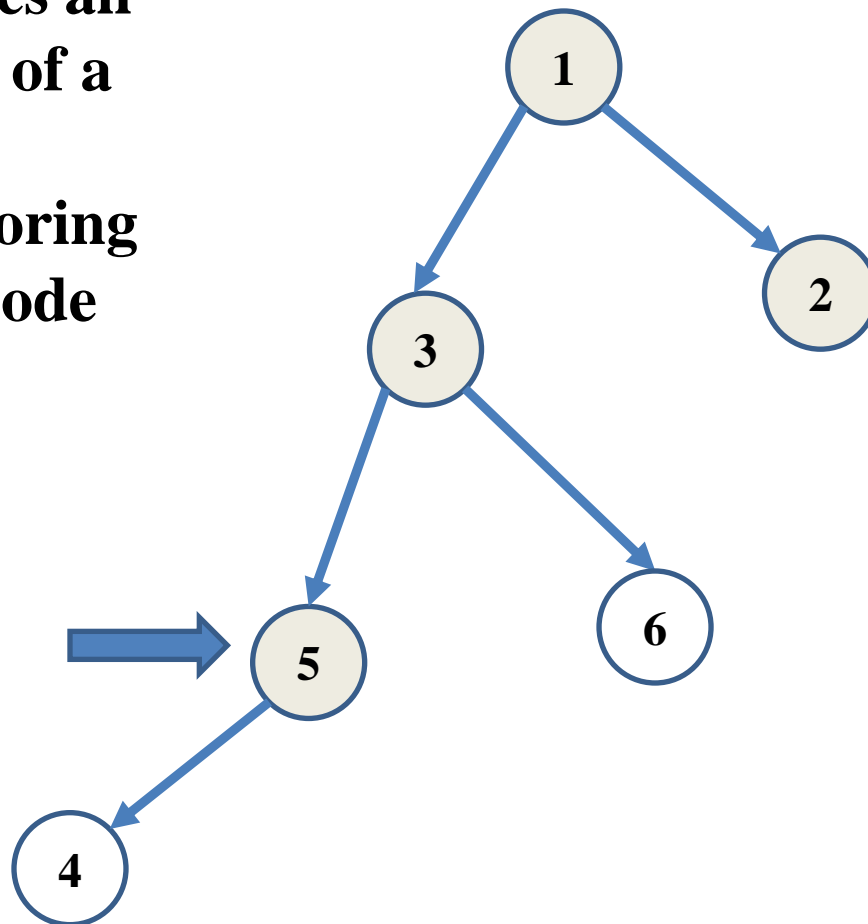
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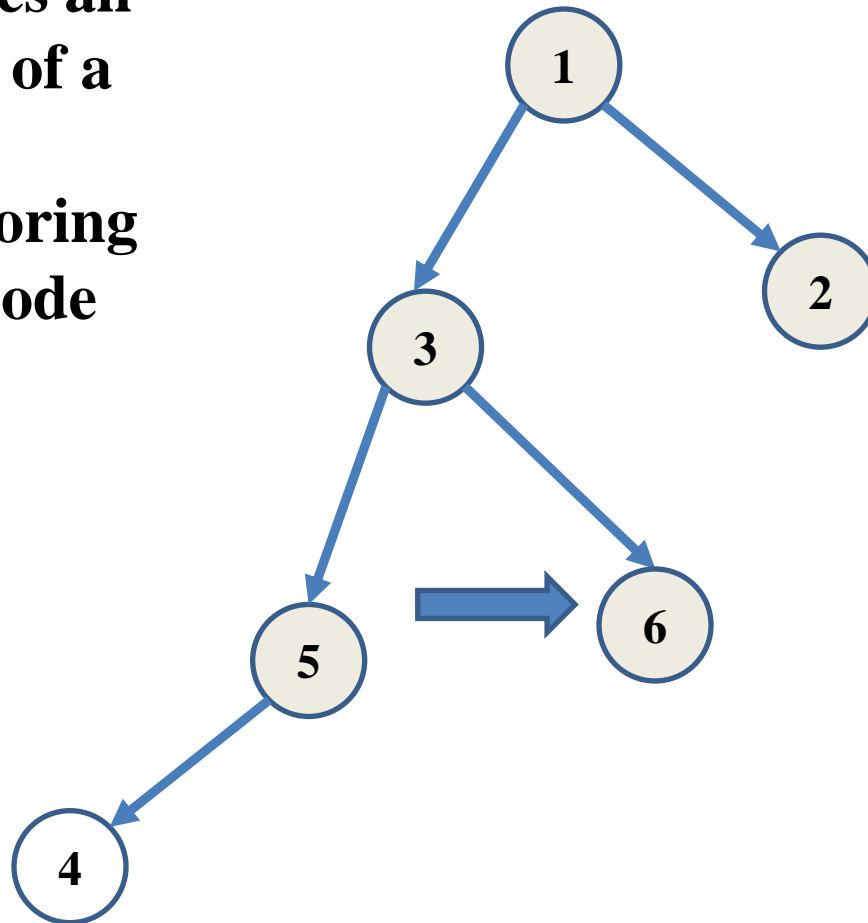
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before exploring  
any other node**



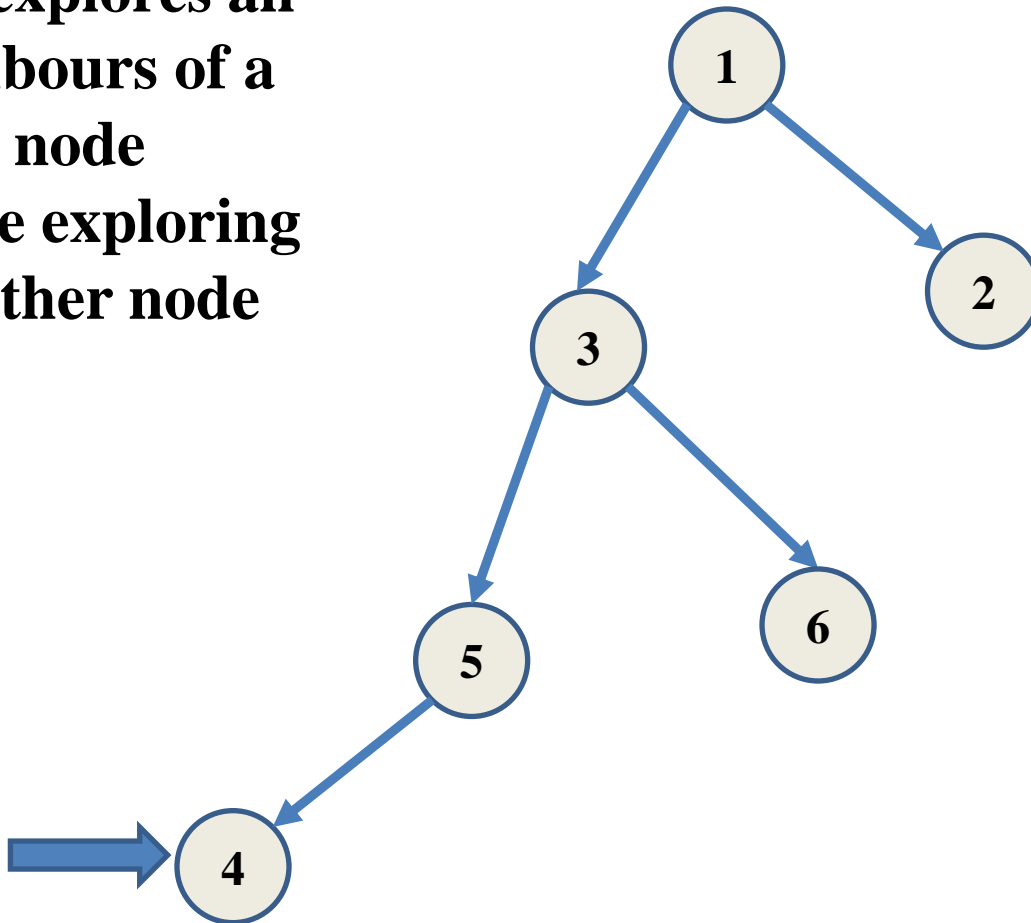
# Breadth First Search

**BFS explores all  
neighbours of a  
given node  
before exploring  
any other node**



# Breadth First Search

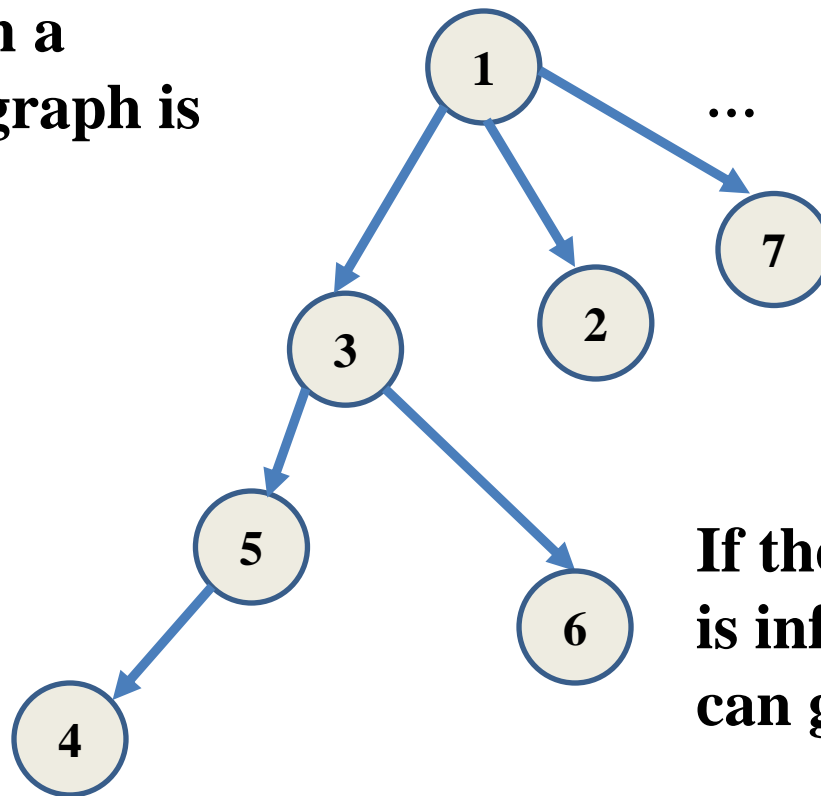
**BFS explores all  
neighbours of a  
given node  
before exploring  
any other node**





# Completeness of BFS

**BFS algorithm  
terminates with a  
solution if the graph is  
finite**



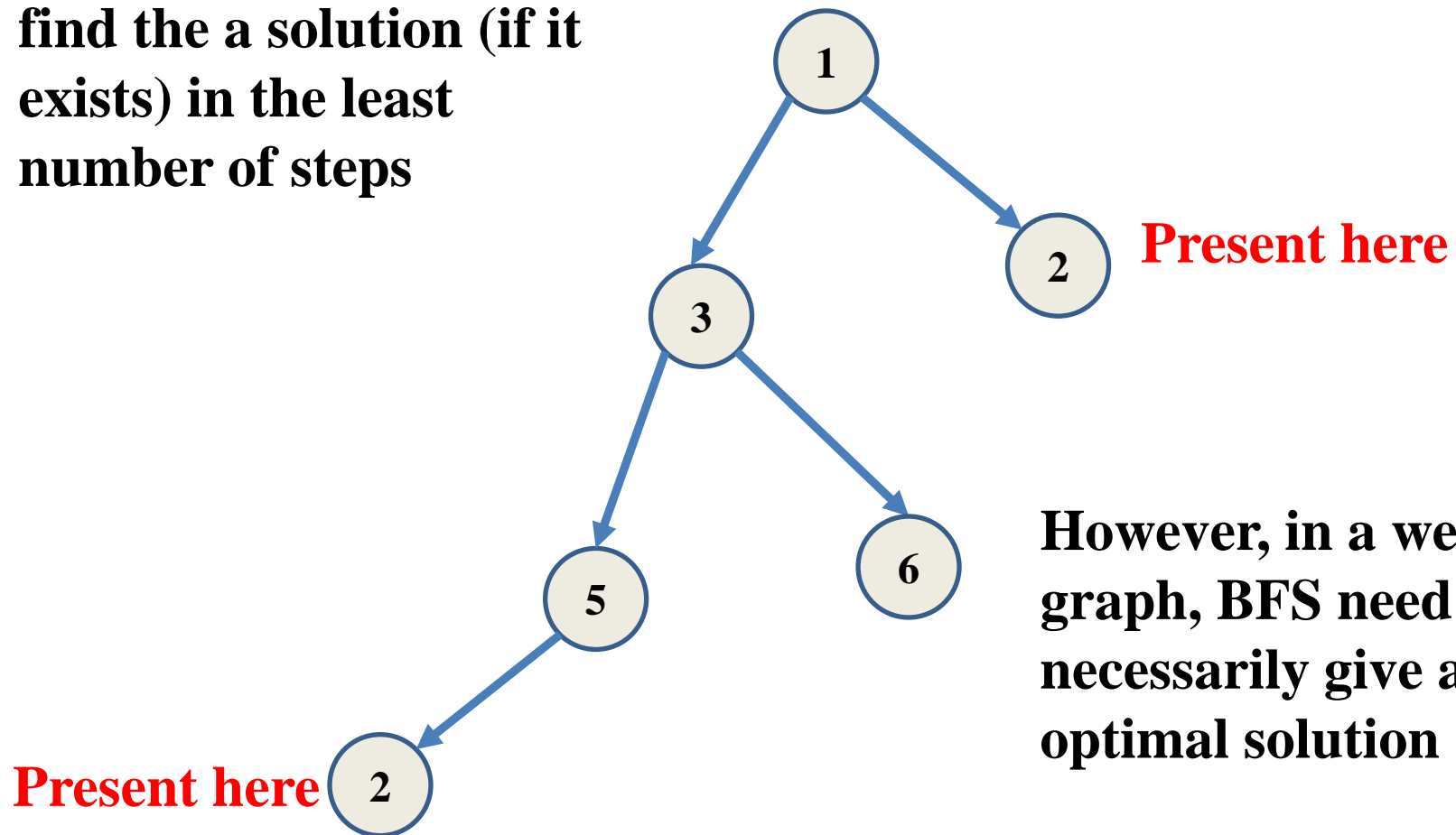
**Potentially continues to  $\infty$**

**If the branching factor  
is infinite, then BFS  
can get stuck in a loop**

# Optimality of BFS

**BFS algorithm will find the a solution (if it exists) in the least number of steps**

**Key = 2**



**However, in a weighted graph, BFS need not necessarily give an optimal solution**

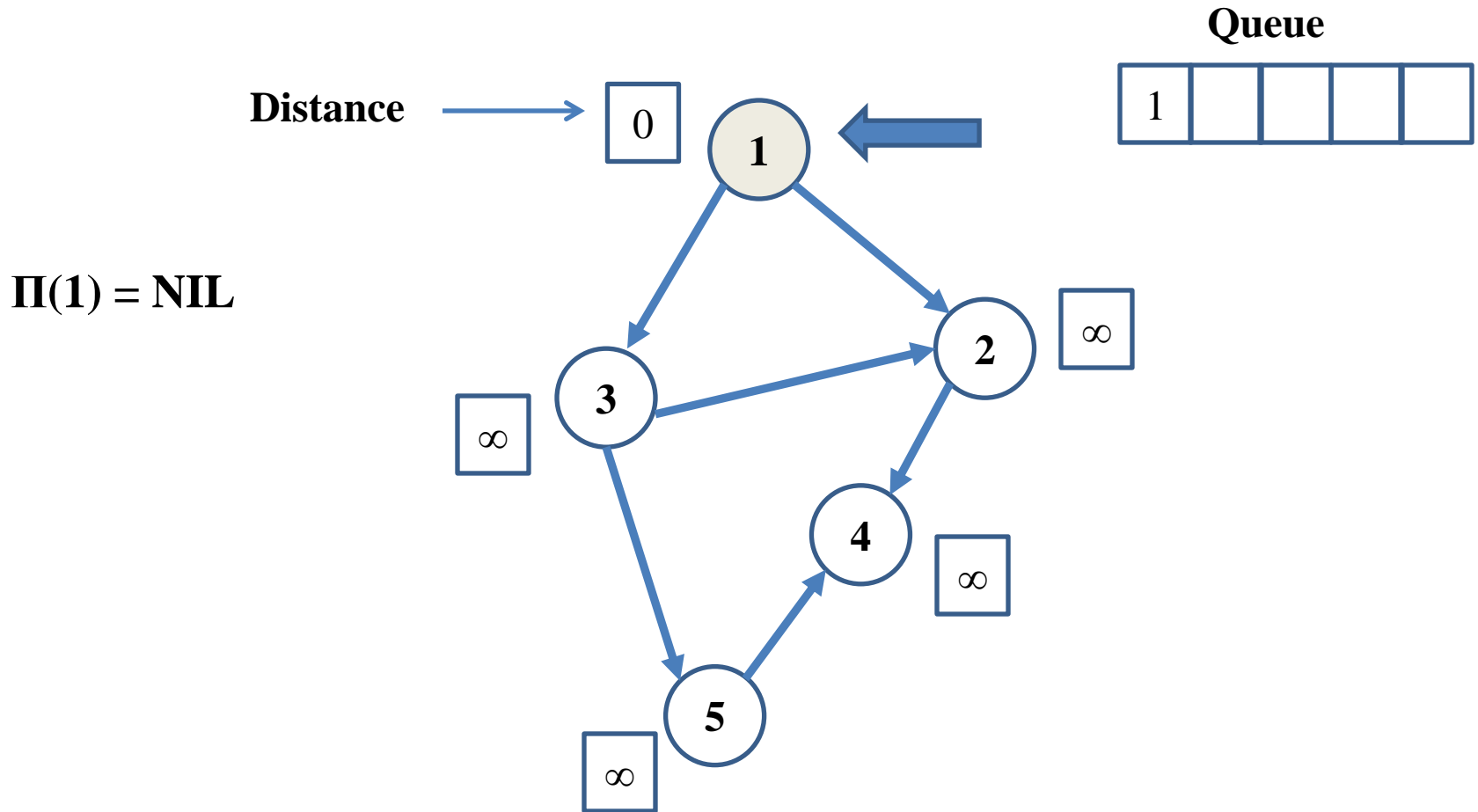
# Applications of BFS

- Cycle detection
- Path Detection
- Finding strongly connected components

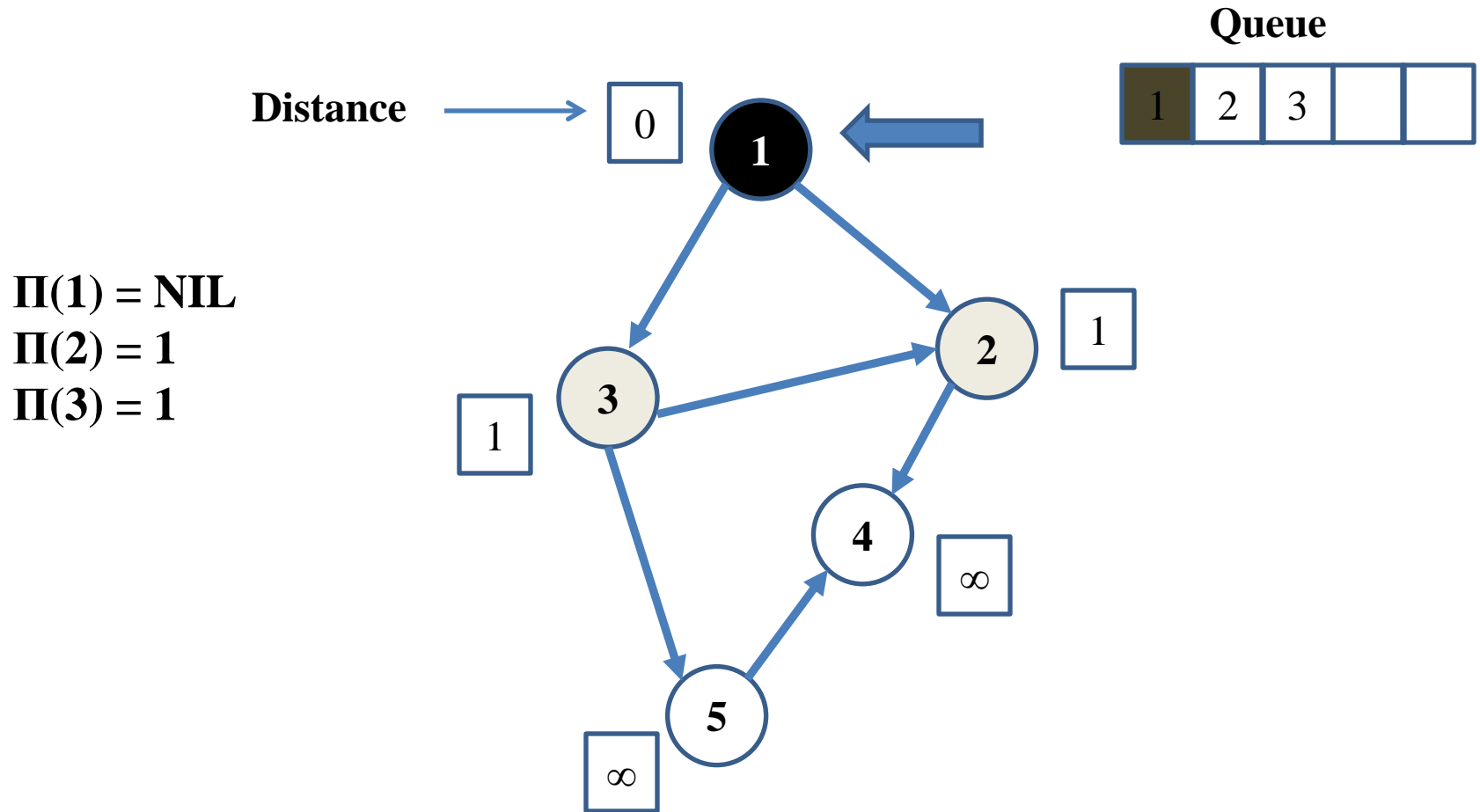
# Associated Notations

- **Nodes are assigned colours**
  - WHITE : The node has not been visited
  - GRAY : The node has been visited, but all of its branches have not been visited completely
  - BLACK : A node and its branches have been explored completely
- Every node is also assigned a distance value which is the number of steps it took to reach that node from the source vertex

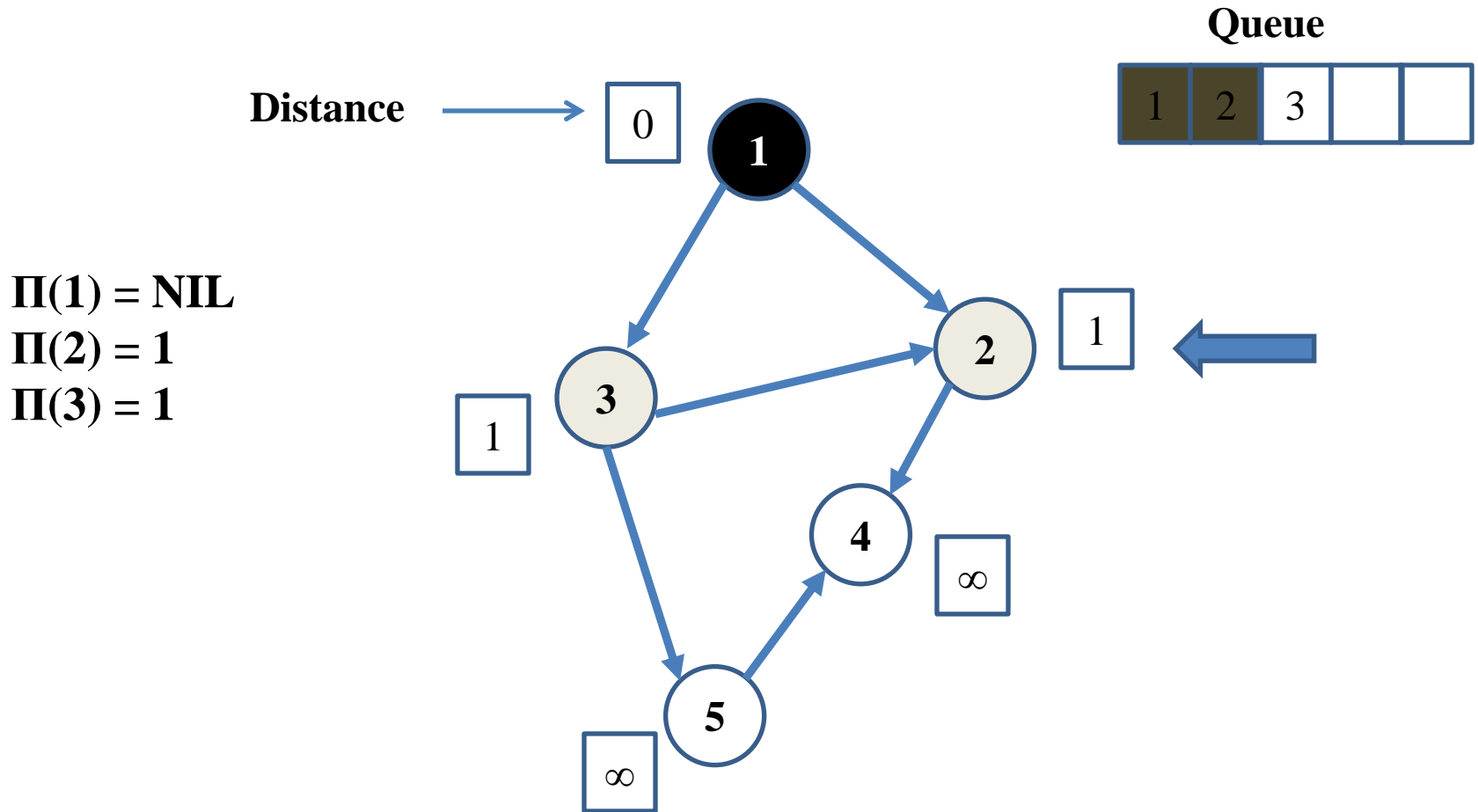
# BFS in Action



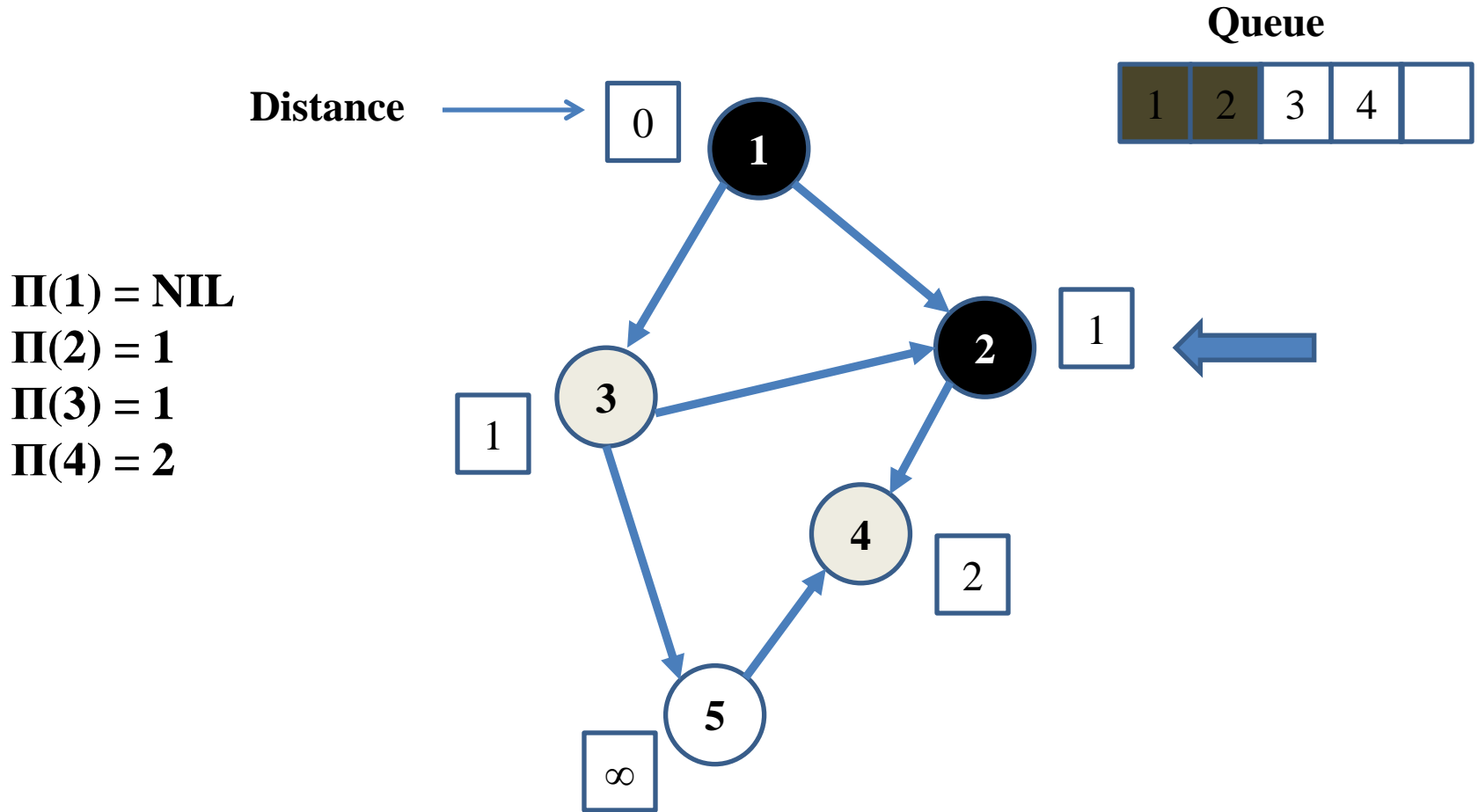
# BFS in Action



# BFS in Action

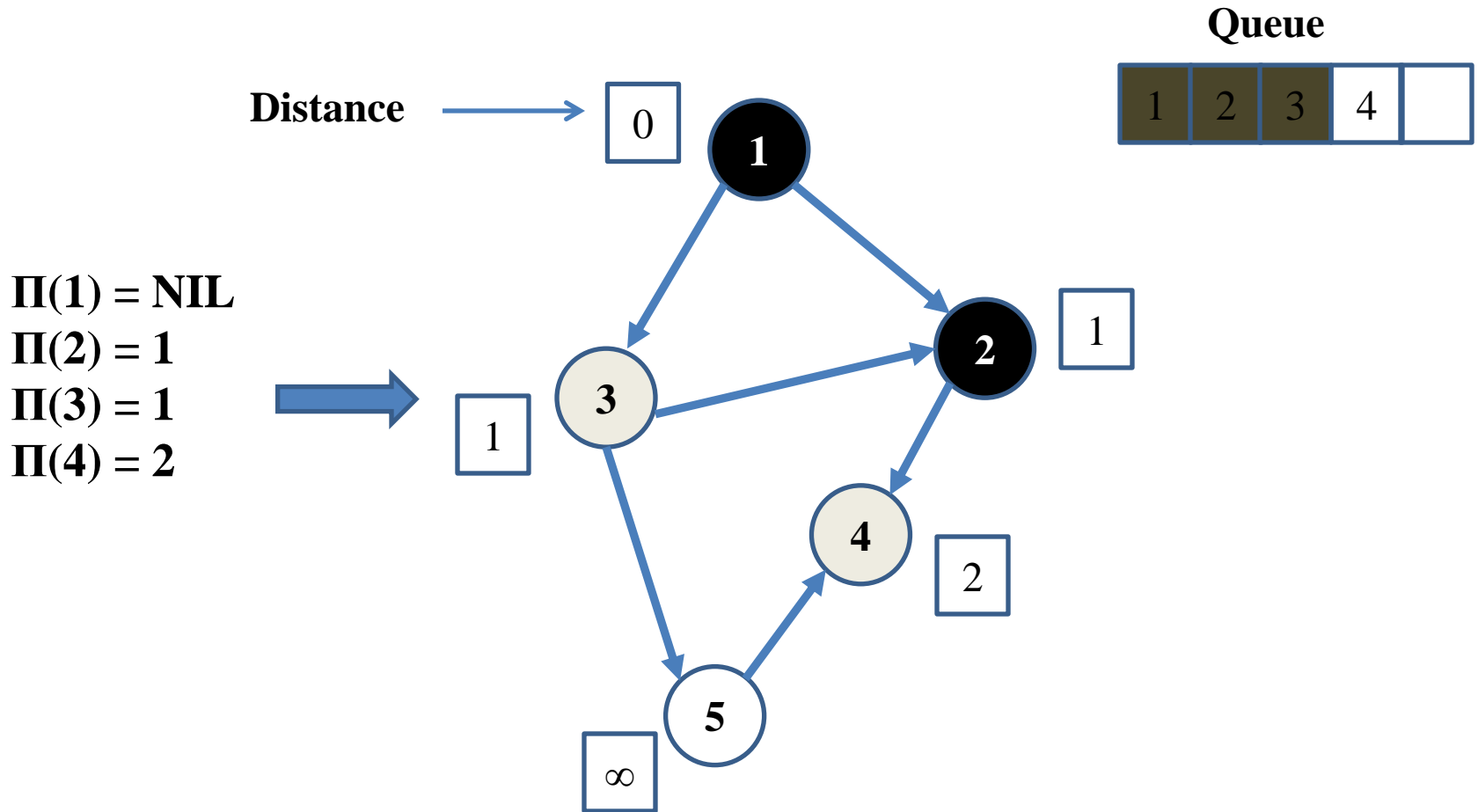


# BFS in Action

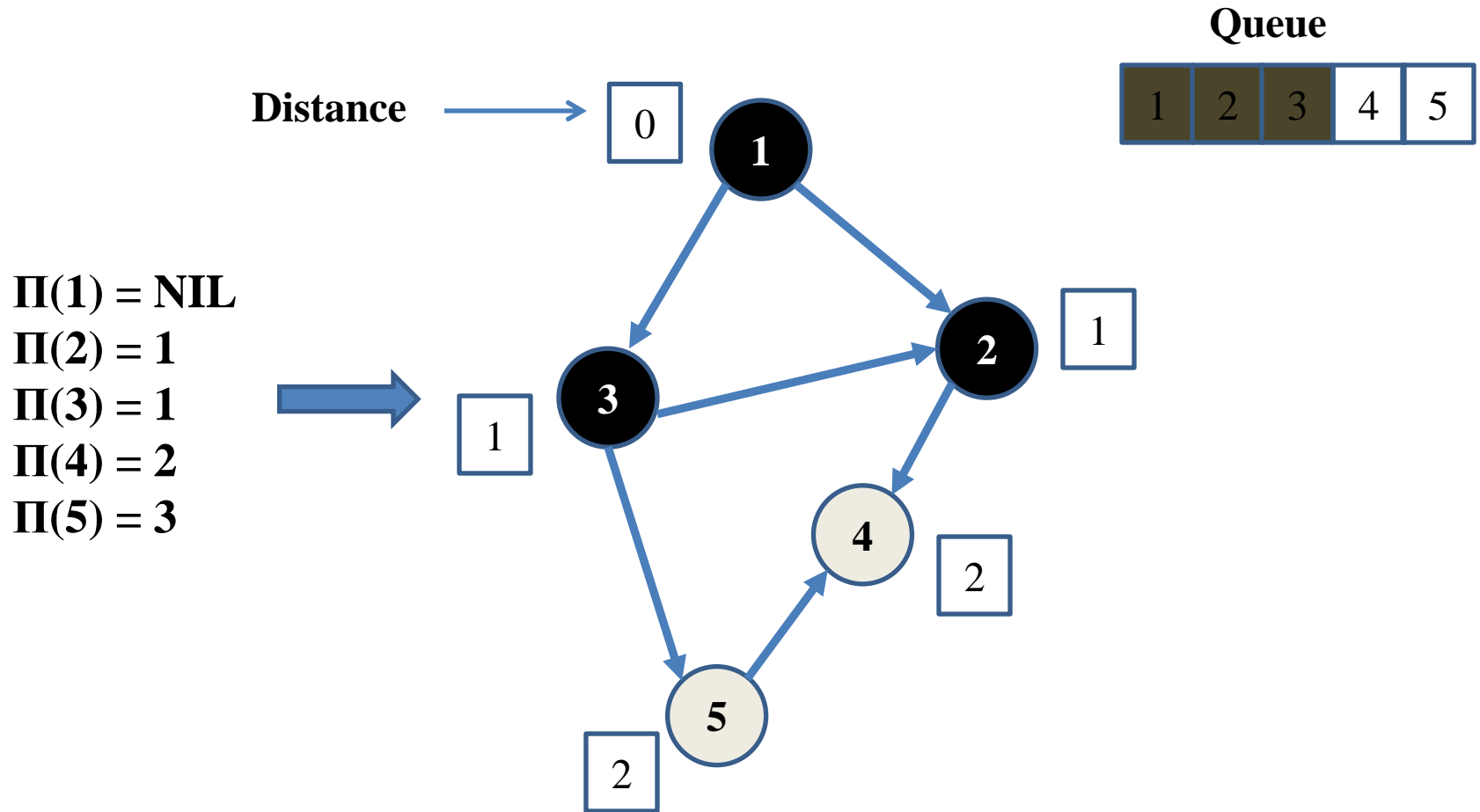




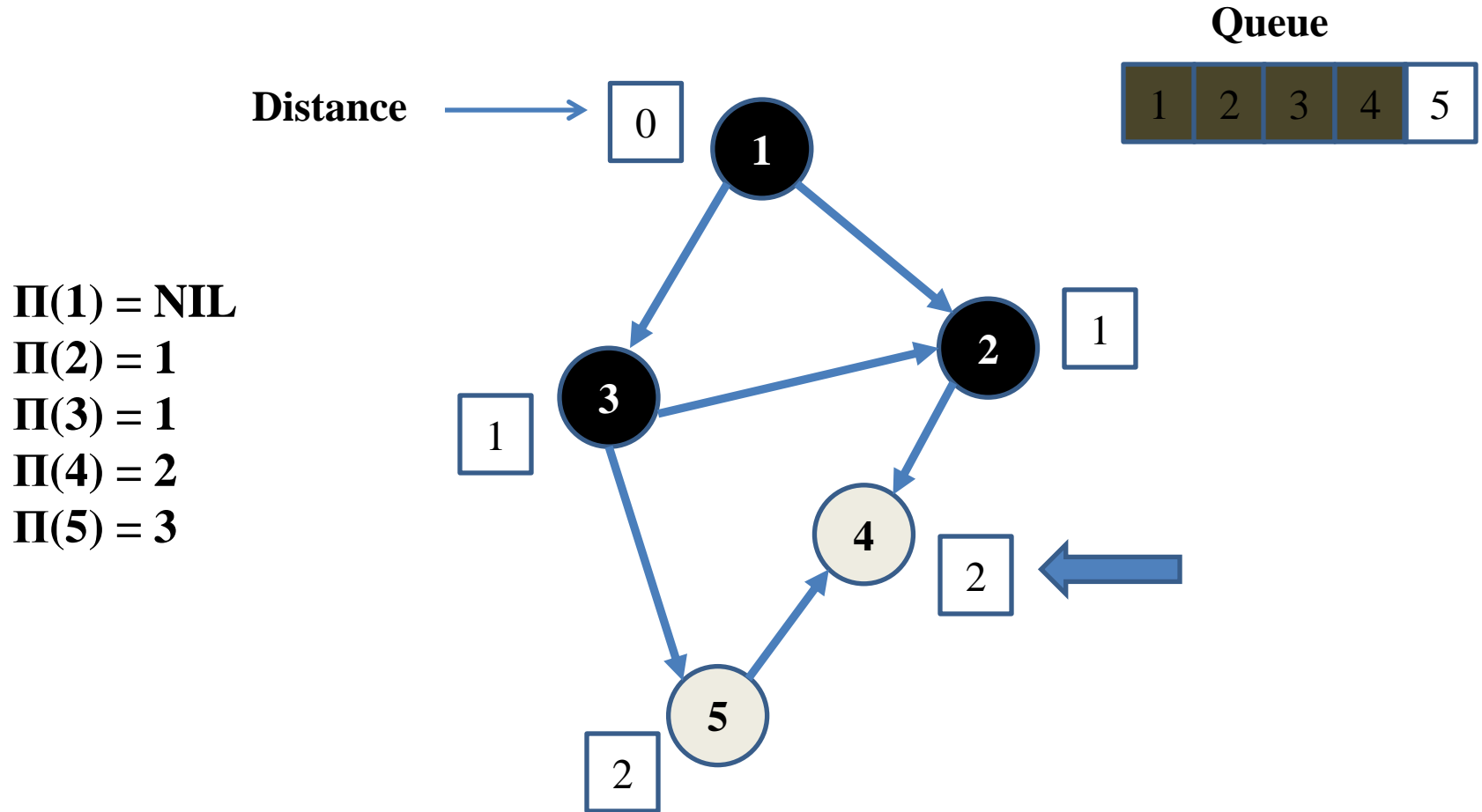
# BFS in Action



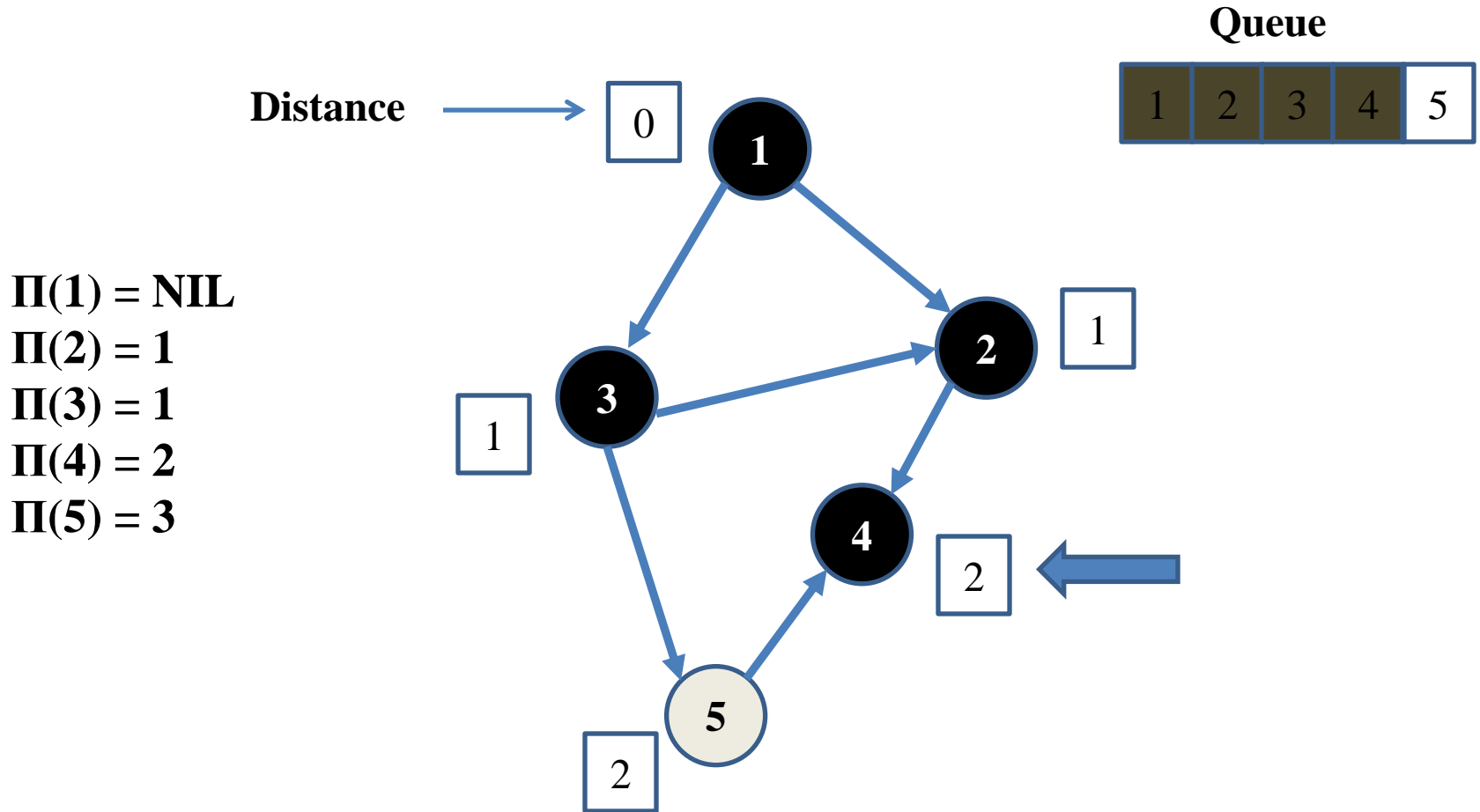
# BFS in Action



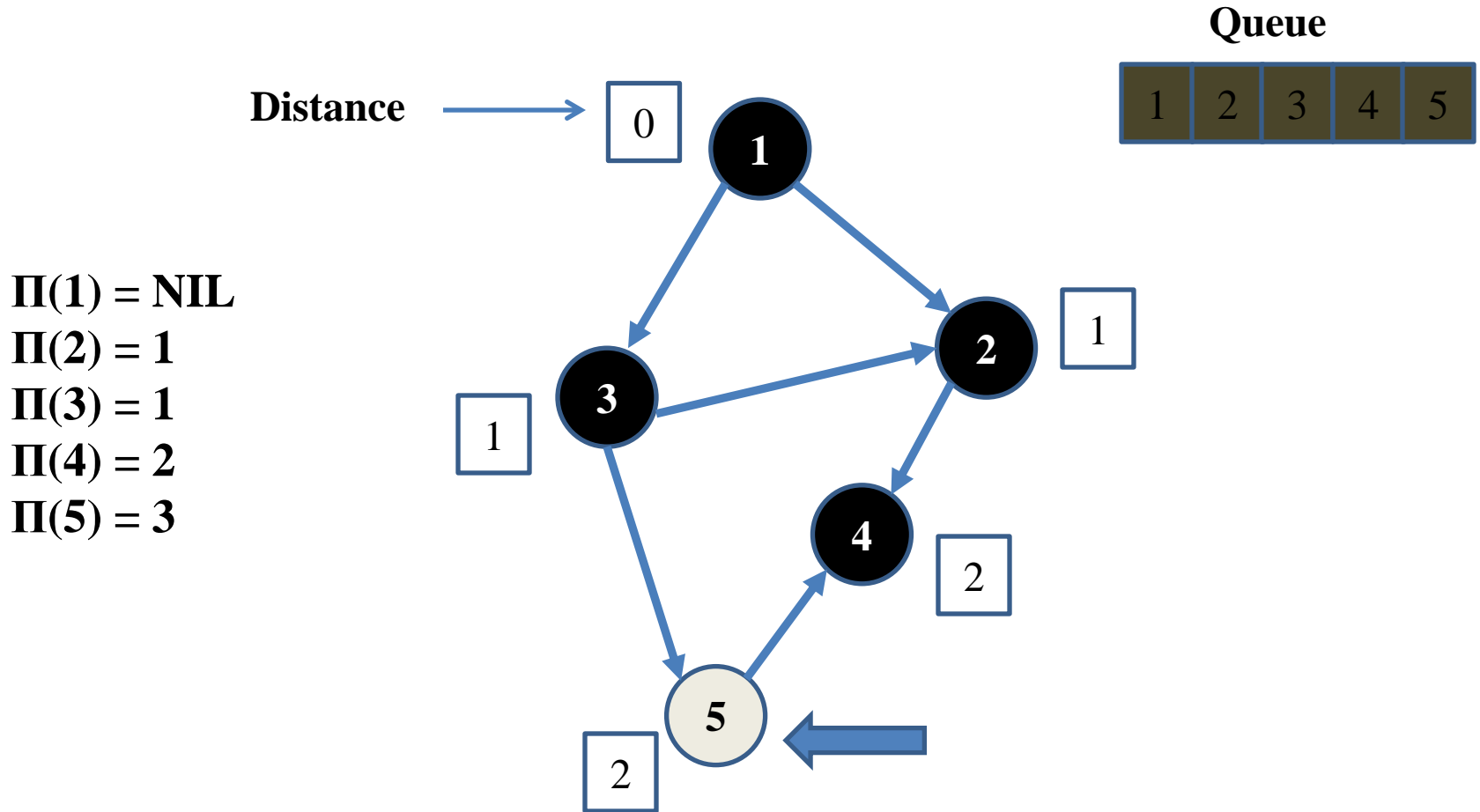
# BFS in Action



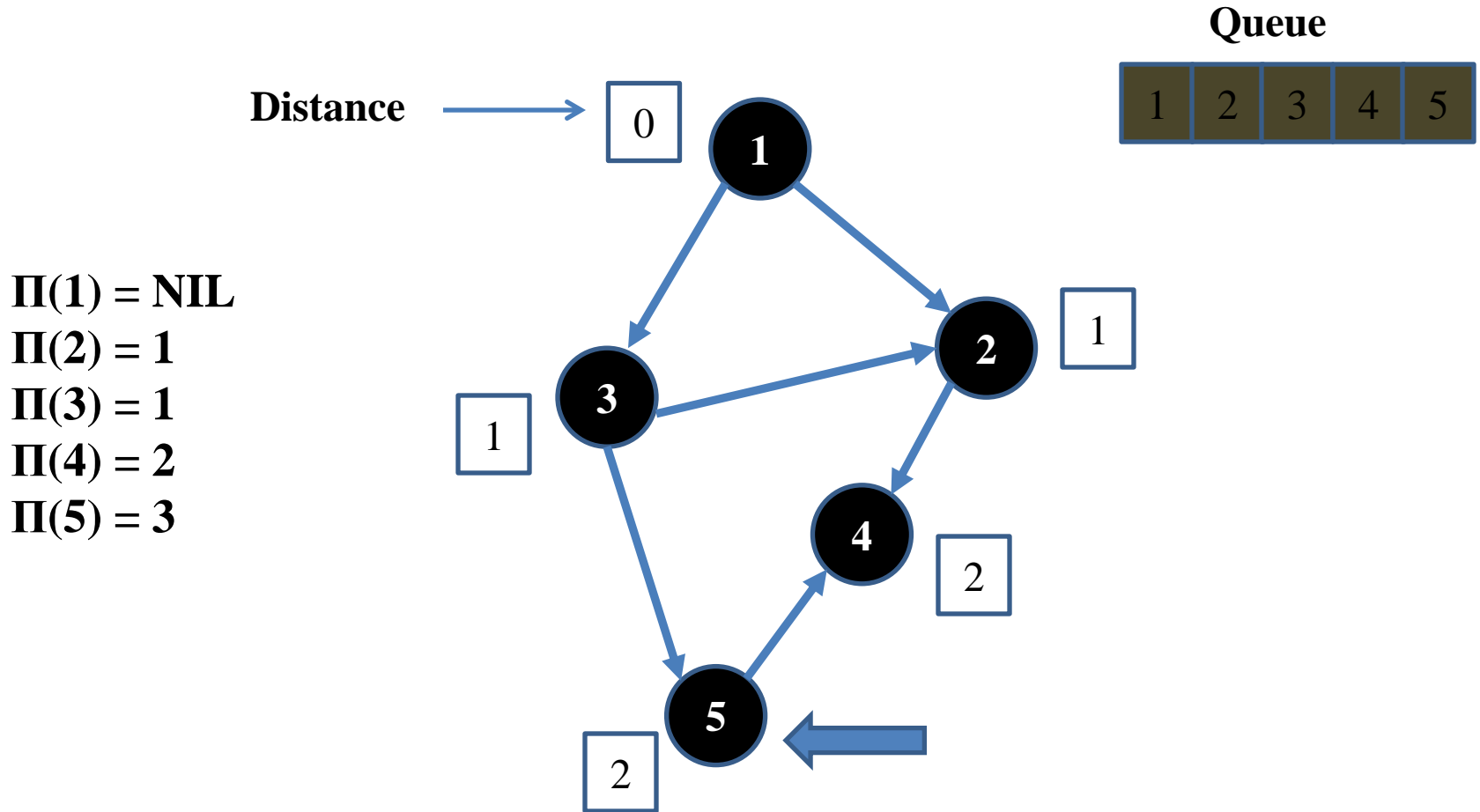
# BFS in Action



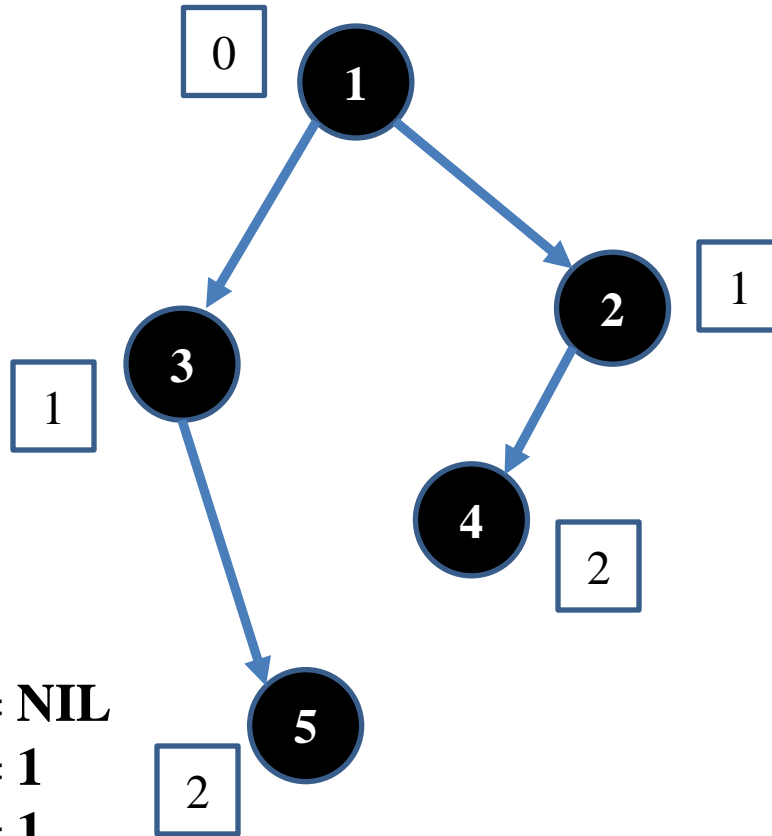
# BFS in Action



# BFS in Action



# BFS Tree



$\Pi(1) = \text{NIL}$

$\Pi(2) = 1$

$\Pi(3) = 1$

$\Pi(4) = 2$

$\Pi(5) = 3$

- The breadth first tree gives the nodes reachable from the source node
- For an unweighted graph, it also shows the shortest path from the source vertex to every other vertex in the graph

# BFS Algorithm

***BFS***( $G, s$ )

*for each* vertex  $u$  *in*  $V[G] - \{s\}$

$colour[u] = WHITE$

$\Pi[u] = NIL$

$d[u] = \infty$

$colour[s] = GRAY$

$d[s] = 0$

$\Pi[s] = NIL$

$Q = \phi$

*ENQUEUE* ( $Q, s$ )



# BFS Algorithm

*while*  $Q \neq \emptyset$

*time* = *time* + 1

$d[u] = \textit{time}$

***for each***  $v$  ***in***  $\textit{Adj}[u]$

***if***  $\textit{colour}[v] = \textit{WHITE}$

$\Pi[v] = u$

$\textit{DFS\_VISIT}(v)$

$\textit{colour}[u] = \textit{BLACK}$

*time* = *time* + 1

$f[u] = \textit{time}$

# Time Complexity of BFS

- Every node is explored EXACTLY ONCE ---  $\Theta(|V|)$
- For every node  $u$ , BFS explores all the edges in  $Adj[u]$
- When summed over all the nodes in the graph, this amounts to the number of edges in the graph

$$\sum_{u \in V} Adj[u] = \Theta(|V|)$$

- Thus, **total complexity of BFS is also  $\Theta(|V| + |E|)$**

**Thank You**