## Data Structures and Algorithms (CS210A)

### Lecture 27

- Quick revision of Depth First Search (DFS) Traversal
- An O(m + n): algorithm for biconnected components of a graph

# Quick revision of Depth First Search (DFS) Traversal

## **DFS traversal of** *G*

DFS(v)

**DFS-traversal(G)** 

}

{ dfn ← 0; For each vertex v∈ V { Visited(v) ← false } For each vertex v ∈ V { If (Visited(v) = false) DFS(v) }

### **DFN** number

### **DFN**[**x**] :

The number at which **x** gets visited during DFS traversal.



## DFS(v) computes a tree rooted at v



If x is ancestor of y then DFN[x] < DFN[y]

Question: Is a DFS tree unique ? Answer: No.

#### **Question**:

Can any rooted tree be obtained through DFS ?

Answer: No.

## Always remember this picture



non-tree edge  $\rightarrow$  back edge

of the graph

## Verifying bi-connectivity of a graph

An O(m + n) time algorithm A <u>single</u> **DFS** traversal

## An O(m + n) time algorithm

- A formal characterization of the problem. (articulation points)
- Exploring <u>relationship</u> between articulation point & DFS tree.

• Using the relation **cleverly** to design an efficient algorithm.





The removal of any of {*v,f,u*} can destroy connectivity.

**v**,**f**,**u** are called the **articulation points** of **G**.

## A formal definition of articulaton point

**Definition:** A vertex **x** is said to be **articulation point** if

∃ *u*,*v* different from *x* 

such that every path between *u* and *v* passes through *x*.



**Observation:** A graph is biconnected if none of its vertices is an articulation point.

#### AIM:

Design an **algorithm** to compute all **articulation points** in a given graph.

### **Some observations**



- A leaf node can never be an a.p. ?
- **Root** is an **a.p**. iff it has two or more children.



### **Necessary** and **Sufficient** condition for **x** to be articulation point



#### Theorem1:

An internal node *x* is **articulation point** iff *x* has <u>**at least**</u> one child *y* s.t. **no** back edge from **subtree(y)** to **ancestor** of *x*.

→ No back edge from subtree(y) going to a vertex "higher" than x.



### **Necessary** and **Sufficient** condition for **x** to be articulation point



#### Theorem1:

An internal node **x** is **articulation point** iff **x** has **at least** one child **y** s.t.

**no** back edge from **subtree(y)** to **ancestor** of **x**.



High\_pt(v):

DFN of the <u>highest ancestor</u> of **v** 

to which there is a back edge from **subtree(v**).

#### Theorem2:

An internal node x is articulation point iff it has a child, say y, in DFS tree such that  $\frac{\text{High}_pt(y)}{2} \quad \text{DFN}(x).$ 

#### Theorem2:



- $\rightarrow$  O(m) time in the worst case to compute High\_pt(v) of a vertex v.
- $\rightarrow$  O(*mn*) time algorithm  $\otimes$

### How to compute **High\_pt(v)** efficiently ?



### How to compute **High\_pt(v)** efficiently ?



**Question:** Can we express High\_pt(v) in terms of its children and proper ancestors?

 $High_pt(v) =$ 

$$\min_{(v,w) \in E} \begin{cases} \text{High_pt}(w) \\ \text{DFN}(w) \end{cases}$$

If w=child(v) If w = proper ancestor of v

# The novel algorithm

Output : an array AP[] s.t. AP[v] = true if and only if v is an articulation point.

### Algorithm for articulation points in a graph G

DFS(v)

}

```
{ Visited(v) \leftarrow true; DFN[v] \leftarrow dfn ++; High_pt[v] \leftarrow \infty;
  For each neighbor w of v
          if (Visited(w) = false)
  {
         { DFS(w); Parent(w) \leftarrow v;
              .....;
              }
          ........
  }
DFS-traversal(G)
{ dfn \leftarrow 0;
  For each vertex v \in V { Visited(v) \leftarrow false; AP[v] \leftarrow false }
  For each vertex v \in V {
                                  If (Visited(v) = false) DFS(v)
                                                                            }
```

### Algorithm for articulation points in a graph G

);

DFS(v)

```
{ Visited(v) \leftarrow true; DFN[v] \leftarrow dfn ++; High_pt[v] \leftarrow \infty;
For each neighbor w of v
```

{ if (Visited(w) = false)

{ Parent(w)  $\leftarrow$  v; DFS(w);

 $\label{eq:High_pt(v)} \leftarrow \min( , \\ \label{eq:High_pt(w)} \geq \mathsf{DFN}[v] \\ \end{tabular}$ 

} `

}

```
DFS-traversal(G)
```

```
{ dfn ← 0;
```

```
For each vertex v \in V {Visited(v) \leftarrow false; AP[v] \leftarrow false }For each vertex v \in V {If (Visited(v) = false) DFS(v) }
```

## Conclusion

**Theorem2**: For a given graph G=(V,E), all articulation points can be computed in O(m + n) time.

### **Data Structures**



**Binary Heap** 



Simplicity

**Binary Search Trees** 

## Неар

**Definition:** a tree data structure where :



# **Operations** on a heap

### **Query Operations**

• Find-min: report the smallest key stored in the heap.

### **Update Operations**

- **CreateHeap(H)** : Create an empty heap **H**.
- Insert(x,H) : Insert a <u>new key</u> with value x into the heap H.
- Extract-min(H) : delete the <u>smallest</u> key from H.
- **Decrease-key**(p,  $\Delta$ , H) : decrease the value of the key p by amount  $\Delta$ .
- Merge(H1,H2) : Merge two heaps H1 and H2.

### Why heaps when we can use a binary search tree ?

Compared to binary search trees, a heap is usually

-- much simpler and

-- more efficient

# **Existing heap data structures**

- Binary heap
- Binomial heap
- Fibonacci heap
- Soft heap

## Can we implement a binary tree using an array ?





**Question:** What does <u>the implementation</u> of a tree data structure <u>require</u>?



Answer: a mechanism to

- access **parent** of a node
- access **children** of a node.

## A complete binary tree



A complete binary of 12 nodes.

## A complete binary tree

