

CSEP 521 Applied Algorithms Spring 2005

Course Introduction
Graph Algorithms

Outline for the Evening

- Course administration
- Algorithm Design Process
- Spanning Tree
 - Depth-First Search
 - In-class exercise
 - Breath-First Search
- Minimum Spanning Tree
- Set Disjoint Union / Find

Lecture 1 - Intro, Graph Algorithms

2

Instructors

- Instructor
 - Richard Ladner
 - ladner@cs.washington.edu
 - 206 543-9347
- TA
 - Neva Cherniavsky
 - (nchernia@cs.washington.edu)

Lecture 1 - Intro, Graph Algorithms

3

Resources

- CSEP 521 Course Web Page
 - <http://www.cs.washington.edu/csep521>
- Papers and Sections from Books
- Recommended Algorithms Book
 - Introduction to Algorithms, 2nd Edition by Cormen, Leiserson, Rivest, and Stein
- E-mail list
 - For information from instructors
 - Check web page to sign up
- Message Board
 - For discussion

Lecture 1 - Intro, Graph Algorithms

4

Engagement by Students

- Weekly Assignments
 - Algorithm design and evaluation
 - Algorithm animation
- In-class activities
- Project with a written report
 - Evaluate several alternative approaches to algorithmically solve a problem
 - Must include readings from literature
 - May include an implementation study
 - May be done in small teams

Lecture 1 - Intro, Graph Algorithms

5

Final Exam and Grading

- There will be no Final Exam
- Percentages
 - Weekly Assignments 60%
 - Project 40%

Lecture 1 - Intro, Graph Algorithms

6

Some Topics

- Graph Algorithms
- Maximum Flow
- Linear Programming
- Data Compression
- Computational Geometry
- Computational Biology

Lecture 1 - Intro, Graph Algorithms

7

Along the Way

- Analysis of algorithms
- Data structures
- NP-completeness
- Dynamic programming
- Greedy algorithms
- Branch-and-bound algorithms
- Approximation algorithms
- Classics of algorithms

Lecture 1 - Intro, Graph Algorithms

8

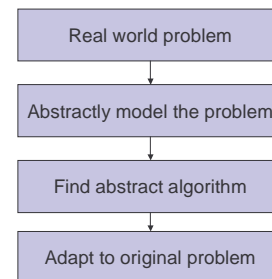
Reading

- Chapter 21 - Disjoint Union / Find
- Chapter 22 - Graph algorithms
- Chapter 23 - Minimum Spanning Tree
- Chapter 24 - Shortest Paths

Lecture 1 - Intro, Graph Algorithms

9

Applied Algorithm Scenario



Lecture 1 - Intro, Graph Algorithms

10

Modeling

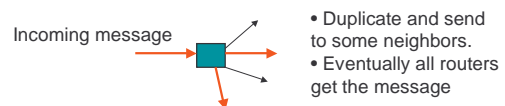
- What kind of algorithm is needed
 - Sorting or Searching
 - Graph Problem
 - Linear Programming
 - Dynamic Programming
 - Clustering
 - Algebra
- Can I find an algorithm or do I have to invent one

Lecture 1 - Intro, Graph Algorithms

11

Broadcasting in a Network

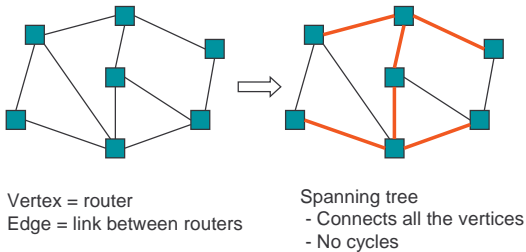
- Network of Routers
 - Organize the routers to efficiently broadcast messages to each other



Lecture 1 - Intro, Graph Algorithms

12

Spanning Tree in a Graph

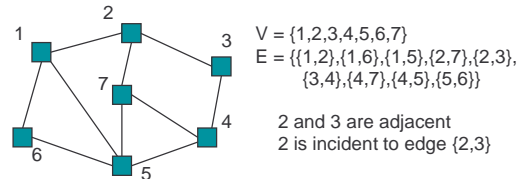


Lecture 1 - Intro, Graph Algorithms

13

Undirected Graph

- $G = (V, E)$
 - V is a set of vertices (or nodes)
 - E is a set of unordered pairs of vertices



Lecture 1 - Intro, Graph Algorithms

14

Spanning Tree Problem

- Input: An undirected graph $G = (V, E)$. G is connected.
- Output: T contained in E such that
 - (V, T) is a connected graph
 - (V, T) has no cycles

Lecture 1 - Intro, Graph Algorithms

15

Depth First Search Algorithm

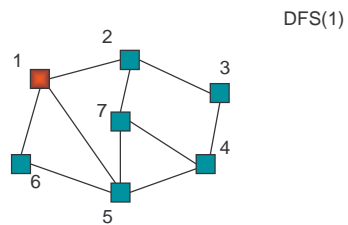
- Recursive marking algorithm
- Initially every vertex is unmarked

```
DFS(i: vertex)
  mark i;
  for each j adjacent to i do
    if j is unmarked then DFS(j)
  end{DFS}
```

Lecture 1 - Intro, Graph Algorithms

16

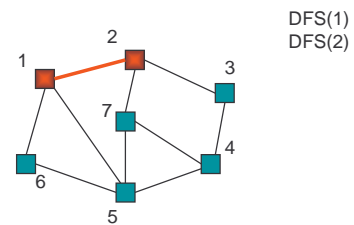
Example of Depth First Search



Lecture 1 - Intro, Graph Algorithms

17

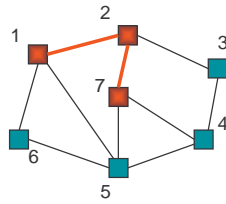
Example Step 2



Lecture 1 - Intro, Graph Algorithms

18

Example Step 3

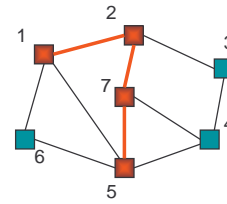


DFS(1)
DFS(2)
DFS(7)

Lecture 1 - Intro, Graph Algorithms

19

Example Step 4

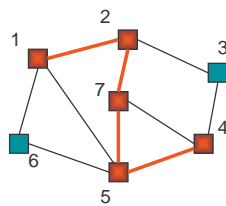


DFS(1)
DFS(2)
DFS(7)
DFS(5)

Lecture 1 - Intro, Graph Algorithms

20

Example Step 5

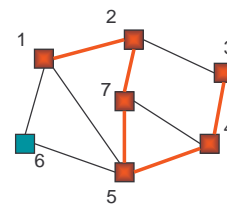


DFS(1)
DFS(2)
DFS(7)
DFS(5)
DFS(4)

Lecture 1 - Intro, Graph Algorithms

21

Example Step 6

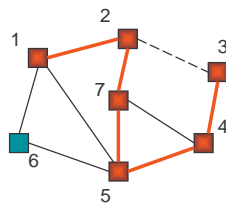


DFS(1)
DFS(2)
DFS(7)
DFS(5)
DFS(4)
DFS(3)

Lecture 1 - Intro, Graph Algorithms

22

Example Step 7

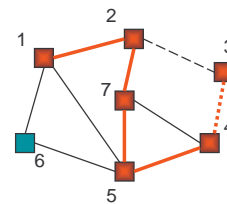


DFS(1)
DFS(2)
DFS(7)
DFS(5)
DFS(4)
DFS(3)

Lecture 1 - Intro, Graph Algorithms

23

Example Step 8

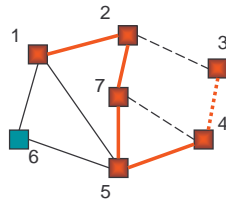


DFS(1)
DFS(2)
DFS(7)
DFS(5)
DFS(4)

Lecture 1 - Intro, Graph Algorithms

24

Example Step 9

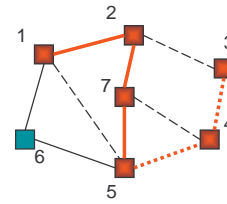


DFS(1)
DFS(2)
DFS(7)
DFS(5)
DFS(4)

Lecture 1 - Intro, Graph Algorithms

25

Example Step 10

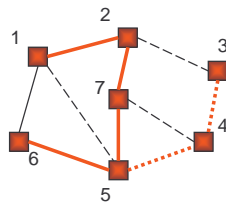


DFS(1)
DFS(2)
DFS(7)
DFS(5)

Lecture 1 - Intro, Graph Algorithms

26

Example Step 11

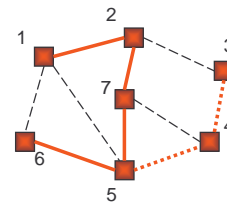


DFS(1)
DFS(2)
DFS(7)
DFS(5)
DFS(6)

Lecture 1 - Intro, Graph Algorithms

27

Example Step 12

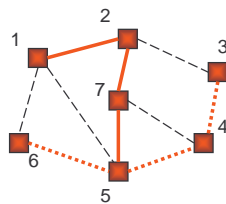


DFS(1)
DFS(2)
DFS(7)
DFS(5)
DFS(6)

Lecture 1 - Intro, Graph Algorithms

28

Example Step 13

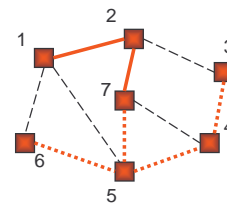


DFS(1)
DFS(2)
DFS(7)
DFS(5)

Lecture 1 - Intro, Graph Algorithms

29

Example Step 14

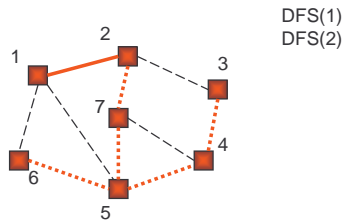


DFS(1)
DFS(2)
DFS(7)

Lecture 1 - Intro, Graph Algorithms

30

Example Step 15

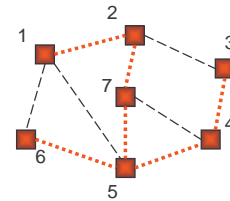


DFS(1)
DFS(2)

Lecture 1 - Intro, Graph Algorithms

31

Example Step 16



DFS(1)

Lecture 1 - Intro, Graph Algorithms

32

Spanning Tree Algorithm

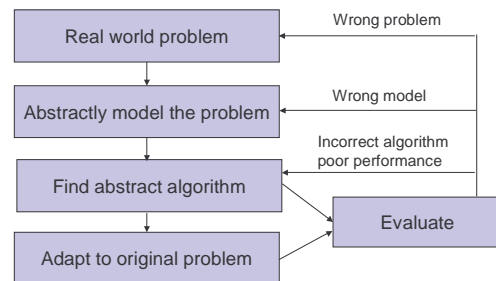
```
ST(i: vertex)
  mark i;
  for each j adjacent to i do
    if j is unmarked then
      Add {i,j} to T;
      ST(j);
  end{ST}
```

```
Main
  T := empty set;
  ST(1);
  end{Main}
```

Lecture 1 - Intro, Graph Algorithms

33

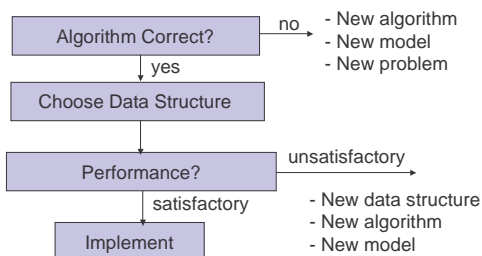
Applied Algorithm Scenario



Lecture 1 - Intro, Graph Algorithms

34

Evaluation Step Expanded



Lecture 1 - Intro, Graph Algorithms

35

Correctness of ST Algorithm

- There are no cycles in T
 - This is an invariant of the algorithm.
 - Each edge added to T goes from a vertex in T to a vertex not in T.
- If G is connected then eventually every vertex is marked. (Proof by contradiction)

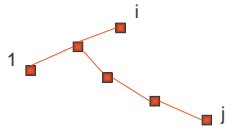


Lecture 1 - Intro, Graph Algorithms

36

Correctness (cont.)

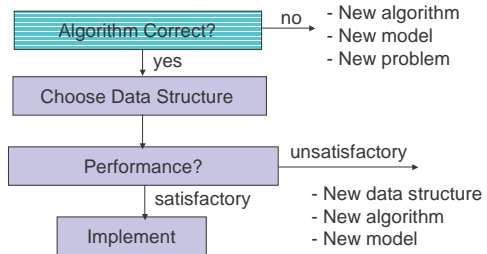
- If G is connected then so is (V, T)



Lecture 1 - Intro, Graph Algorithms

37

Data Structure Step



Lecture 1 - Intro, Graph Algorithms

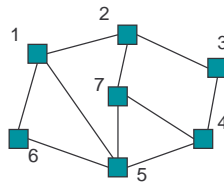
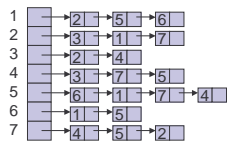
38

Edge List and Adjacency Lists

- List of edges

1	5	1	2	2	3	5	7	5	5
2	1	6	7	3	4	6	4	7	4

- Adjacency lists

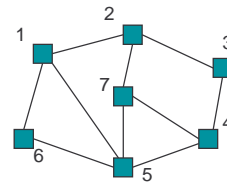


Lecture 1 - Intro, Graph Algorithms

39

Adjacency Matrix

	1	2	3	4	5	6	7
1	0	1	0	0	1	1	0
2	1	0	1	0	0	0	1
3	0	1	0	1	0	0	0
4	0	0	1	0	1	0	1
5	1	0	0	1	0	1	1
6	1	0	0	0	1	0	0
7	0	1	0	1	1	0	0



Lecture 1 - Intro, Graph Algorithms

40

Data Structure Choice

- Edge list
 - Simple but does not support depth first search
- Adjacency lists
 - Good for sparse graphs
 - Supports depth first search
- Adjacency matrix
 - Good for dense graphs
 - Supports depth first search

Lecture 1 - Intro, Graph Algorithms

41

Spanning Tree with Adjacency Lists

```

ST(i: vertex)
  M[i] := 1;
  v := G[i];
  while not(v = null)
    j := v.vertex;
    if M[j] = 0 then
      Add (i,j) to T;
      ST(j);
    v := v.next;
  end{ST}
    
```

```

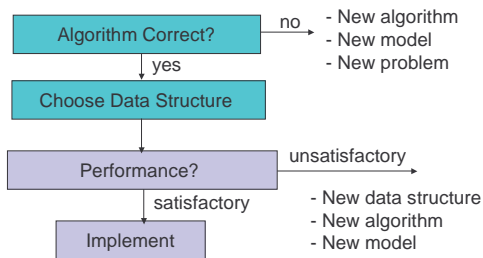
Main
  G is array of adjacency lists;
  M[i] := 0 for all i;
  T is empty;
  Spanning_Tree(1);
end{Main}
    
```

M is the marking array
Node of linked list
vertex next

Lecture 1 - Intro, Graph Algorithms

42

Performance Step



Lecture 1 - Intro, Graph Algorithms

43

Performance of ST Algorithm

- n vertices and m edges
- Connected graph
- Storage complexity $O(m)$
- Time complexity $O(m)$

Lecture 1 - Intro, Graph Algorithms

44

Other Uses of Depth First Search

- Popularized by Hopcroft and Tarjan 1973
- Connected components
- Biconnected components
- Strongly connected components in directed graphs
- topological sorting of a acyclic directed graphs

Lecture 1 - Intro, Graph Algorithms

45

Depth-First Search in Directed Graphs

- Discovery and Finish Times
- Initially $D[i] = F[i] = 0$, $\text{time} = 1$

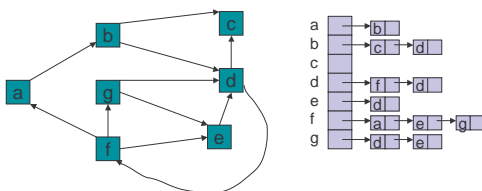
```

DFS(i: vertex)
  D[i] := time;
  time++;
  v := G[i];
  for each vertex j adjacent to i do
    if D[j] = 0 then DFS(j)
  F[i] := time;
  time++;
end{DFS}
  
```

Lecture 1 - Intro, Graph Algorithms

46

Example



- Compute the discovery and finish times
- Classify the edges

Lecture 1 - Intro, Graph Algorithms

47

Edge Classification

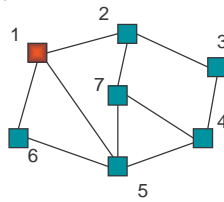
- Forward Edge (i, j)
 - $D[i] < D[j] < F[j] < F[i]$
- Backward Edge
 - $D[j] < D[i] < F[i] < F[j]$
- Cross Edge
 - $D[j] < F[j] < D[i] < F[i]$
- Note – A directed graph is acyclic if and only if it has no backward edges in a DFS.

Lecture 1 - Intro, Graph Algorithms

48

ST using Breadth First Search 1

- Uses a queue to order search

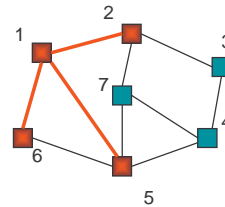


Queue = 1

Lecture 1 - Intro, Graph Algorithms

49

Breadth First Search 2

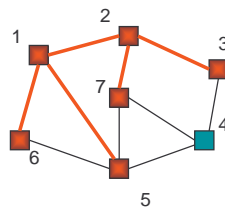


Queue = 2,6,5

Lecture 1 - Intro, Graph Algorithms

50

Breadth First Search 3

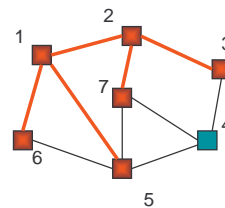


Queue = 6,5,7,3

Lecture 1 - Intro, Graph Algorithms

51

Breadth First Search 4

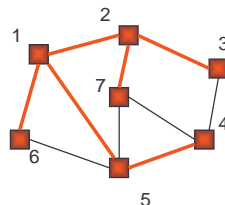


Queue = 5,7,3

Lecture 1 - Intro, Graph Algorithms

52

Breadth First Search 5

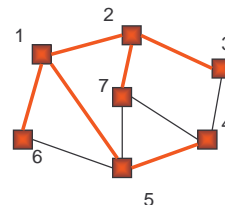


Queue = 7,3,4

Lecture 1 - Intro, Graph Algorithms

53

Breadth First Search 6

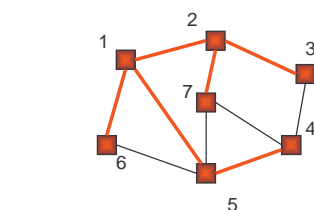


Queue = 3,4

Lecture 1 - Intro, Graph Algorithms

54

Breadth First Search 7

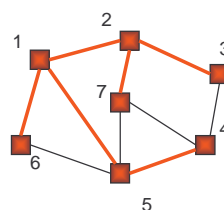


Queue = 4

Lecture 1 - Intro, Graph Algorithms

55

Breadth First Search 8



Queue =

Lecture 1 - Intro, Graph Algorithms

56

Spanning Tree using Breadth First Search

```
BFS
Initialize T to be empty;
Initialize Q to be empty;
Enqueue(1,Q) and mark 1;
while Q is not empty do
  i := Dequeue(Q);
  for each j adjacent to i do
    if j is not marked then
      add {i,j} to T;
      Enqueue(j,Q) and mark j;
end(BFS)
```

Lecture 1 - Intro, Graph Algorithms

57

Depth First vs Breadth First

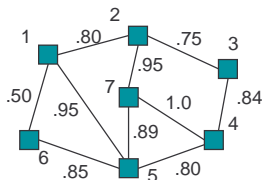
- Depth First
 - Stack or recursion
 - Many applications
- Breadth First
 - Queue (recursion no help)
 - Can be used to find shortest paths from the start vertex

Lecture 1 - Intro, Graph Algorithms

58

Best Spanning Tree

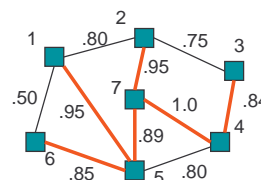
- Each edge has the probability that it won't fail
- Find the spanning tree that is least likely to fail



Lecture 1 - Intro, Graph Algorithms

59

Example of a Spanning Tree



Probability of success = $.85 \times .95 \times .89 \times .95 \times 1.0 \times .84$
= .5735

Lecture 1 - Intro, Graph Algorithms

60

Minimum Spanning Tree Problem

- Input: Undirected Graph $G = (V, E)$ and a cost function C from E to the reals. $C(e)$ is the cost of edge e .
- Output: A spanning tree T with minimum total cost. That is: T that minimizes

$$C(T) = \sum_{e \in T} C(e)$$

Lecture 1 - Intro, Graph Algorithms

61

Reducing Best to Minimum

Let $P(e)$ be the probability that an edge doesn't fail.

Define:

$$C(e) = -\log_{10}(P(e))$$

Minimizing $\sum_{e \in T} C(e)$

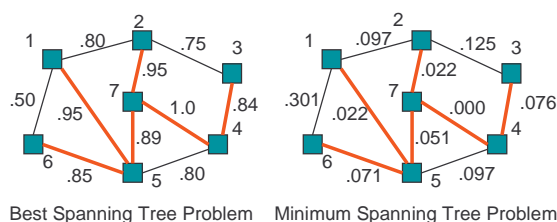
is equivalent to maximizing $\prod_{e \in T} P(e)$

because $\prod_{e \in T} P(e) = 10^{-\sum_{e \in T} C(e)}$

Lecture 1 - Intro, Graph Algorithms

62

Example of Reduction



Best Spanning Tree Problem Minimum Spanning Tree Problem

Lecture 1 - Intro, Graph Algorithms

63

Minimum Spanning Tree

- Boruvka 1926
- Kruskal 1956
- Prim 1957 also by Jarnik 1930
- Karger, Klein, Tarjan 1995
 - Randomized linear time algorithm
 - Probably not practical, but very interesting

Lecture 1 - Intro, Graph Algorithms

64

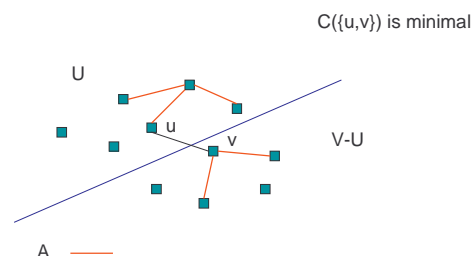
MST Optimality Principle

- $G = (V, E)$ with costs C . G connected.
- Let (V, A) be a subgraph of G that is contained in a minimum spanning tree. Let U be a set such that no edge in A has one end in U and one end in $V-U$. Let $C(\{u, v\})$ minimal and u in U and v in $V-U$. Let A' be A with $\{u, v\}$ added. Then (V, A') is contained in a minimum spanning tree.

Lecture 1 - Intro, Graph Algorithms

65

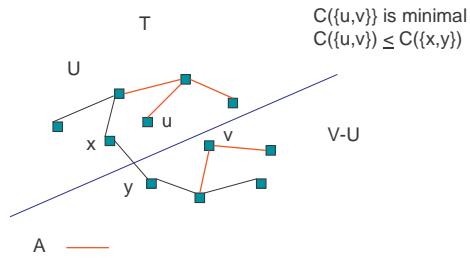
Proof of Optimality Principle



Lecture 1 - Intro, Graph Algorithms

66

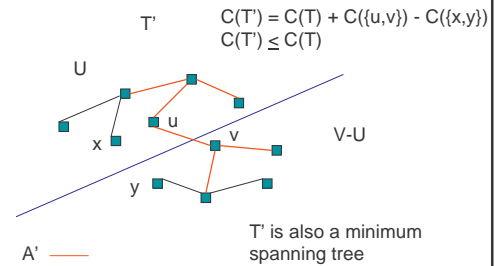
Proof of Optimality Principle



Lecture 1 - Intro, Graph Algorithms

67

Proof of Optimality Principle



Lecture 1 - Intro, Graph Algorithms

68

Kruskal's Greedy Algorithm

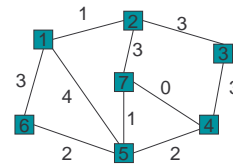
Sort the edges by increasing cost;
 Initialize A to be empty;
 For each edge e chosen in increasing order do
 if adding e does not form a cycle then
 add e to A

Invariant: A is always contained in some minimum spanning tree

Lecture 1 - Intro, Graph Algorithms

69

Example of Kruskal 1

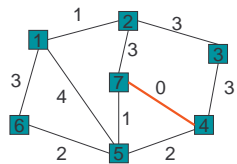


$\{7,4\}$ $\{2,1\}$ $\{7,5\}$ $\{5,6\}$ $\{5,4\}$ $\{1,6\}$ $\{2,7\}$ $\{2,3\}$ $\{3,4\}$ $\{1,5\}$
 0 1 1 2 2 3 3 3 3 4

Lecture 1 - Intro, Graph Algorithms

70

Example of Kruskal 2

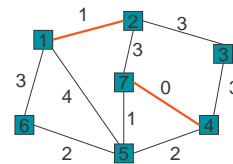


~~$\{7,4\}$~~ $\{2,1\}$ $\{7,5\}$ $\{5,6\}$ $\{5,4\}$ $\{1,6\}$ $\{2,7\}$ $\{2,3\}$ $\{3,4\}$ $\{1,5\}$
 0 1 1 2 2 3 3 3 3 4

Lecture 1 - Intro, Graph Algorithms

71

Example of Kruskal 2

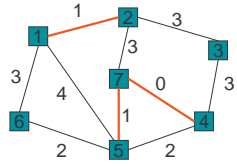


~~$\{7,4\}$~~ ~~$\{2,1\}$~~ $\{7,5\}$ $\{5,6\}$ $\{5,4\}$ $\{1,6\}$ $\{2,7\}$ $\{2,3\}$ $\{3,4\}$ $\{1,5\}$
 0 1 1 2 2 3 3 3 3 4

Lecture 1 - Intro, Graph Algorithms

72

Example of Kruskal 3



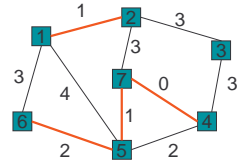
~~{7,4}~~ ~~{2,1}~~ ~~{7,5}~~ {5,6} {5,4} {1,6} {2,7} {2,3} {3,4} {1,5}

0 1 1 2 2 3 3 3 3 4

Lecture 1 - Intro, Graph Algorithms

73

Example of Kruskal 4



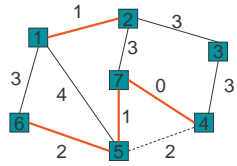
~~{7,4}~~ ~~{2,1}~~ ~~{7,5}~~ {5,6} {5,4} {1,6} {2,7} {2,3} {3,4} {1,5}

0 1 1 2 2 3 3 3 3 4

Lecture 1 - Intro, Graph Algorithms

74

Example of Kruskal 5



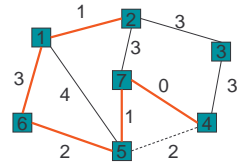
~~{7,4}~~ ~~{2,1}~~ ~~{7,5}~~ {5,6} {5,4} {1,6} {2,7} {2,3} {3,4} {1,5}

0 1 1 2 2 3 3 3 3 4

Lecture 1 - Intro, Graph Algorithms

75

Example of Kruskal 6



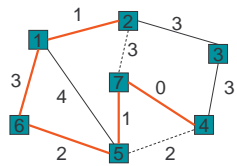
~~{7,4}~~ ~~{2,1}~~ ~~{7,5}~~ {5,6} {5,4} {1,6} {2,7} {2,3} {3,4} {1,5}

0 1 1 2 2 3 3 3 3 4

Lecture 1 - Intro, Graph Algorithms

76

Example of Kruskal 7



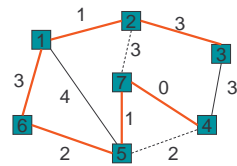
~~{7,4}~~ ~~{2,1}~~ ~~{7,5}~~ {5,6} {5,4} {1,6} {2,7} {2,3} {3,4} {1,5}

0 1 1 2 2 3 3 3 3 4

Lecture 1 - Intro, Graph Algorithms

77

Example of Kruskal 7



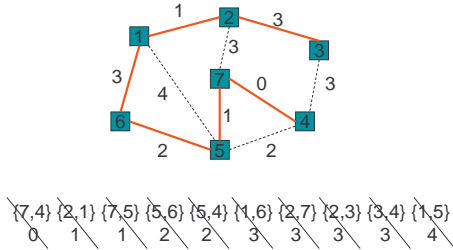
~~{7,4}~~ ~~{2,1}~~ ~~{7,5}~~ {5,6} {5,4} {1,6} {2,7} {2,3} {3,4} {1,5}

0 1 1 2 2 3 3 3 3 4

Lecture 1 - Intro, Graph Algorithms

78

Example of Kruskal 8,9



Lecture 1 - Intro, Graph Algorithms

79

Data Structures for Kruskal

- Sorted edge list

$\{7,4\} \{2,1\} \{7,5\} \{5,6\} \{5,4\} \{1,6\} \{2,7\} \{2,3\} \{3,4\} \{1,5\}$
0 1 1 2 2 3 3 3 3 4

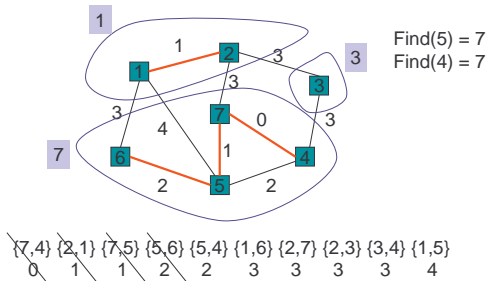
- Disjoint Union / Find

- Union(a,b) - union the disjoint sets named by a and b
- Find(a) returns the name of the set containing a

Lecture 1 - Intro, Graph Algorithms

80

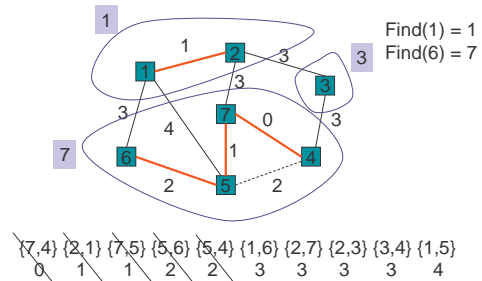
Example of DU/F 1



Lecture 1 - Intro, Graph Algorithms

81

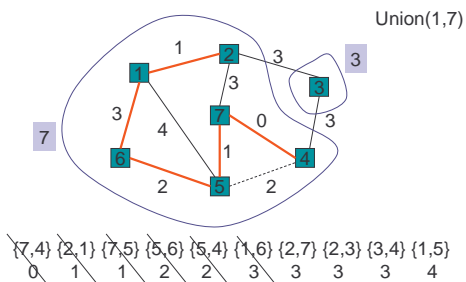
Example of DU/F 2



Lecture 1 - Intro, Graph Algorithms

82

Example of DU/F 3



Lecture 1 - Intro, Graph Algorithms

83

Kruskal's Algorithm with DU / F

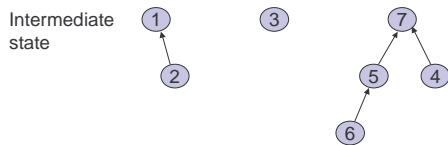
```
Sort the edges by increasing cost;
Initialize A to be empty;
for each edge {i,j} chosen in increasing order do
  u := Find(i);
  v := Find(j);
  if not(u = v) then
    add {i,j} to A;
    Union(u,v);
```

Lecture 1 - Intro, Graph Algorithms

84

Up Tree for DU/F

Initial state 1 2 3 4 5 6 7

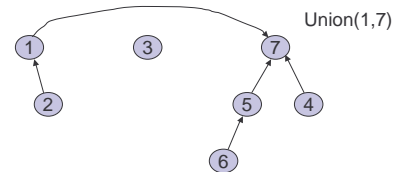


Lecture 1 - Intro, Graph Algorithms

85

DU/F Operation

- Find(i) - follow pointer to root and return the root.
- Union(i,j) - assuming i and j roots, point i to j.

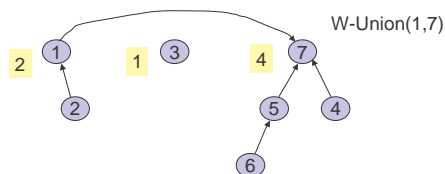


Lecture 1 - Intro, Graph Algorithms

86

Weighted Union

- Weighted Union
 - Always point the smaller tree to the root of the larger tree

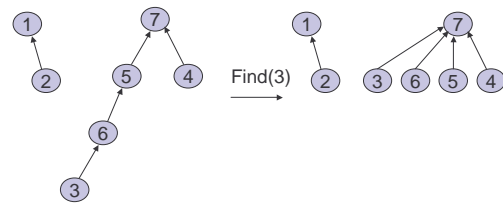


Lecture 1 - Intro, Graph Algorithms

87

Path Compression

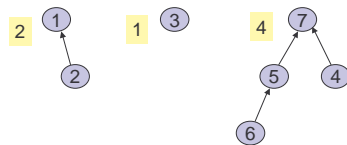
- On a Find operation point all the nodes on the search path directly to the root.



Lecture 1 - Intro, Graph Algorithms

88

Elegant Array Implementation



	1	2	3	4	5	6	7
up	0	1	0	7	7	5	0
weight	2		1				4

Lecture 1 - Intro, Graph Algorithms

89

Up Tree Pseudo-Code

```

PC-Find(i : index)
  r := i;
  while not(up[r] = 0) do
    r := up[r];
  k := up[i];
  while not(k = r) do
    up[i] := r;
    i := k;
    k := up[k];
  return(r);
end{Find}
    
```

```

W-Union(i,j : index)
  // i and j are roots
  wi := weight[i];
  wj := weight[j];
  if wi < wj then
    up[i] := j;
    weight[j] := wi + wj;
  else
    up[j] := i;
    weight[i] := wi + wj;
  end{W-Union}
    
```

Lecture 1 - Intro, Graph Algorithms

90

Disjoint Union / Find Notes

- Worst case time complexity for a W-Union is $O(1)$ and for a PC-Find is $O(\log n)$.
- Time complexity for m operations on n elements is $O(m \log^* n)$ where $\log^* n$ is a very slow growing function. Essentially constant time per operation!
- Using “ranked union” gives an even better bound theoretically.

Lecture 1 - Intro, Graph Algorithms

91

Performance of W-Union / PC-Find

- The time complexity of PC-Find is $O(\log n)$.
- An up tree formed by W-Union of height h has at least 2^h nodes. Inductive Proof.

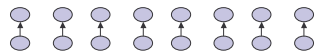


Lecture 1 - Intro, Graph Algorithms

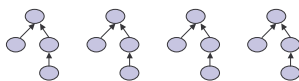
92

Worst Case for PC-Find

$n/2$ Weighted Unions



$n/4$ Weighted Unions

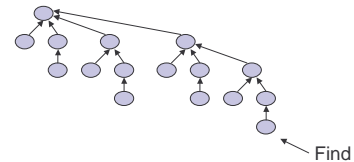


Lecture 1 - Intro, Graph Algorithms

93

Example of Worst Cast (cont')

After $n - 1 = n/2 + n/4 + \dots + 1$ Weighted Unions



If there are $n = 2^k$ nodes then there are k pointers on the longest path to root.

Lecture 1 - Intro, Graph Algorithms

94

Amortized Complexity

- For disjoint union / find with weighted union and path compression.
 - average time per operation is essentially a constant.
 - worst case time for a PC-Find is $O(\log n)$.
- An individual operation can be costly, but over time the average cost per operation is not.

Lecture 1 - Intro, Graph Algorithms

95