Lecture 13 Algorithms & Data Structures

Goldsmiths Computing

January 28, 2019

Outline

Introduction

Graphs

Spanning trees

Path finding

Outline

Introduction

- 1. Implicit data structures (...cont'd)
 - Heaps

VLE activities

Binary search quiz

Statistics so far:

- · A attempts: average mark B
- C students: average mark D
 - E under 4.00, F over 6.99, G at 10.00

Quiz closes at 16:00 on Friday 1st February

- · no extensions
- · grade is
 - 0 (for no attempt)
 - $30 + 70 \times (\text{score}/10)^2$

Hashing quiz

Hashing quiz

Hashing quiz

Binary search submission

Introduction

Graphs

Spanning trees

Path finding

Motivation

Many, many problems can be expressed in terms of graphs.

Definition

A graph is a set of vertices (nodes) which are linked by zero or more edges from nodes to nodes, each of which can have a weight.

Operations

```
vertices return the collection of vertices {v} in the graph
edges return the collection of edges {(u,v)} in the graph
addVertex[v] add vertex v to the graph
addEdge[e] add edge e to the graph
neighbours[v] return the collection of vertices directly reachable from v
weight[u,v] return the weight of the edge between u and v
constructor(V, E) make a new graph with with given vertex and edge
collection
```

Edge operations

from return the source vertex of this edge
to return the destination vertex of this edge
weight return the weight of this edge

Representations

adjacency matrix a matrix of edge information linking vertices adjacency list an array of vertices, each vertex contaning edges from that vertex

edge list a list of edges in the graph, along with a set of vertices

Other definitions

Directed

A directed graph has edges that are one-directional: an edge from u to v does not imply an edge from v to u.

Undirected

An undirected graph has two-directional edges: an edge from u to v with weight w implies an edge from v to u with weight w.

Tree

A tree (in the context of graphs) is an undirected graph where any two vertices are joined by exactly one path.

Directed acyclic

A directed acyclic graph or DAG is a directed graph where no sequence of edges returns to its starting point.

1. Reading:

- · CLRS, chapter 22
- · Drozdek, section 8.1
- · DPV, section 3.1

Introduction

Graphs

Spanning trees

Path finding

Motivation

- Internet routing
- · Electricity, cable, road networks
- Maze generation

Definition

A tree T is a spanning tree of a graph G if it is: a subgraph of G includes only edges that are present in G; and spans G includes all vertices of G

Properties

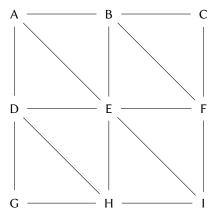
If G has |V| vertices, a spanning tree has:

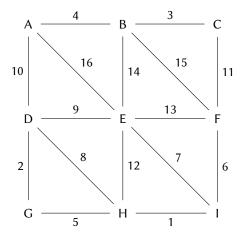
- |V| vertices
- |V|-1 edges (proof?)

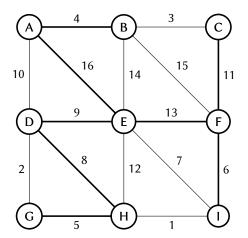
A B C

D E F

G H I







Minimum spanning tree

A minimum spanning tree for graph G is a spanning tree of graph G whose edge weights sum to the minimum possible total weight for that graph.

Prim's algorithm

```
function PRIMMST(G) vs \leftarrow vertices(G) \\ T \leftarrow new \ Graph(first(vs), \{\!\!\!\}) \\ while \ |T| < |G| \ do \\ E \leftarrow \{e \mid e \in edges(G) \land \\ from(e) \in vertices(T) \land to(e) \notin vertices(T) \} \\ newE \leftarrow argmin_{e \in E} \ weight(e) \\ newV \leftarrow to(newE) \\ AddVertex(T,newV); \ AddEdges(T,newE) \\ end \ while \\ end \ function
```

Prim's algorithm

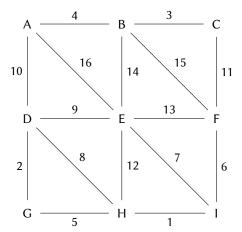
```
function PRIMMST(G) vs \leftarrow vertices(G) \\ T \leftarrow new \ Graph(first(vs),\{\}) \\ while |T| < |G| \ do \\ E \leftarrow \{e \mid e \in edges(G) \land \\ from(e) \in vertices(T) \land to(e) \notin vertices(T)\} \\ newE \leftarrow argmin_{e \in E} \ weight(e) \\ newV \leftarrow to(newE) \\ Add Vertex(T,newV); \ Add Edges(T,newE) \\ end \ while \\ end \ function \\
```

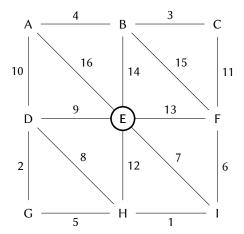
- 1. Initialise the minimum spanning tree with a vertex from the graph.
- 2. Until all the vertices are included in the tree,
 - find the edge with smallest weight that links a vertex in the tree so far with a vertex not yet in the tree;
 - · add that edge, and the new vertex, to the minimum spanning tree.

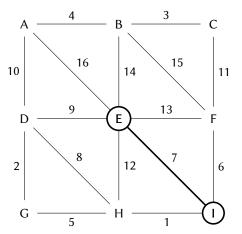


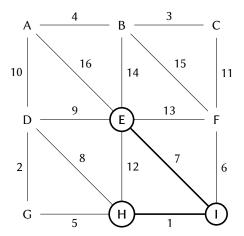
Prim's algorithm

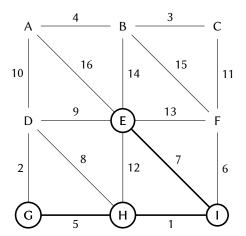
```
function PRIMMST(G)
    vs \leftarrow vertices(G)
    T \leftarrow new Graph(FIRST(vs),{})
    while |T| < |G| do
        newE \leftarrow NIL; newV \leftarrow NIL; w \leftarrow \infty
        for e \in EDGES(G) \land FROM(e) \in VERTICES(T) \land TO(e) \notin VERTICES(T) do
             if WEIGHT(e) < w then
                 w \leftarrow weight(e); newE \leftarrow e; newV \leftarrow to(e)
             end if
        end for
        ADDVERTEX(T,newV); ADDEDGE(T,newE)
    end while
end function
```

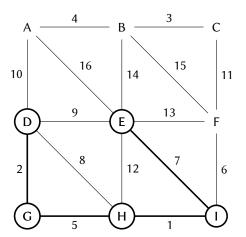


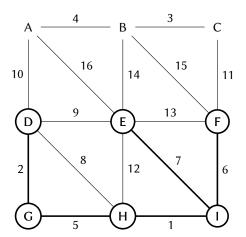




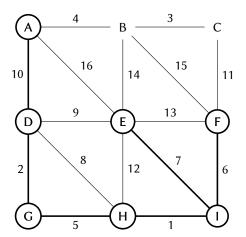




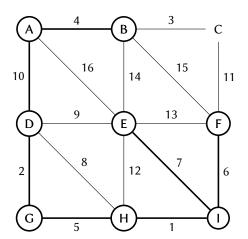




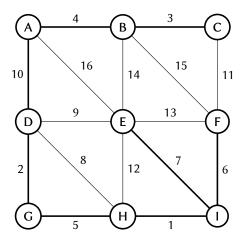
Prim's algorithm: example



Prim's algorithm: example



Prim's algorithm: example



Prim's algorithm: proof sketch

By contradiction. Let P be the spanning tree generated by Prim's algorithm on G, and T be the minimum spanning tree.

- if P = T, we are done.
- if P ≠ T,
 - there is an edge e in P not in T;
 - we added that edge to P at some point, joining a set of vertices V in Prim's tree with one of the set of vertices G-V;
 - find the edge f in T that joins V with G-V;
 - the tree T-f+e must have lower cost than T (why?) and is a spanning tree (why?).

Kruskal's algorithm

```
function KruskalmST(G)

vs \leftarrow vertices(G)

T \leftarrow new Graph(vs,\{\})

Z \leftarrow new DisjointSet()

for v \in vs do

Make-set(Z,v)

end for

for (u,v) \in edges(G) sorted by weight do

if find(Z,u) \neq find(Z,v) then

AddEdge(T,(u,v))

Union(Z,u,v)

end if

end for

end function
```

Kruskal's algorithm

```
function KruskalMST(G)

vs ← vertices(G)

T ← new Graph(vs,{})

Z ← new DisjointSet()

for v ∈ vs do

MAKE-SET(Z,v)

end for

for (u,v) ∈ edges(G) sorted by weight do

if find(Z,u) ≠ find(Z,v) then

AddEdge(T,(u,v))

Union(Z,u,v)

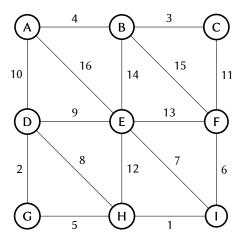
end if

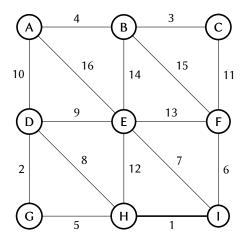
end for

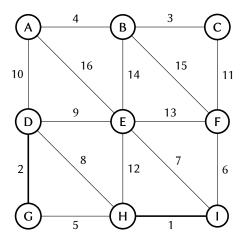
end function
```

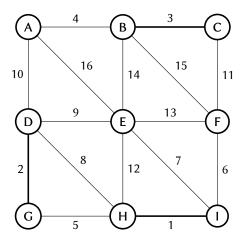
- 1. Initialise the minimum spanning tree with all vertices and no edges.
- 2. Put each vertex into its own equivalence class
- 3. Iterate over all the edges in the graph, ordered by weight:
 - · add the edge to the tree if it joins two different equivalence classes;
 - · make the edge's two vertices' equivalence classes be the same

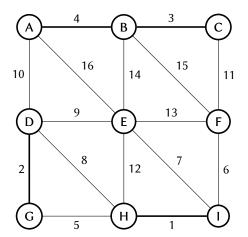


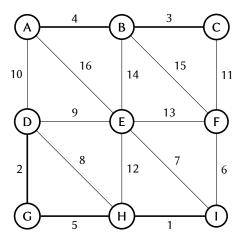


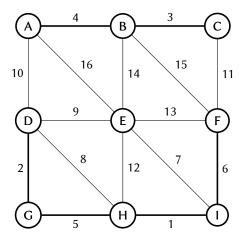


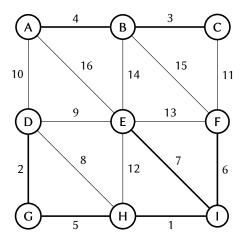


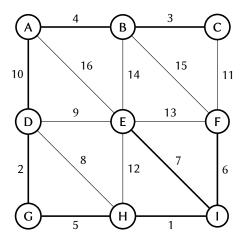












Kruskal's algorithm: proof

Similar to Prim's algorithm: consider the first edge added in the spanning tree K that is not in T

Work

- 1. Reading:
 - · CLRS, chapter 23
 - · Drozdek, section 8.5
 - DPV, section 5.1
- 2. Questions:

```
CLRS Exercises 23.1-7, 23.2-1, 23.2-2, 23.2-4, 23.2-5
CLRS Problem 23-1
DPV Exercises 5.1, 5.2, 5.5
```

- 3. Complete the proof of Kruskal's algorithm.
- 4. Choose a concrete representation for graphs and edge sets. For your choices, what is the time complexity of Prim's algorithm?

Outline

Introduction

Graphs

Spanning trees

Path finding



Path finding

Motivation

- · Exploration of known, partially known or unknown surroundings
- Component of various AI solutions
 - especially agents exploring some space:
 - ... game enemies
 - ... NPCs
 - · ... self-driving cars

Definition

Single-source shortest path

- from a single source node:
 - find the shortest path to every node in the graph
 - · stop early if we have a specific target node

"Begin at the beginning," the King said gravely, "and go on till you come to the end: then stop."



"Begin at the beginning," the King said gravely, "and go on till you come to the end: then stop."

Initial state Start at the start node

"Begin at the beginning," the King said gravely, "and go on till you come to the end: then stop."

Initial state

Start at the start node

Goal state

Stop when we have found a path (optimally: shortest) to the target node

• (if no target: stop when there are no nodes without the shortest path known)

"Begin at the beginning," the King said gravely, "and go on till you come to the end: then stop."

Initial state

Start at the start node

Goal state

Stop when we have found a path (optimally: shortest) to the target node

 (if no target: stop when there are no nodes without the shortest path known)

State expansion

Explore the graph using neighbours of already-visited nodes

Greedy best-first search

Explore the graph using the neighbour of the current node which is closest to the target.

```
function GBFS(G,start,end)

current \leftarrow start

result \leftarrow new queue()

while current \neq end do

ENQUEUE(result,current)

ns \leftarrow NEIGHBOURS(G,current)

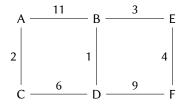
current \leftarrow arg min_{n \in ns} d(n, end)

end while

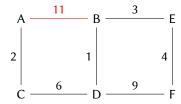
return result

end function
```

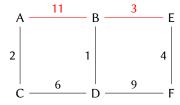
Node	d(n,F)
A	14
В	6
C	12
D	7
E	4
F	0



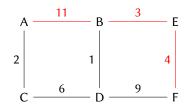
Node	d(n,F)
Α	14
В	6
С	12
D	7
E	4
F	0



Node	d(n,F)
Α	14
В	6
C	12
D	7
E	4
F	0



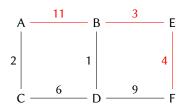
Node	d(n,F)
Α	14
В	6
C	12
D	7
Ε	4
F	0



Result

path
$$A \rightarrow B \rightarrow E \rightarrow F$$
 distance 18

Node	d(n,F)
A	14
В	6
C	12
D	7
Ε	4
F	0



Result

path
$$A \rightarrow B \rightarrow E \rightarrow F$$
 distance 18

Problems

- · does not necessarily find a solution!
- · not guaranteed optimal

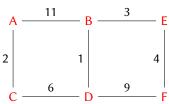
Dijkstra's algorithm

Explore the graph using the neighbour of the already-visited nodes with the smallest distance from the start node

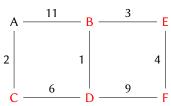
```
function DIJKSTRA(G,start,end)
    dist ← new table(); prev ← new table()
    Q ← new min-heap(dist)
    for v \in G do
        dist[v] \leftarrow 0 if v = start else \infty; INSERT(Q,v)
    end for
    while ¬ EMPTY(Q) do
        u \leftarrow \text{EXTRACT-MIN}(Q)
        if u = end then
            s ← new stack()
            while u # start do
                PUSH(s,u); u \leftarrow prev[u]
            end while
            return s
        end if
        for v \in \text{NEIGHBOURS}(G,u) do
            d \leftarrow dist[u] + weight(G,u,v)
            if d < dist[v] then
                 dist[v] \leftarrow d; prev[v] \leftarrow u; dist[v] \leftarrow u; dist[v] \leftarrow u
            end if
        end for
    end while
end function
```

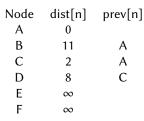
Node dist[n] prev[n]
A 0

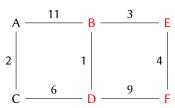
 $\begin{array}{ccc} B & \infty \\ C & \infty \\ D & \infty \\ E & \infty \\ F & \infty \end{array}$



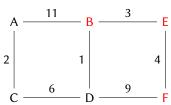
Node	dist[n]	prev[n]
Α	0	
В	11	Α
C	2	Α
D	∞	
Ε	∞	
F	∞	



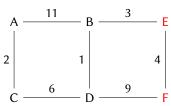




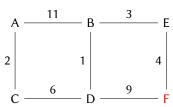
Node	dist[n]	prev[n]
Α	0	
В	9	D
C	2	Α
D	8	С
Ε	∞	
F	17	D



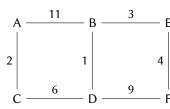
Node	dist[n]	prev[n
Α	0	
В	9	D
C	2	Α
D	8	C
Ε	12	В
F	17	D



Node	dist[n]	prev[n
Α	0	
В	9	D
C	2	Α
D	8	C
Ε	12	В
F	16	Ε



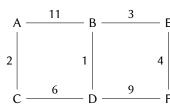
Node	dist[n]	prev[n]
Α	0	
В	9	D
C	2	Α
D	8	С
Ε	12	В
F	16	Е



Result

path
$$A \rightarrow C \rightarrow D \rightarrow B \rightarrow E \rightarrow F$$
 distance 16

Node	dist[n]	prev[n]
Α	0	
В	9	D
C	2	Α
D	8	C
Ε	12	В
F	16	Ε



Result

path
$$A \rightarrow C \rightarrow D \rightarrow B \rightarrow E \rightarrow F$$

distance 16

Note

- · requires all non-negative weights
- · guaranteed to find shortest path
- · need priority queue (min-heap) for efficient operation
- does not use distance estimate information



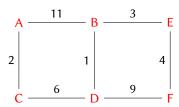


Explore the graph using the neighbour of the already-visited nodes with the smallest estimated distance from the start node to the target node

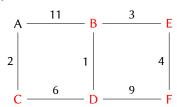
```
function A*(G,start,end)
    dist ← new table(); prev ← new table()
    Q \leftarrow new min-heap(dist)
    for v \in G do
        dist[v] \leftarrow 0 if v = start else \infty; INSERT(Q,v)
    end for
    while ¬ EMPTY(Q) do
        u \leftarrow \text{EXTRACT-MIN}(Q)
        if u = end then
            s ← new stack()
            while u # start do
                 PUSH(s,u); u \leftarrow prev[u]
            end while
            return s
        end if
        for v \in \text{NEIGHBOURS}(G,u) do
            d \leftarrow dist[u] + weight(G,u,v) + H(v)
            if d < dist[v] then
                 dist[v] \leftarrow d; prev[v] \leftarrow u; dist[v] \leftarrow u; dist[v] \leftarrow u
            end if
        end for
    end while
end function
```

Path finding

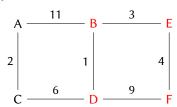
Node	d(n,F)	dist[n]	prev[n]
Α	14	0	
В	6	∞	
C	12	∞	
D	7	∞	
Ε	4	∞	
F	0	∞	



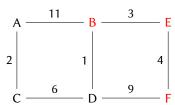
Node	d(n,F)	dist[n]	prev[n]
Α	14	0	
В	6	11	Α
C	12	2	Α
D	7	∞	
E	4	∞	
F	0	m	



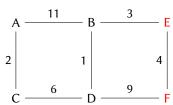
Node	d(n,F)	dist[n]	prev[n]
Α	14	0	
В	6	11	Α
C	12	2	Α
D	7	8	C
Ε	4	∞	
F	0	∞	



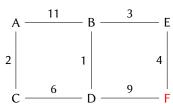
Node	d(n,F)	dist[n]	prev[n
Α	14	0	
В	6	9	D
C	12	2	Α
D	7	8	C
Ε	4	∞	
F	0	17	D



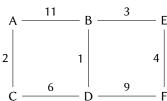
Node	d(n,F)	dist[n]	prev[n]
Α	14	0	
В	6	9	D
C	12	2	Α
D	7	8	C
Ε	4	12	В
F	0	17	D



Node	d(n,F)	dist[n]	prev[n]
Α	14	0	
В	6	9	D
C	12	2	Α
D	7	8	C
Ε	4	12	В
F	0	16	Е



Node	d(n,F)	dist[n]	prev[n]	
Α	14	0		A -
В	6	9	D	
C	12	2	Α	2
D	7	8	C	
Ε	4	12	В	
F	0	16	Ε	C -



Result

path
$$A \rightarrow C \rightarrow D \rightarrow B \rightarrow E \rightarrow F$$

distance 16

Node	d(n,F)	dist[n]	prev[n]		11		3	
Α	14	0		A —		— В —		— E
В	6	9	D					
C	12	2	Α	2		1		4
D	7	8	C					
Е	4	12	В		6		9	_
F	0	16	E	C —		— D —		— r

Result

path
$$A \rightarrow C \rightarrow D \rightarrow B \rightarrow E \rightarrow F$$

distance 16

Note

- · generalisation of Dijkstra's algorithm
- distance estimation н must be admissible
 - · lower bound
 - · non-negative
 - (Dijkstra's algorithm is A* with H(n) = 0)



Work

- 1. Reading
 - CLRS, chapter 24
 - Drozdek, sections 8.2, 8.3
- 2. Questions from CLRS

Exercises 24.3-1