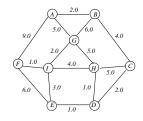
# Weighted Graphs

## ${\sf Definition}\ 1$

A weighted (directed or undirected graph) is a pair  $(\mathcal{G}, W)$  consisting of a graph  $\mathcal{G} = (V, E)$  and a weight function  $W : E \to \mathbb{R}$ .

In this lecture, we always assume that weights are non-negative, i.e., that  $W(e) \ge 0$  for all  $e \in E$ .

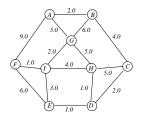
Example



ADS: lects 14 & 15 - slide 1 -

# Representations of Weighted Graphs (as Matrices)

Algorithms and Data Structures: Minimum Spanning Trees I and II - Prim's Algorithm



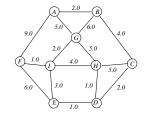
### Adjacency Matrix

( 0	2.0	0	0	0	9.0	5.0	0	0 )
2.0	0	4.0	0	0	0	6.0	0	0
0	4.0	0	2.0	0	0	0	5.0	0
0	0	2.0	0	1.0	0	0	1.0	0
0	0	0	1.0	0	6.0	0	0	3.0
9.0	0	0	0	6.0	0	0	0	1.0
5.0	6.0	0	0	0	0	0	5.0	2.0
0	0	5.0	1.0	0	0	5.0	0	4.0
( 0	0	0	0	3.0	1.0	2.0	4.0	o /

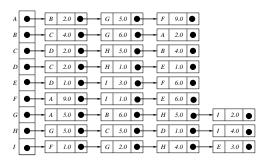
ADS: lects 14 & 15 – slide 3 –

ADS: lects 14 & 15 - slide 2 -

Representations of Weighted Graphs (Adjacency List)



Adjacency Lists



# Connecting Sites

#### Problem

Given a collection of *sites* and *costs* of connecting them, find a minimum cost way of connecting all sites.

### Our Graph Model

- Sites are vertices of a weighted graph, and (non-negative) weights of the edges represent the cost of connecting their endpoints.
- It is reasonable to assume that the graph is undirected and connected.
- The *cost* of a *subgraph* is the sum of the costs of its edges.
- The problem is to find a subgraph of minimum cost that connects all vertices.

# Spanning Trees

 $\mathfrak{G} = (V, E)$  undirected connected graph and W weight function.  $\mathfrak{H} = (V^H, E^H)$  with  $V^H \subseteq V$  and  $E^H \subseteq E$  subgraph of  $\mathfrak{G}$ .

• The *weight* of  $\mathcal H$  is the number

$$W(\mathcal{H}) = \sum_{e \in E^H} W(e).$$

•  $\mathcal{H}$  is a spanning subgraph of  $\mathcal{G}$  if  $V^H = V$ .

### Observation 2

A connected spanning subgraph of minimum weight is a tree.

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# Prim's Algorithm

#### Idea

"Grow" an MST out of a single vertex by always adding "fringe" (neighbouring) edges of minimum weight.

A *fringe edge* for a subtree  $\mathcal{T}$  of a graph is an edge with exactly one endpoint in  $\mathcal{T}$  (so e = (u, v) with  $u \in \mathcal{T}$  and  $v \notin \mathcal{T}$ ).

### Algorithm PRIM(G, W)

- 1.  $\ensuremath{\mathfrak{T}} \leftarrow$  one vertex tree with arbitrary vertex of  $\ensuremath{\mathfrak{G}}$
- 2. while there is a fringe edge do
- 3. add fringe edge of minimum weight to T
- 4. return  $\mathfrak{T}$

Note that this is another use of the greedy strategy.

## ADS: lects 14 & 15 - slide 5 -

# Minimum Spanning Trees

 $(\mathcal{G}, \textit{W})$  undirected connected weighted graph

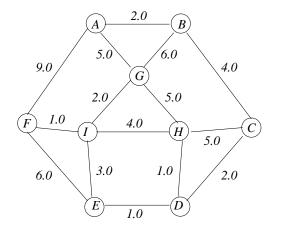
## Definition 3

A minimum spanning tree (MST) of  $\mathcal{G}$  is a connected spanning subgraph  $\mathcal{T}$  of  $\mathcal{G}$  of minimum weight.

### The minimum spanning tree problem:

Given: Undirected connected weighted graph  $(\mathfrak{G}, W)$ Output: An MST of  $\mathfrak{G}$ 

## Example



#### Correctness of Prim's algorithm

1. Throughout the execution of  $\mathrm{PRIM},\,\mathcal{T}$  remains a tree.

*Proof:* To show this we need to show that throughout the execution of the algorithm, T is (i) always connected and (ii) never contains a cycle.

(i) Only edges with an endpoint in  ${\mathfrak T}$  are added to  ${\mathfrak T},$  so  ${\mathfrak T}$  remains connected.

(ii) We never add any edge which has *both* endpoints in  $\mathcal{T}$  (we only allow a single endpoint), so the algorithm will never construct a cycle.

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#### Correctness of Prim's algorithm (cont'd)

2. All vertices will eventually be added to  $\ensuremath{\mathbb{T}}.$ 

*Proof:* by *contradiction* ... (depends on our assumption that the graph  $\mathcal{G}$  was connected.)

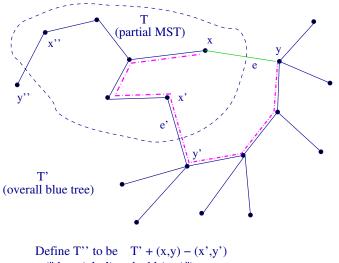
- Suppose w is a vertex that never gets added to T (as usual, in proof by contradiction, we suppose the opposite of what we want).
- Let v = v<sub>0</sub>e<sub>1</sub>v<sub>1</sub>e<sub>2</sub>...v<sub>n</sub> = w be a path from some vertex v inside T to w (we know such a path must exist, because G is connected). Let v<sub>i</sub> be the first vertex on this path that never got added to T.
- ► After v<sub>i-1</sub> was added to T, e<sub>i</sub> = (v<sub>i-1</sub>, v<sub>i</sub>) would have become a fringe edge. Also, it would have remained as a fringe edge unless v<sub>i</sub> was added to T.
- So eventually v<sub>i</sub> must have been added, because Prims algorithm only stops if there are no fringe edges. So our assumption was wrong. So we must have w in T for every vertex w.

#### ADS: lects 14 & 15 – slide 10 –

#### Correctness of Prim's algorithm (cont'd)

- Throughout the execution of PRIM, T is contained in some MST of G. Proof: (by Induction)
  - Suppose that T is contained in an MST T' and that fringe edge e = (x, y) is then added to T by PRIM. We shall prove that T + e is contained in some MST T" (not necessarily T').
  - ▶ case (i): If *e* is contained in T', our proof is easy, we simply let T'' = T'.
  - case (ii): Otherwise, if e ∉ ℑ', consider the unique path 𝔅 from x to y in 𝔅' (𝔅 is the pink path in the example overleaf).

Then  $\mathcal{P}$  contains *exactly one* fringe edge e' = (x', y') (same names in example).



("drop (x',y') and add (x,y)")

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## Towards an Implementation

#### Improvement

- Instead of fringe edges, we think about adding *fringe vertices* to the tree
- A fringe vertex is a vertex y not in T that is an endpoint of a fringe edge.
- The weight of a fringe vertex y is

 $\min\{W(e) \mid e = (x, y) \text{ a fringe edge}\}\$ 

(ie, the best weight that could "bring y into the MST")

To be able to recover the tree, every time we "bring a fringe vertex y into the tree", we store its *parent* in the tree.

We will store the fringe vertices in a priority queue.

Correctness of Prim's algorithm (cont'd)

- 3. case (ii) cont'd
  - Then W(e) ≤ W(e').
    (otherwise e' would definitely have been added before e)
  - Let  $\mathfrak{T}'' = \mathfrak{T}' + e e'$ .
  - ▶ T" is a tree.

Why? Well, we drop e' = (x', y'), which splits the global MST T into two components:  $\mathfrak{T}'_{x'}$  and the other subtree  $\mathfrak{T}'_{y'} = \mathfrak{T}' \setminus \mathfrak{T}'_{x'}$ . We know x and y are now in different components after this split, because we have *broken* the unique path  $\mathcal{P}$  between x and y in  $\mathfrak{T}'$ . Hence we can add e = (x, y) to *re-join*  $\mathfrak{T}'_{x'}$  and  $\mathfrak{T}'_{y'}$  without making a cycle.

 $\mathfrak{T}''$  has the same vertices as  $\mathfrak{T}',$  thus it is a spanning tree.

• Moreover,  $W(\mathcal{T}'') = W(\mathcal{T}') + W(e) - W(e')$ , and because we know  $W(e) \leq W(e')$ , this gives  $W(\mathcal{T}'') \leq W(\mathcal{T}')$ , thus  $\mathcal{T}''$  is also a MST.

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# Priority Queues with Decreasing Key

A *Priority Queue* is an ADT for storing a collection of elements with an associated *key*. The following methods are supported:

- INSERT(e, k): Insert element *e* with key *k*.
- ► GET-MIN(): Return an element with minimum key; an error occurs if the priority queue is empty.
- EXTRACT-MIN(): Return and remove an element with minimum key; an error if the priority queue is empty.
- ► IS-EMPTY(): Return TRUE if the priority queue is empty and FALSE otherwise.

To update the keys during the execution of  ${\rm PRIM},$  we need priority queues supporting the following additional method:

 DECREASE-KEY(e, k): Set the key of e to k and update the priority queue. It is assumed that k is smaller than of equal to the old key of e.

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# Implementation of Prim's Algorithm

A 1	$\mathbf{D}_{\mathbf{D}}$				
Algo	rithm $PRIM(\mathcal{G}, W)$	-	Algorithm $\operatorname{ReLAX}(y,$		
1.	Initialise parent array $\pi$ :	1.	$w \leftarrow W(y, z)$		
	$\pi[v] \leftarrow  ext{NIL}$ for all vertices $v$	2.	if weight $[z] = c$		
2.	0,	3.	weight[z		
	weight[ $v$ ] $\leftarrow \infty$ for all $v$	4.	$\pi[z] \leftarrow$		
3.	Initialise inMST array:	5.	Q.INSER		
Л	$inMST[v] \leftarrow false \text{ for all } v$	6.	else if $(w < we)$		
4.		7.	not (		
	$v \leftarrow \text{arbitrary vertex of } \mathcal{G}$	8.	_		
	Q.INSERT(v, 0)		-		
7.	weight $[v] = 0;$	9.	$\pi[z] \leftarrow j$		
8.	while not( $Q$ .Is-EMPTY()) do	10.	Q.Decf		
9.	$y \leftarrow Q.$ Extract-Min()				
10.	$inMST[y] \leftarrow true$				
11.	for all z adjacent to y do				
12.	RELAX(y,z)				
13.	return $\pi$				
		]			

thm RELAX(y, z)  $w \leftarrow W(y, z)$ if weight $[z] = \infty$  then weight $[z] \leftarrow w$   $\pi[z] \leftarrow y$  Q.INSERT(z, w)else if (w < weight[z] and not (inMST[z])) then weight $[z] \leftarrow w$   $\pi[z] \leftarrow y$ Q.DECREASE KEY(z, w)

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# Analysis of **PRIM's algorithm** (RELAX)

- Decreasing the time needed to execute INSERT and DECREASE-KEY, the execution of RELAX requires time Θ(1).
- $\blacktriangleright$   $\operatorname{INSERT}$  is executed once for every vertex, which requires time

$$\Theta(\mathbf{n} \cdot \mathbf{T}_{\text{INSERT}}(\mathbf{n}))$$

 DECREASE-KEY is executed at most once for every edge. This can require time of size

$$\Theta(\boldsymbol{m} \cdot \boldsymbol{T}_{\text{Decrease-Key}}(\boldsymbol{n}))$$

Overall, we get

$$T_{\text{PRIM}}(n,m) = \Theta\left(n\left(T_{\text{EXTRACT-MIN}}(n) + T_{\text{INSERT}}(n)\right) + mT_{\text{DECREASE-KEY}}(n)\right)$$

# Analysis of $\operatorname{PRIM}\nolimits$ 's algorithm

Let n be the number of vertices and m the number of edges of the input graph.

- ▶ Lines 1-7, 13 of Prim require  $\Theta(n)$  time altogether.
- ▶ Q will extract each of the n vertices of G once. Thus the loop at lines 8-12 is iterated n times.

Thus, disregarding (for now) the time to execute the inner loop (lines 11-12) the execution of the loop requires time

 $\Theta(\mathbf{n} \cdot \mathbf{T}_{\text{EXTRACT-MIN}}(\mathbf{n}))$ 

The inner loop is executed at most once for each edge (and at least once for each edge). So its execution requires time

 $\Theta(m \cdot T_{\text{RELAX}}(n,m)).$ 

ADS: lects 14 & 15 - slide 18 -

# Priority Queue Implementations

- Array: Elements simply stored in an array.
- Heap: Elements are stored in a binary heap (see [CLRS] Section 6.5)
- Fibonacci Heap: Sophisticated variant of the simple binary heap (see [CLRS] Chapters 19 and 20)

method		running time			
	Array	Heap	Fibonacci Heap		
INSERT	$\Theta(1)$	$\Theta(\lg n)$	$\Theta(1)$		
Extract-Min	$\Theta(n)$	$\Theta(\lg n)$	$\Theta(\lg n)$		
Decrease-Key	$\Theta(1)$	$\Theta(\lg n)$	$\Theta(1)$ (amortised)		

## Running-time of PRIM

### Remarks

 $T_{\text{PRIM}}(n,m) = \Theta\left(n\left(T_{\text{EXTRACT-MIN}}(n) + T_{\text{INSERT}}(n)\right) + mT_{\text{DECREASE-Key}}(n)\right)$ 

Which Priority Queue implementation?

With array implementation of priority queue:

 $T_{\rm PRIM}(n,m) = \Theta(n^2).$ 

• With heap implementation of priority queue:

$$T_{\mathrm{PRIM}}(n,m) = \Theta((n+m)\lg(n)).$$

• With Fibonacci heap implementation of priority queue:

$$T_{\text{PRIM}}(n,m) = \Theta(n \lg(n) + m).$$

(*n* being the number of vertices and *m* the number of edges)

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## **Reading Assignment**

[CLRS] Chapter 23.

#### **Problems**

- 1. Exercises 23.1-1, 23.1-2, 23.1-4 of [CLRS]
- 2. In line 3 of Prim's algorithm, there may be more than one fringe edge of minimum weight. Suppose we add all these minimum edges in one step. Does the algorithm still compute a MST?
- 3. Prove that our *implementation* of Prim's algorithm on slide 6 is correct i.e., that it computes an MST.

- The Fibonacci heap implementation is mainly of theoretical interest. It is not much used in practice because it is very complicated and the constants hidden in the Θ-notation are large.
- For dense graphs with m = Θ(n<sup>2</sup>), the array implementation is probably the best, because it is so simple.
- For sparser graphs with m ∈ O(<sup>n<sup>2</sup></sup>/<sub>lgn</sub>), the heap implementation is a good alternative, since it is still quite simple, but more efficient for smaller m.

Instead of using binary heaps, the use of *d*-ary heaps for some  $d \ge 1$  can speed up the algorithm (see [Sedgewick] for a discussion of practical implementations of Prims algorithm).

ADS: lects 14 & 15 - slide 22 -