Algorithms

 \checkmark

ROBERT SEDGEWICK | KEVIN WAYNE

Algorithms

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4.3 MINIMUM SPANNING TREES

introduction

- greedy algorithm
- edge-weighted graph API
- Kruskal's algorithm
- Prim's algorithm
- context

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KRUSKAL'S ALGORITHM DEMO



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Consider edges in ascending order of weight.

• Add next edge to tree *T* unless doing so would create a cycle.

graph edges sorted by weight



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does not create a cycle

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Kruskal's algorithm: visualization



Kruskal's algorithm: visualization



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PRIM'S ALGORITHM DEMO

Prim's algorithm

lazy implementation

eager implementation

Algorithms

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- Start with vertex 0 and greedily grow tree *T*.
- Add to *T* the min weight edge with exactly one endpoint in *T*.
- Repeat until V 1 edges.



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MST edges

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MST edges
Prim's algorithm demo

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```
MST edges
```

Prim's algorithm: visualization



Prim's algorithm: visualization



Prim's algorithm: implementation challenge

Challenge. Find the min weight edge with exactly one endpoint in *T*.

- *E*
- V
- $\log E$
- $\log^* E$
- 1



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Lazy solution. Maintain a PQ of edges with (at least) one endpoint in *T*.

- Key = edge; priority = weight of edge.
- Delete-min to determine next edge e = v w to add to T.



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- Disregard if both endpoints v and w are marked (both in T).



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- Key = edge; priority = weight of edge.
- Delete-min to determine next edge e = v w to add to T.
- Disregard if both endpoints v and w are marked (both in T).
- Otherwise, let w be the unmarked vertex (not in T):
 - add to PQ any edge incident to w (assuming other endpoint not in T)
 - add e to T and mark w



Prim(graph G)

```
PQ = empty priority queue of edges
color all vertices grey
Visit(0)
while (|A| < n - 1)
     (u,v) = PQ.DeleteMin()
     if u or v is grey
         A = A \cup (u, v)
     if u is grey
          Visit(u)
     else // v is grey
          Visit(v)
```

Visit(vertex u) color u black for all edges (u,v) if v is grey PQ.insert((u,v)) **Proposition.** Lazy Prim's algorithm computes the MST in time proportional to $E \log E$ and extra space proportional to E (in the worst case).

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Pf.

operation	frequency	binary heap	
delete min	E	$\log E$	
insert	E	log E	

Prim's algorithm: eager implementation

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Observation. For each vertex *v*, need only min weight edge connecting *v* to *T*.



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• MST includes at most one edge connecting v to T. Why?



Observation. For each vertex *v*, need only min weight edge connecting *v* to *T*.

- MST includes at most one edge connecting v to T. Why?
- If MST includes such an edge, it can take cheapest such edge. Why?



pq has at most one entry per vertex

Eager solution. Maintain a PQ of vertices connected by an edge to T, where priority of vertex v = weight of min weight edge connecting v to T.



pq has at most one entry per vertex

Eager solution. Maintain a PQ of vertices connected by an edge to T, where priority of vertex v = weight of min weight edge connecting v to T.

• Delete min vertex v and add its associated edge e = v - w to T.



pq has at most one entry per vertex

Eager solution. Maintain a PQ of vertices connected by an edge to T, where priority of vertex v = weight of min weight edge connecting v to T.

- Delete min vertex v and add its associated edge e = v w to T.
- Update PQ by considering all edges e = v x incident to v
 - ignore if x is already in T
 - add x to PQ if not already on it
 - decrease priority of x if v-x becomes min weight edge connecting x

to T



Eager implementation of Prim's algorithm

Prim(graph G)

PQ = empty priority queue of vertices

```
cost = array of size n
```

```
edge = array of size n
```

color all vertices grey

Visit(0)

```
while(PQ not empty)
```

```
u = PQ.DeleteMin()
A = A \cup edge[u]
Visit(u)
```

Visit(vertex u)

color u black

```
for all edges (u,v)
```

 $if v is grey \\ color v red \\ PQ.insert(v, w(u,v)) \\ cost[v] = w(u,v) \\ edge[v] = (u,v) \\ elseif (v is red) and (w(u,v) < cost[v]) \\ PQ.DecreaseKey(v, w(u,v)) \\ cost[v] = w(u,v) \\ edge[v] = (u,v) \\ edge[v] = (u,v) \\$

PQ implementation	insert	delete-min	decrease-key	total
unordered array	1	V	1	V^2

Bottom line.

• Array implementation optimal for dense graphs.

PQ implementation	insert	delete-min	decrease-key	total
unordered array	1	V	1	V^2
binary heap	log V	log V	log V	$E \log V$

Bottom line.

- Array implementation optimal for dense graphs.
- Binary heap much faster for sparse graphs.

PQ implementation	insert	delete-min	decrease-key	total
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binary heap	$\log V$	log V	$\log V$	$E \log V$
d-way heap	$\log_d V$	$d \log_d V$	$\log_d V$	$E \log_{E/V} V$

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- 4-way heap worth the trouble in performance-critical situations.

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Fibonacci heap	1 †	$\log V^{\dagger}$	1 †	$E + V \log V$

† amortized

Bottom line.

- Array implementation optimal for dense graphs.
- Binary heap much faster for sparse graphs.
- 4-way heap worth the trouble in performance-critical situations.
- Fibonacci heap best in theory, but not worth implementing.

Challenge. Would adding edge v-w to tree T create a cycle? If not, add it.

- E + V
- V
- log *V*
- log* V
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Challenge. Would adding edge v-w to tree *T* create a cycle? If not, add it.

- log *V*
- log* V ← use the union-find data structure !
- 1



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Efficient solution. Use the union-find data structure.

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- If *v* and *w* are in same set, then adding *v*-*w* would create a cycle.



Case 1: adding v-w creates a cycle
Kruskal's algorithm: implementation challenge

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- Maintain a set for each connected component in *T*.
- If *v* and *w* are in same set, then adding *v*-*w* would create a cycle.
- To add *v*-*w* to *T*, merge sets containing *v* and *w*.



Case 1: adding v-w creates a cycle

Case 2: add v-w to T and merge sets containing v and w

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Kruskal's algorithm: Java implementation



Proposition. Kruskal's algorithm computes MST in time proportional to $E \log E$ (in the worst case).

Kruskal's algorithm: running time

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Pf.	operation	frequency	time per op
	build pq	1	E
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	connected	E	$\log^* V^{\dagger}$

† amortized bound using weighted quick union with path compression

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+ amortized bound using weighted quick union with path compression

```
recall: \log^* V \le 5 in this universe
Remark. If edges are already sorted, order of growth is E \log^* V.
```

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Does a linear-time MST algorithm exist?

year worst case	discovered by
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Remark. Linear-time randomized MST algorithm (Karger-Klein-Tarjan 1995).

Euclidean MST

Given *N* points in the plane, find MST connecting them, where the distances between point pairs are their Euclidean distances.



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Brute force. Compute ~ N^2 / 2 distances and run Prim's algorithm.

Euclidean MST

Given *N* points in the plane, find MST connecting them, where the distances between point pairs are their Euclidean distances.



Brute force. Compute ~ N^2 / 2 distances and run Prim's algorithm. Ingenuity. Exploit geometry and do it in ~ $c N \log N$.

Scientific application: clustering

k-clustering. Divide a set of objects classify into k coherent groups.Distance function. Numeric value specifying "closeness" of two objects.

Goal. Divide into clusters so that objects in different clusters are far apart.



outbreak of cholera deaths in London in 1850s (Nina Mishra)

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Applications.

- Routing in mobile ad hoc networks.
- Document categorization for web search.
- Similarity searching in medical image databases.
- Skycat: cluster 10⁹ sky objects into stars, quasars, galaxies.

k-clustering. Divide a set of objects classify into k coherent groups.Distance function. Numeric value specifying "closeness" of two objects.

Single link. Distance between two clusters equals the distance between the two closest objects (one in each cluster).

Single-link clustering. Given an integer *k*, find a *k*-clustering that maximizes the distance between two closest clusters.



"Well-known" algorithm in science literature for single-link clustering:

- Form V clusters of one object each.
- Find the closest pair of objects such that each object is in a different cluster, and merge the two clusters.
- Repeat until there are exactly *k* clusters.



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Alternate solution. Run Prim; then delete k - 1 max weight edges.

Dendrogram of cancers in human

Tumors in similar tissues cluster together.



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