



# Algorithm design and analysis

### — Greedy graph algorithms —

#### Silvio Guimarães

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# Algorithm design and analysis — Graphs —

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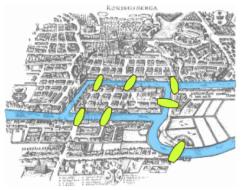
- Model pairwise relationships (edges) between objects (nodes or vertices).
- ► Undirected graph G = (V, E): set V of nodes and set E of edges, where E ⊆ V × V. Elements of E are unordered pairs.
- ▶ Directed graph G = (V, E): set V of nodes and set E of edges, where  $E \subseteq V \times V$ . Elements of E are ordered pairs.

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- ▶ Problems involving graphs have a rich history dating back to Euler.

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- ▶ Problems involving graphs have a rich history dating back to Euler.







## Algorithm design and analysis

— Shortest Paths —

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### Shortest Path Problem

- G = (V, E) is a connected directed graph. Each edge e has a length l<sub>e</sub> ≥ 0.
- ► V has n nodes and E has m edges.
- ► Length of a path *P* is the sum of lengths of the edges in *P*.
- ► Goal is to determine the shortest path from some start node s to each node in V.

Aside: If G is undirected, convert to a directed graph by replacing each edge in G by two directed edges.

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#### SHORTEST PATHS

**INSTANCE** A directed graph G(V, E), a function  $I : E \to \mathbb{R}^+$ , and a node  $s \in V$ 

**SOLUTION** A set  $\{P_u, u \in V\}$ , where  $P_u$  is the shortest path in *G* from *s* to *u*.

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Greedy graph algorithm

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Algorithm: Shortest path algorithm - Dijkstra

**input** : A graph G = (V, E), a weight map W and a source node s. **output**: The distances of the vertices from s

- 1 Let S be the set of explored nodes;
- **2** foreach  $u \in S$  do store distance  $d[u] = \infty$ ;
- 3 Initially d[s] = 0 and S = s;

```
4 while S \neq V do

5 Select a node v \notin S with at least one edge from S for which

d'(v) = \min_{e=(u,v):u \in S} d[u] + W(e) is as small as possible;
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6 | Add v to S and define d[v] = d'[v];
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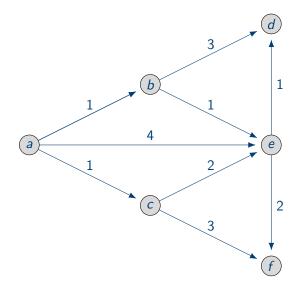
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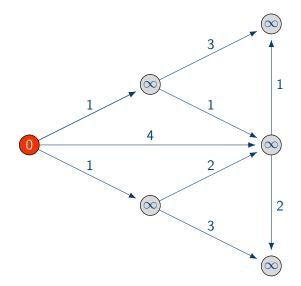
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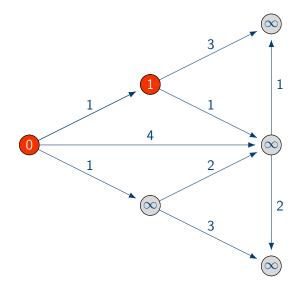
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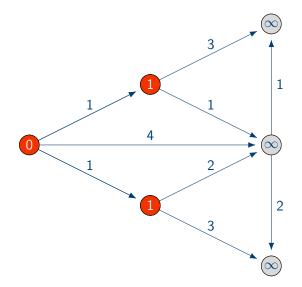
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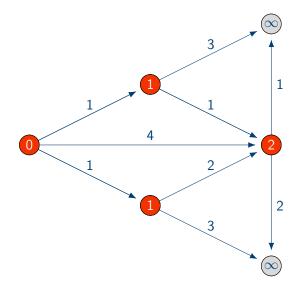
Can modify algorithm to compute the shortest paths themselves: record the predecessor u that minimizes d'(v).

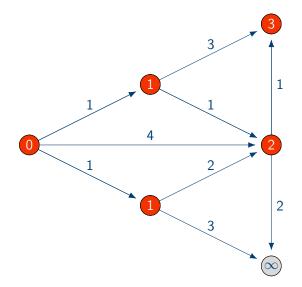


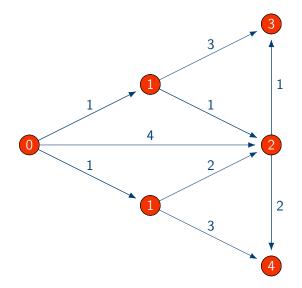










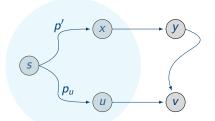


### **Proof of Correctness**

- Let  $P_u$  be the shortest path computed for a node u.
- Claim:  $P_u$  is the shortest path from s to u.
- Prove by induction on the size of *S*.
  - Base case: |S| = 1. The only node in S is s.
  - Inductive step: we add the node v to S. Let u be the v's predecessor on the path P<sub>v</sub>. Could there be a shorter path P from s to v?

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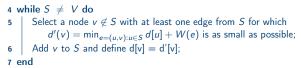
The alternate s - v path Pthrough x and y already too long by the time it had left the set S

### Comments about Dijkstra's Algorithm

- ► Algorithm cannot handle negative edge lengths.
- Union of shortest paths output form a tree. Why?

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► How many iterations are there of the while loop? .

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  - How many iterations are there of the while loop? n-1.
  - In each iteration, for each node v ∉ S, compute min<sub>e=(u,v),u∈S</sub> d(u) + l<sub>e</sub>.

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- In each iteration, for each node v ∉ S, compute min<sub>e=(u,v),u∈S</sub> d(u) + l<sub>e</sub>.
- Running time per iteration is .

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- How many iterations are there of the while loop? n-1.
- In each iteration, for each node v ∉ S, compute min<sub>e=(u,v),u∈S</sub> d(u) + l<sub>e</sub>.
- Running time per iteration is O(m), yielding an overall running time of O(nm).

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• Observation: If we add v to S, d'(w) changes only for v's neighbours.

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- Observation: If we add v to S, d'(w) changes only for v's neighbours.
- Store the minima d'(v) for each node  $v \in V S$  in a priority queue.
- ▶ Determine the next node *v* to add to *S* using EXTRACTMIN.
- After adding v, for each neighbour w of v, compute  $d(v) + l_{(v,w)}$ .
- ▶ If  $d(v) + l_{(v,w)} < d'(w)$ , update w's key using CHANGEKEY.

#### Algorithm: Shortest path algorithm - Dijkstra

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- ► How many times are EXTRACTMIN and CHANGEKEY invoked?

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- ▶ If  $d(v) + l_{(v,w)} < d'(w)$ , update w's key using CHANGEKEY.
- ▶ How many times are EXTRACTMIN and CHANGEKEY invoked? n-1 and *m* times, respectively.

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- ▶ If  $d(v) + l_{(v,w)} < d'(w)$ , update w's key using CHANGEKEY.
- ► How many times are EXTRACTMIN and CHANGEKEY invoked? n 1 and m times, respectively. Total running time is O(m log n).





# Algorithm design and analysis — Minimum Spanning Trees —

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- ► Connect a set of nodes using a set of edges with certain properties.
- Input is usually a graph and the desired network (the output) should use subset of edges in the graph.
- ► Example: connect all nodes using a cycle of shortest total length.

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- Input is usually a graph and the desired network (the output) should use subset of edges in the graph.
- Example: connect all nodes using a cycle of shortest total length. This problem is the NP-complete traveling salesman problem.

# Minimum Spanning Tree (MST)

- ► Given an undirected graph G = (V, E) with a cost c<sub>e</sub> > 0 associated with each edge e ∈ E.
- ▶ Find a subset *T* of edges such that the graph (V, T) is connected and the cost  $\sum_{e \in T} c_e$  is as small as possible.

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#### MINIMUM SPANNING TREE

- **INSTANCE** An undirected graph G = (V, E) and a function  $c : E \to \mathbb{R}^+$
- **SOLUTION** A set  $T \subseteq E$  of edges such that (V, T) is connected and the  $\sum_{e \in T} c_e$  is as small as possible.

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#### MINIMUM SPANNING TREE

- **INSTANCE** An undirected graph G = (V, E) and a function  $c : E \to \mathbb{R}^+$
- **SOLUTION** A set  $T \subseteq E$  of edges such that (V, T) is connected and the  $\sum_{e \in T} c_e$  is as small as possible.
- Claim: If T is a minimum-cost solution to this network design problem then (V, T) is a tree.
- A subset T of E is a spanning tree of G if (V, T) is a tree.

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 Increasing cost order Process edges in increasing order of cost. Discard an edge if it creates a cycle.
 Dijkstra-like Start from a node s and grow T outward from s: add the node that can be attached most cheaply to current tree.
 Decreasing cost order Delete edges in order of decreasing cost as

long as graph remains connected.

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   Increasing cost order Process edges in increasing order of cost. Discard an edge if it creates a cycle.
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  - add the node that can be attached most cheaply to current tree.
  - **Decreasing cost order** Delete edges in order of decreasing cost as long as graph remains connected.
- Which of these algorithms works?

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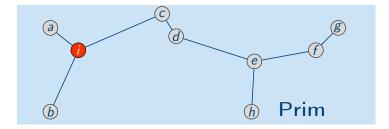
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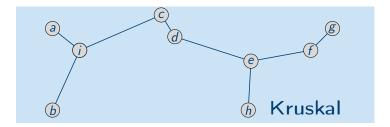
Discard an edge if it creates a cycle. Kruskal's algorithm

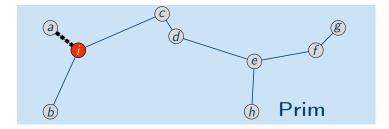
**Dijkstra-like** Start from a node s and grow T outward from s: add the node that can be attached most cheaply to current tree. Prim's algorithm

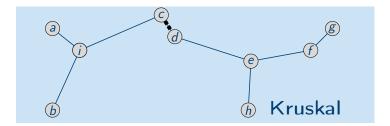
**Decreasing cost order** *Delete edges in order of decreasing cost as long as graph remains connected. Reverse-Delete algorithm* 

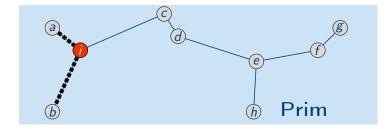
Which of these algorithms works? All of them!

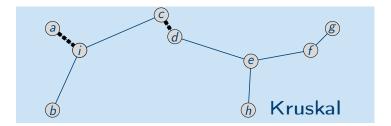


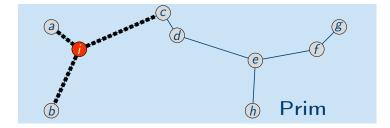


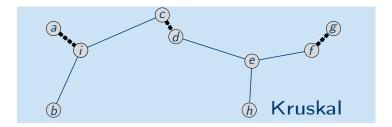


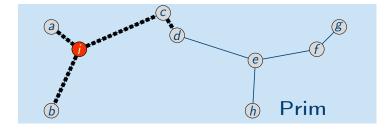


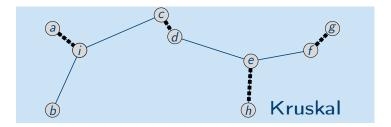


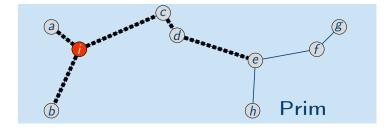


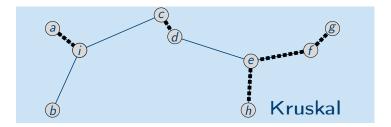




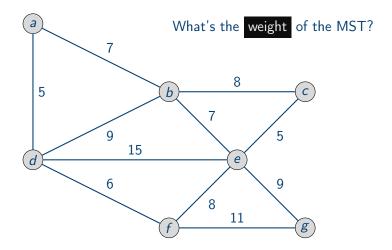




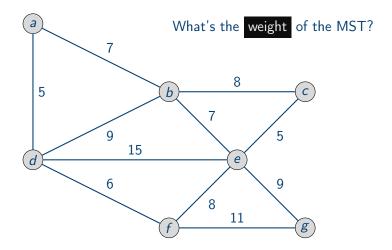




# Example of Prim's Algorithm

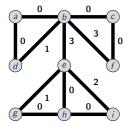


# Example of Kruskal's Algorithm



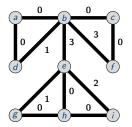
► A cut in a graph G = (V, E) is a set of edges whose removal disconnects the graph (into two or more connected components).

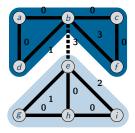
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#### Silvio Guimarães

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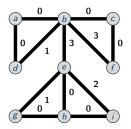


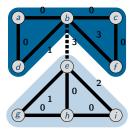


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#### Greedy graph algorithm

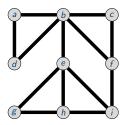
- ► A cut in a graph G = (V, E) is a set of edges whose removal disconnects the graph (into two or more connected components).
- Every set S ⊂ V (S cannot be empty or the entire set V) has a corresponding cut: cut(S) is the set of edges (v, w) such that v ∈ S and w ∈ V − S.



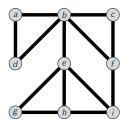


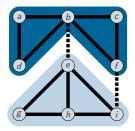
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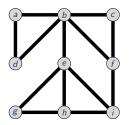
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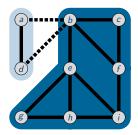




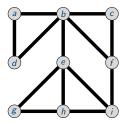
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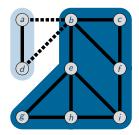
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- ► cut(S) is a cut because deleting the edges in cut(S) disconnects S from V S.





#### ► When is it **safe** to include an edge in an MST?

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- Proof: exchange argument. If a supposed MST T does not contain e, show that there is a tree with smaller cost than T that contains e.

- ▶ Let *F* be the set of all edges that satisfy the cut property.
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- ► *F* is the unique MST.
- ► Kruskal's and Prim's algorithms compute *F* efficiently.

# **Optimality of Kruskal's Algorithm**

#### Kruskal's algorithm:

- ► Start with an empty set *T* of edges.
- ► Process edges in *E* in non decreasing order of cost.
- Add the next edge e to T only if adding e does not create a cycle.
   Discard e if it creates a cycle.
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- ► Claim: Kruskal's algorithm outputs an MST.
  - 1. For every edge e added, demonstrate the existence of S and V S such that e and S satisfy the cut property.
  - 2. Prove that the algorithm computes a spanning tree.

- Prim's algorithm: Maintain a tree (S, U)
  - Start with an **arbitrary** node  $s \in S$  and  $U = \emptyset$ .
  - ► Add the node v to S and the edge e to U that minimize

$$\min_{e=(u,v), u\in S, v\notin S} c_e \equiv \min_{e\in \operatorname{cut}(S)} c_e.$$

- Stop when S = V.
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- Claim: Prim's algorithm outputs an MST.
  - 1. Prove that every edge inserted satisfies the cut property.
  - 2. Prove that the graph constructed is a spanning tree.

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### • Reverse-Delete algorithm: Maintain a set E' of edges.

- Start with E' = E.
- Process edges in non increasing order of cost.
- ► Delete the next edge *e* from E' only if (V, E') is connected after removal.
- Stop after processing all the edges.
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- ► Claim: the Reverse-Delete algorithm outputs an MST.
  - 1. Show that every edge deleted belongs to no MST.
  - 2. Prove that the graph remaining at the end is a spanning tree.

- To handle multiple edges with the same weight, perturb each length by a random infinitesimal amount.
- Any algorithm that constructs a spanning tree by including edges that satisfy the cut property and deleting edges that satisfy the cycle property will yield an MST!





# Algorithm design and analysis

— Implementation —

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### Implementing Prim's Algorithm

- Maintain a tree (S, U).
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- Sorting edges takes  $O(m \log n)$  time.
- Implementation is very similar to Dijkstra's algorithm.
- ► Maintain S and store attachment costs a(v) = min<sub>e∈cut(S)</sub> c<sub>e</sub> for every node v ∈ V − S in a priority queue.
- ► At each step, extract minimum v from priority queue and update the attachment costs of the neighbours of v.
- ► Total of n 1 EXTRACTMIN and m CHANGEKEY operations, yielding a running time of  $O(m \log n)$ .

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- ► Add the next edge *e* to *T* only if adding *e* does not create a cycle.
- Sorting edges takes  $O(m \log n)$  time.
- Key question: "Does adding e = (u, v) to T create a cycle?"
  - ► Maintain set of connected components of *T*.
  - ► **FIND**(u): return the name of the connected component of *T* that *u* belongs to.
  - UNION(A, B): merge connected components A and B.
- Answering the question: Adding e creates a cycle if and only if FIND(u) = FIND(v). If not, execute UNION(FIND(u), FIND(v)).

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  - ► Each FIND takes O(1) time, k invocations of UNION take O(k log k) time in total.
  - Each FIND takes O(log n) time and each invocation of UNION takes O(1) time.

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  - Each FIND takes O(log n) time and each invocation of UNION takes O(1) time.
- Total running time of Kruskal's algorithm is  $O(m \log n)$ .





# Algorithm design and analysis

— Huffman code —

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Feb 2023

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How do we know when the next symbol begins?

Use a separation symbol (like the pause in Morse), or make sure that there is no ambiguity by ensuring that no code is a prefix of another one

A prefix code for a set S is a function c that maps each  $x \in S$  to 1s and 0s in such a way that for  $x, y \in S, x \neq y, c(x)$  is not a prefix of c(y).

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Suppose frequencies are known in a text of 1G characters:  $f_a = 0.4$ ,  $f_e = 0.2$ ,  $f_k = 0.2$ ,  $f_l = 0.1$ ,  $f_u = 0.1$ . What is the size of the encoded text?

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$$2^{fa} + 2^{fe} + 3^{fk} + 2^{fl} + 4^{fu} = 2.4 \text{ G bits}$$



### How to greedly compute a prefix tree to encode an alphabet?

### Suppose frequencies are known: $f_a = 0.32, f_e = 0.25, f_k = 0.20, f_l = 0.18, f_u = 0.05.$ How to create an encoding to minimize the size of a text?

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```
Algorithm: Huffman code
   input : A set S of elements with their frequencies.
   output: A prefix tree
 1 if S = 2 then
       return a tree with root and 2 leaves;
 3 else
      let v and z be lowest-frequency letters in S:
      S' = S
 5
      remove y and z from S';
 6
      insert new letter w in S' with f_w = f_v + f_z;
 7
      T' = Huffman(S'):
 8
      T = add two children y and z to leaf w from T';
 9
      return T:
10
11 end
```