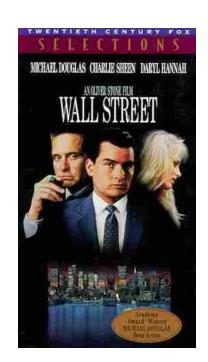
## Coin Changing

Greed is good. Greed is right. Greed works.
Greed clarifies, cuts through, and captures the essence of the evolutionary spirit.

- Gordon Gecko (Michael Douglas)





## Coin Changing

Goal. Given currency denominations: 1, 5, 10, 25, 100, devise a method to pay amount to customer using fewest number of coins.

Ex: 34¢.



Cashier's algorithm. At each iteration, add coin of the largest value that does not take us past the amount to be paid.

Ex: \$2.89.



## Coin-Changing: Analysis of Greedy Algorithm

Observation. Greedy algorithm is sub-optimal for US postal denominations: 1, 10, 21, 34, 70, 100, 350, 1225, 1500.

Counterexample. 140¢.

• Greedy: 100, 34, 1, 1, 1, 1, 1.

• Optimal: 70, 70.



















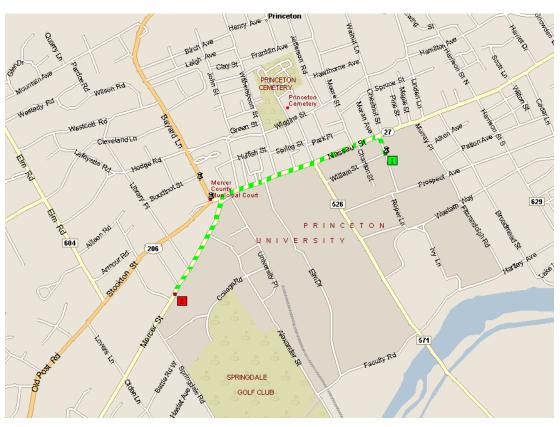
#### Coin-Changing: Analysis of Greedy Algorithm

Theorem. Greed is optimal for U.S. coinage: 1, 5, 10, 25, 100. Pf. (by induction on x)

- Consider optimal way to change  $c_k \le x < c_{k+1}$ : greedy takes coin k.
- We claim that any optimal solution must also take coin k.
  - if not, it needs enough coins of type  $c_1, ..., c_{k-1}$  to add up to x
  - table below indicates no optimal solution can do this
- Problem reduces to coin-changing  $x c_k$  cents, which, by induction, is optimally solved by greedy algorithm. ■

k	c <sub>k</sub>	All optimal solutions must satisfy	Max value of coins 1, 2,, k-1 in any OPT
1	1	P ≤ 4	-
2	5	N ≤ 1	4
3	10	N + D ≤ 2	4 + 5 = 9
4	25	Q ≤ 3	20 + 4 = 24
5	100	no limit	75 + 24 = 99

# 4.4 Shortest Paths in a Graph



shortest path from Princeton CS department to Einstein's house

## Edsger W. Dijkstra

The question of whether computers can think is like the question of whether submarines can swim.

Do only what only you can do.

In their capacity as a tool, computers will be but a ripple on the surface of our culture. In their capacity as intellectual challenge, they are without precedent in the cultural history of mankind.

The use of COBOL cripples the mind; its teaching should, therefore, be regarded as a criminal offence.

APL is a mistake, carried through to perfection. It is the language of the future for the programming techniques of the past: it creates a new generation of coding bums.



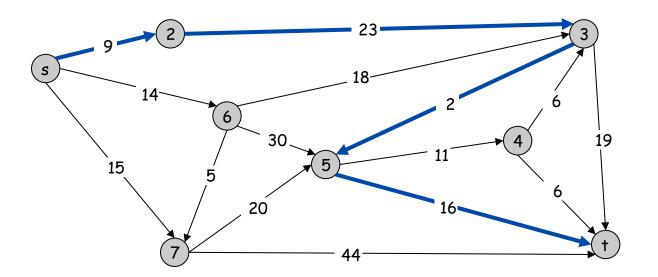
#### Shortest Path Problem

#### Shortest path network.

- Directed graph G = (V, E).
- Source s, destination t.
- Length  $\ell_e$  = length of edge e.

Shortest path problem: find shortest directed path from s to t.

cost of path = sum of edge costs in path



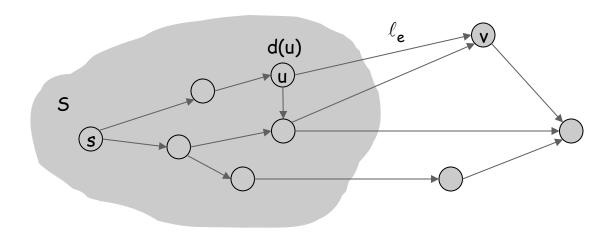
Cost of path s-2-3-5-t = 9 + 23 + 2 + 16 = 48.

## Dijkstra's Algorithm

#### Dijkstra's algorithm.

- Maintain a set of explored nodes S for which we have determined the shortest path distance d(u) from s to u.
- Initialize  $S = \{s\}, d(s) = 0$ .
- Repeatedly choose unexplored node v which minimizes

$$\pi(v) = \min_{e = (u,v) : u \in S} d(u) + \ell_e,$$
 shortest path to some u in explored part, followed by a single edge (u, v)

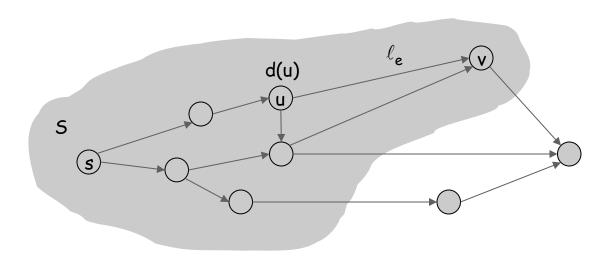


## Dijkstra's Algorithm

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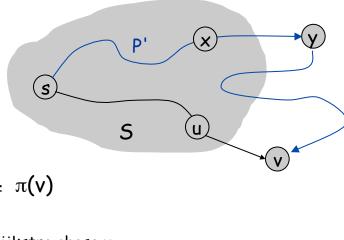
## Dijkstra's Algorithm: Proof of Correctness

Invariant. For each node  $u \in S$ , d(u) is the length of the shortest s-u path. Pf. (by induction on |S|)

Base case: |S| = 1 is trivial.

Inductive hypothesis: Assume true for  $|S| = k \ge 1$ .

- Let v be next node added to S, and let u-v be the chosen edge.
- The shortest s-u path plus (u, v) is an s-v path of length  $\pi(v)$ .
- Consider any s-v path P. We'll see that it's no shorter than  $\pi(v)$ .
- Let x-y be the first edge in P that leaves 5, and let P' be the subpath to x.
- P is already too long as soon as it leaves S.



$$\ell\left(P\right) \geq \ell\left(P'\right) + \ell\left(x,y\right) \geq d(x) + \ell\left(x,y\right) \geq \pi(y) \geq \pi(v)$$

$$\uparrow \qquad \uparrow \qquad \uparrow \qquad \uparrow$$

$$\text{nonnegative inductive defin of } \pi(y) \qquad \text{Dijkstra chose } v$$

$$\text{weights} \qquad \text{hypothesis} \qquad \text{instead of } y$$

## Dijkstra's Algorithm: Implementation

For each unexplored node, explicitly maintain  $\pi(v) = \min_{e = (u,v): u \in S} d(u) + \ell_e$ .

- Next node to explore = node with minimum  $\pi(v)$ .
- When exploring v, for each incident edge e = (v, w), update

$$\pi(w) = \min \left\{ \pi(w), \ \pi(v) + \ell_e \right\}.$$

Efficient implementation. Maintain a priority queue of unexplored nodes, prioritized by  $\pi(v)$ .

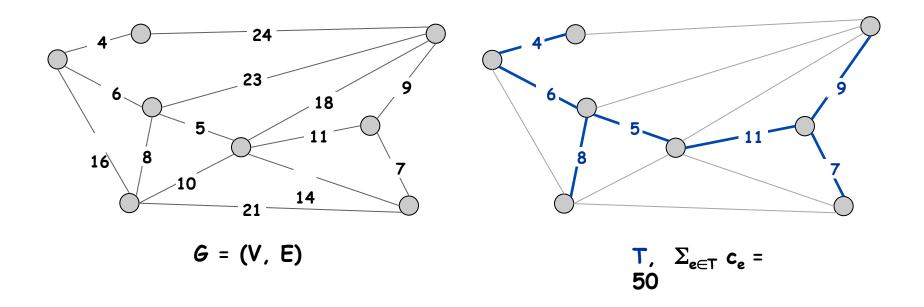
PQ Operation	Dijkstra	Array	Binary heap	d-way Heap	Fib heap †
Insert	n	n	log n	d log <sub>d</sub> n	1
ExtractMin	n	n	log n	d log <sub>d</sub> n	log n
ChangeKey	m	1	log n	log <sub>d</sub> n	1
IsEmpty	n	1	1	1	1
Total		n <sup>2</sup>	m log n	m log <sub>m/n</sub> n	m + n log n

<sup>†</sup> Individual ops are amortized bounds

# 4.5 Minimum Spanning Tree

## Minimum Spanning Tree

Minimum spanning tree. Given a connected graph G = (V, E) with real-valued edge weights  $c_e$ , an MST is a subset of the edges  $T \subseteq E$  such that T is a spanning tree whose sum of edge weights is minimized.



Cayley's Theorem. There are  $n^{n-2}$  spanning trees of  $K_n$ .

can't solve by brute force

#### **Applications**

#### MST is fundamental problem with diverse applications.

- Network design.
  - telephone, electrical, hydraulic, TV cable, computer, road
- Approximation algorithms for NP-hard problems.
  - traveling salesperson problem, Steiner tree
- Indirect applications.
  - max bottleneck paths
  - LDPC codes for error correction
  - image registration with Renyi entropy
  - learning salient features for real-time face verification
  - reducing data storage in sequencing amino acids in a protein
  - model locality of particle interactions in turbulent fluid flows
  - autoconfig protocol for Ethernet bridging to avoid cycles in a network
- Cluster analysis.

## Greedy Algorithms

Kruskal's algorithm. Start with  $T = \phi$ . Consider edges in ascending order of cost. Insert edge e in T unless doing so would create a cycle.

Reverse-Delete algorithm. Start with T = E. Consider edges in descending order of cost. Delete edge e from T unless doing so would disconnect T.

Prim's algorithm. Start with some root node s and greedily grow a tree T from s outward. At each step, add the cheapest edge e to T that has exactly one endpoint in T.

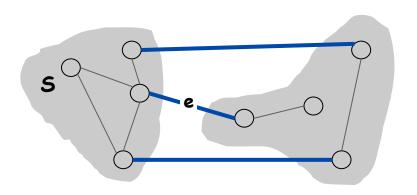
Remark. All three algorithms produce an MST.

## Greedy Algorithms

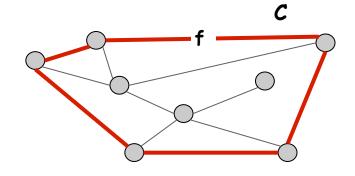
Simplifying assumption. All edge costs  $c_e$  are distinct.

Cut property. Let S be any subset of nodes, and let e be the min cost edge with exactly one endpoint in S. Then the MST contains e.

Cycle property. Let C be any cycle, and let f be the max cost edge belonging to C. Then the MST does not contain f.



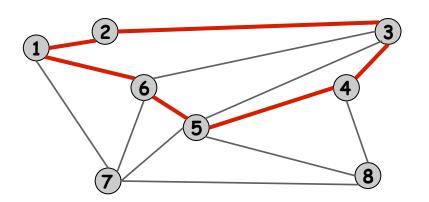
e is in the MST



f is not in the MST

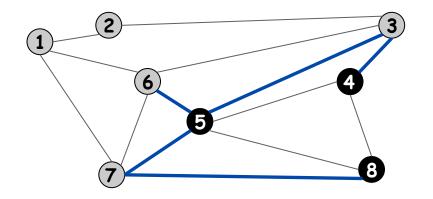
## Cycles and Cuts

Cycle. Set of edges the form a-b, b-c, c-d, ..., y-z, z-a.



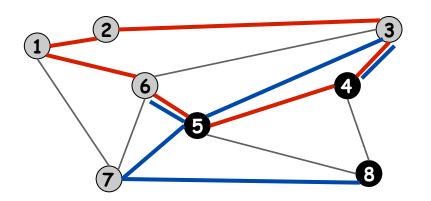
Cycle 
$$C = 1-2, 2-3, 3-4, 4-5, 5-6, 6-1$$

Cutset. A cut is a subset of nodes S. The corresponding cutset D is the subset of edges with exactly one endpoint in S.

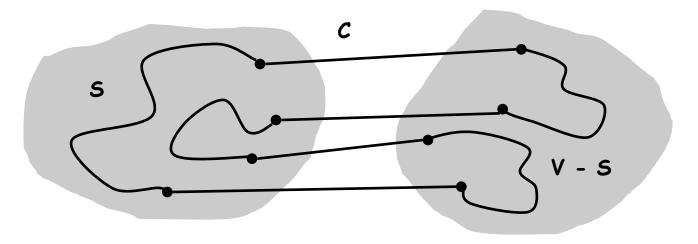


## Cycle-Cut Intersection

Claim. A cycle and a cutset intersect in an even number of edges.



#### Pf. (by picture)



## Greedy Algorithms

Simplifying assumption. All edge costs  $c_e$  are distinct.

Cut property. Let S be any subset of nodes, and let e be the min cost edge with exactly one endpoint in S. Then the MST T\* contains e.

#### Pf. (exchange argument)

- Suppose e does not belong to T\*, and let's see what happens.
- Adding e to T\* creates a cycle C in T\*.
- Edge e is both in the cycle C and in the cutset D corresponding to S  $\Rightarrow$  there exists another edge, say f, that is in both C and D.
- $T' = T^* \cup \{e\} \{f\}$  is also a spanning tree.
- Since  $c_e < c_f$ ,  $cost(T') < cost(T^*)$ .
- This is a contradiction. ■

