

COL702:Advanced Data Structures and Algorithms

Thanks to Miles Jones, Russell Impagliazzo, and Sanjoy Dasgupta at UCSD for these slides.

Paths in graphs

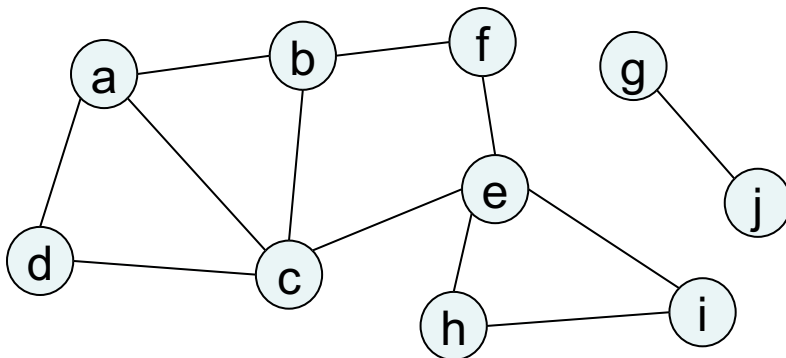
The classic 15-puzzle



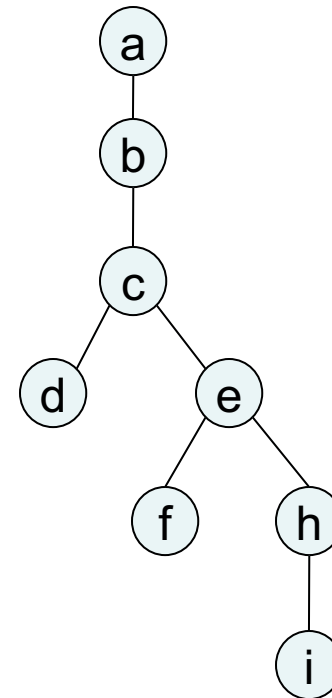
Graph $G = (V, E)$

$V = \{\text{configurations of puzzle}\}$

E : edges between neighboring configurations



explore(G,a):

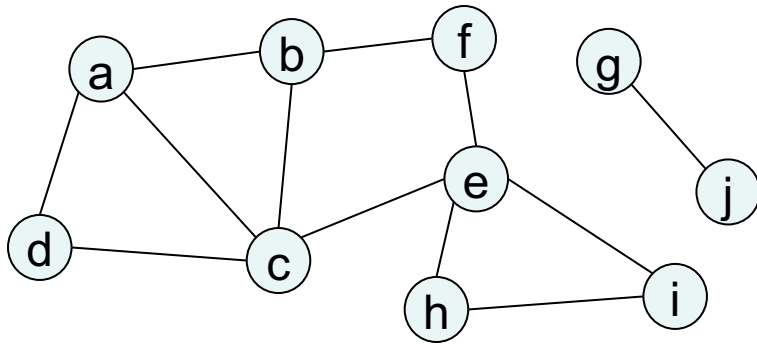


Finds a path from a to i.

But this isn't the shortest possible path!

Distances in graphs

Distance between two nodes
= length of shortest path between them



$\text{dist}(a,e) = ?$
 $\text{dist}(d,g) = ?$

Suppose we want to compute distances from some starting node s to all other nodes in G .

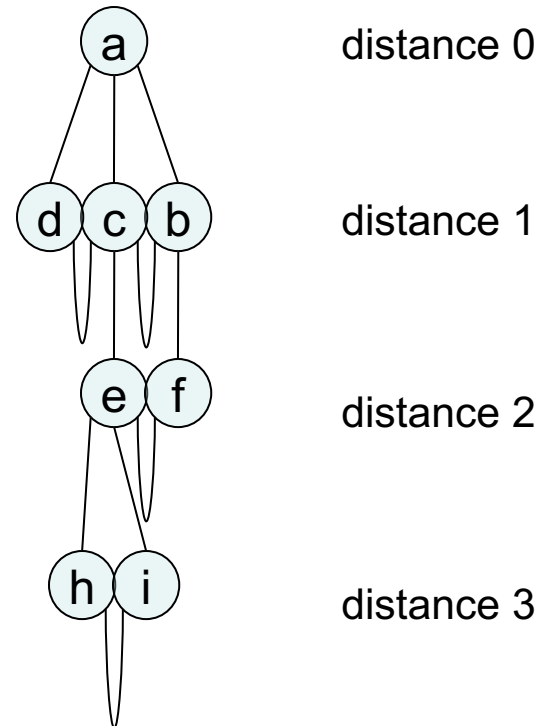
Strategy: layer-by-layer

first, nodes at distance 0

then, nodes at distance 1

then, nodes at distance 2, etc.

Physical model:
Vertex – ping-pong ball
Edge – piece of string



Breadth-first search

Suppose we have seen all nodes at distance $\leq d$.

How to get the next layer?

Solution:

A node is at distance $d+1$ if:

it is adjacent to some node at distance d

it hasn't been seen yet

```
procedure bfs(G,s)
```

```
input: graph  $G = (V,E)$ ; node  $s$  in  $V$   
output: for each node  $u$ ,  $\text{dist}[u]$  is  
set to its distance from  $s$ 
```

```
for  $u$  in  $V$ :
```

```
     $\text{dist}[u] = \infty$ 
```

```
 $\text{dist}[s] = 0$ 
```

```
 $Q = [s]$  // queue containing just  $s$ 
```

```
while  $Q$  is not empty:
```

```
     $u = \text{eject}(Q)$ 
```

```
    for each edge  $(u,v)$  in  $E$ :
```

```
        if  $\text{dist}[v] = \infty$ :
```

```
             $\text{inject}(Q,v)$ 
```

```
             $\text{dist}[v] = \text{dist}[u]+1$ 
```

BFS example

```
procedure bfs(G,s)
```

```
  for u in V:
```

```
    dist[u] =  $\infty$ 
```

```
    prev[u] = nil
```

```
dist[s] = 0
```

```
Q = [s] // queue containing just s
```

```
while Q is not empty:
```

```
  u = eject(Q)
```

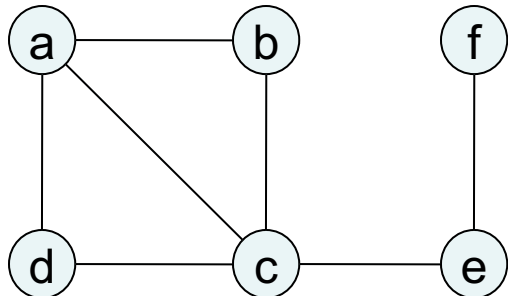
```
  for each edge (u,v) in E:
```

```
    if dist[v] =  $\infty$ :
```

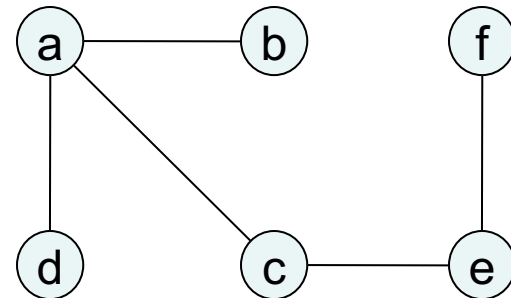
```
      inject(Q,v)
```

```
      dist[v] = dist[u]+1
```

```
      prev[v] = u
```



Queue	Distances					
	a	b	c	d	e	f
[a]	0	∞	∞	∞	∞	∞
[bcd]	0	1	1	1	∞	∞
[cd]	0	1	1	1	∞	∞
[de]	0	1	1	1	2	∞
[e]	0	1	1	1	2	∞
[f]	0	1	1	1	2	3
[]	0	1	1	1	2	3



Shortest path tree

Why does BFS work?

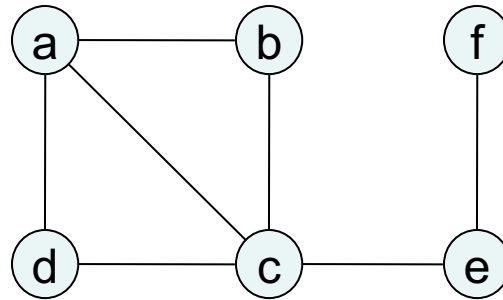
```
procedure bfs(G,s)
  for u in V:
    dist[u] = ∞
  dist[s] = 0
  Q = [s]
  while Q is not empty:
    u = eject(Q)
    for each edge (u,v) in E:
      if dist[v] = ∞:
        inject(Q,v)
        dist[v] = dist[u]+1
```

Claim For any distance $d = 0, 1, 2, \dots$, there is a point in time at which:

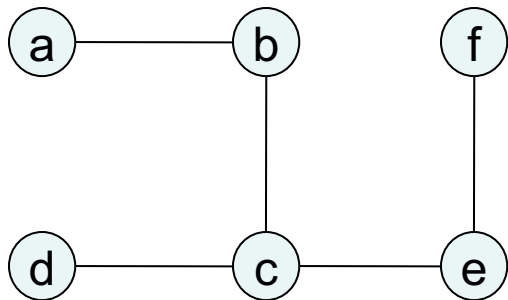
- (i) all nodes at distance $\leq d$ have their `dist[]` values correctly set
- (ii) all other nodes have `dist[] = ∞`
- (iii) the queue `Q` contains exactly the nodes at distance d

Running time: $O(V + E)$, like DFS

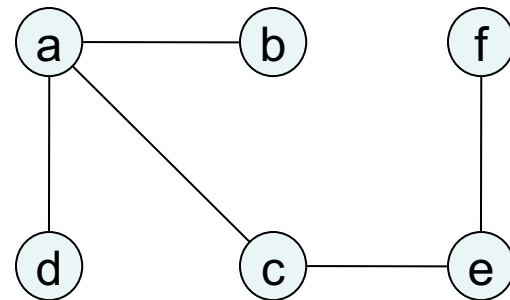
Two search strategies



Depth-first



Breadth-first



Edge lengths

BFS treats all edges as having the same length.
This is rarely true in applications.

Denote the length of edge $e = (u,v)$ by $l(e)$ or l_e or $l(u,v)$



COL351: Slides for Lecture Component 10

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Edge lengths

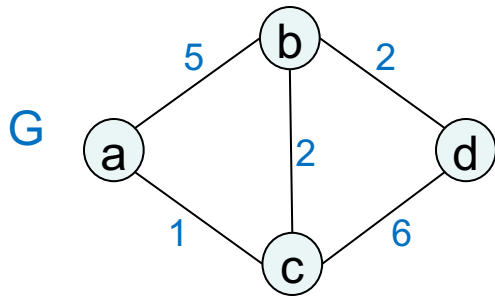
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Denote the length of edge $e = (u,v)$ by $l(e)$ or l_e or $l(u,v)$



Extending BFS

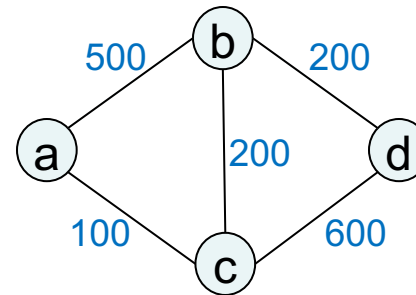
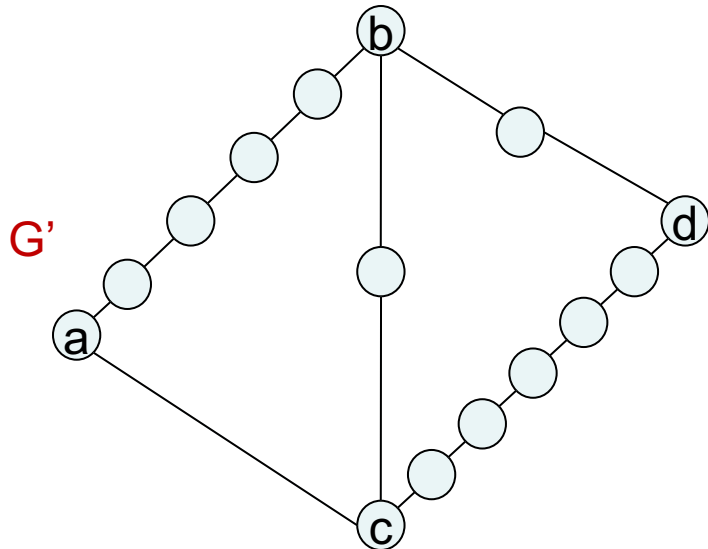
Suppose G has positive integral edge lengths



- (i) G' has unit-length edges
 - (ii) For the “real” nodes, distance in G = distance in G'
- So run BFS on G' !

Problem: efficiency

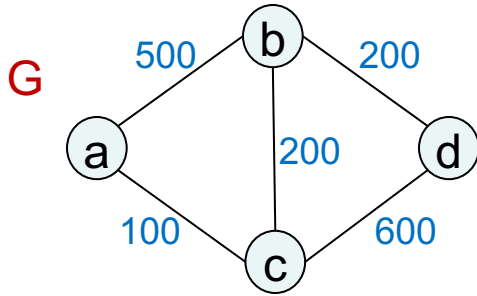
Simple trick: add *dummy nodes*



If edge lengths in G are large:

- (i) G' is enormous
- (ii) BFS wastes a lot of time computing distances to dummy nodes we don't care about

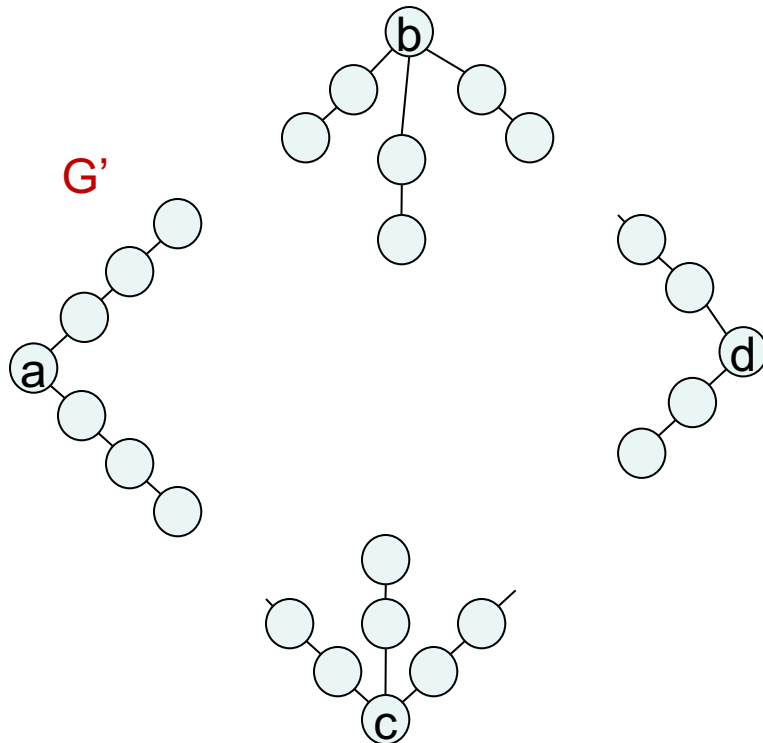
Extending BFS



First 99 time steps: BFS (on G') slowly advances along $a \rightarrow b$ and $a \rightarrow c$. Boring!

Can we snooze and have an **alarm** wake up us whenever BFS reaches a *real* node?

Alarm for each real node: estimated time of arrival based on edges currently being traversed.



- $T = 0$ set alarms for b (500), c (100)
snooze
- $T = 100$ wake up, BFS is at c
set alarms for b (300), d (700)
snooze
- $T = 300$ wake up, BFS is at b
set alarm for d (500)
snooze
- $T = 500$ wake up, BFS is at d

- $\text{dist}[c] = 100$
- $\text{dist}[b] = 300$
- $\text{dist}[d] = 500$

Alarm clock algorithm

(Given graph G and starting node s)

set an alarm for node s at time 0

if the next alarm goes off at time T , for node u :

distance[u] = T

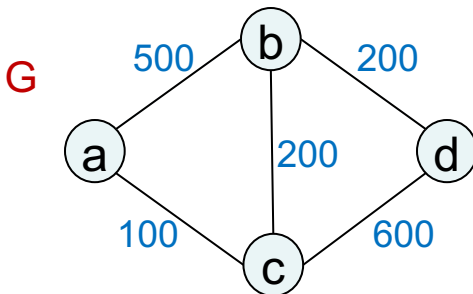
for each edge (u,v) in E :

if no alarm for v , set one for $T + l(u,v)$

if there is an alarm for v , but later than $T + l(u,v)$, then reset to this earlier time

Exactly simulates BFS on G' ...

we no longer need to construct G' !



How to implement alarm?

Answer: priority queue (aka heap)

A priority queue H stores:

- a set of elements (our nodes)

- associated key values (alarm times)

and supports these operations:

insert(H,x)	insert new element into H	set a new alarm
deletemin(H)	return element with smallest key value, remove from H	which alarm is going off next?
decreasekey(H,x)	allow x 's key value to be decreased	allow alarm to be reset to an earlier time
makequeue(S)	make a queue out of the elements in S (and their keys)	initialize alarms

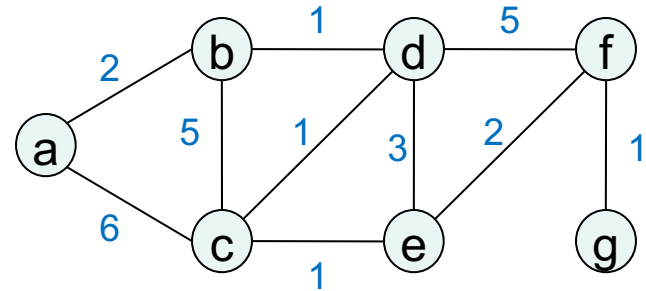
Dijkstra's algorithm

```
procedure dijkstra(G,l,s)
```

```
input: graph  $G = (V,E)$ ; node  $s$ ;  
       positive edge lengths  $l_e$   
output: for each node  $u$ ,  $\text{dist}[u]$  is  
        set to its distance from  $s$ 
```

```
for  $u$  in  $V$ :  
     $\text{dist}[u] = \infty$   
 $\text{dist}[s] = 0$   
 $H = \text{makequeue}(V)$  // key =  $\text{dist}[]$ 
```

```
while  $H$  is not empty:  
     $u = \text{deletemin}(H)$   
    for each edge  $(u,v)$  in  $E$ :  
        if  $\text{dist}[v] > \text{dist}[u] + l(u,v)$ :  
             $\text{dist}[v] = \text{dist}[u] + l(u,v)$   
             $\text{decreasekey}(H,v)$ 
```



Another example

```
procedure dijkstra(G,l,s)
```

```
for u in V:
```

```
  dist[u] =  $\infty$ 
```

```
  prev[u] = nil
```

```
dist[s] = 0
```

```
H = makequeue(V) // key = dist[]
```

```
while H is not empty:
```

```
  u = deletemin(H)
```

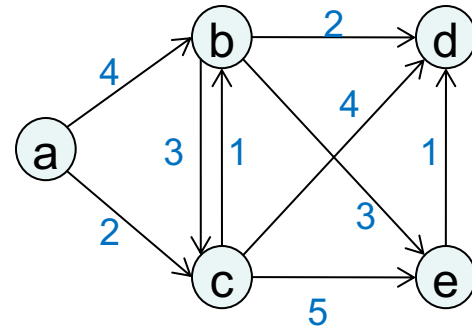
```
  for each edge (u,v) in E:
```

```
    if dist[v] > dist[u] + l(u,v):
```

```
      dist[v] = dist[u] + l(u,v)
```

```
      prev[v] = u
```

```
      decreasekey(H,v)
```



Running time

```
procedure dijkstra(G,l,s)

for u in V:
    dist[u] = ∞
dist[s] = 0
H = makequeue(V) // key = dist[]

while H is not empty:
    u = deletemin(H)
    for each edge (u,v) in E:
        if dist[v] > dist[u] + l(u,v):
            dist[v] = dist[u] + l(u,v)
            decreasekey(H,v)
```

Time:

$O(V + E) +$

$V \times \text{deletemin} +$

$V \times \text{insert} +$

$E \times \text{decreasekey}$

COL351: Slides for Lecture Component 11

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Running time

```
procedure dijkstra(G,l,s)

for u in V:
    dist[u] = ∞
dist[s] = 0
H = makequeue(V) // key = dist[]

while H is not empty:
    u = deletemin(H)
    for each edge (u,v) in E:
        if dist[v] > dist[u] + l(u,v):
            dist[v] = dist[u] + l(u,v)
            decreasekey(H,v)
```

Time:

$O(V + E) +$

$V \times \text{deletemin} +$

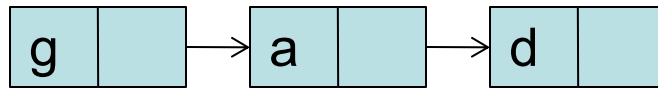
$V \times \text{insert} +$

$E \times \text{decreasekey}$

Depends on priority queue
implementation:
eg. binary heap $O(E \log V)$

Linked list implementation

Linked list, unordered



insert:

decreasekey:

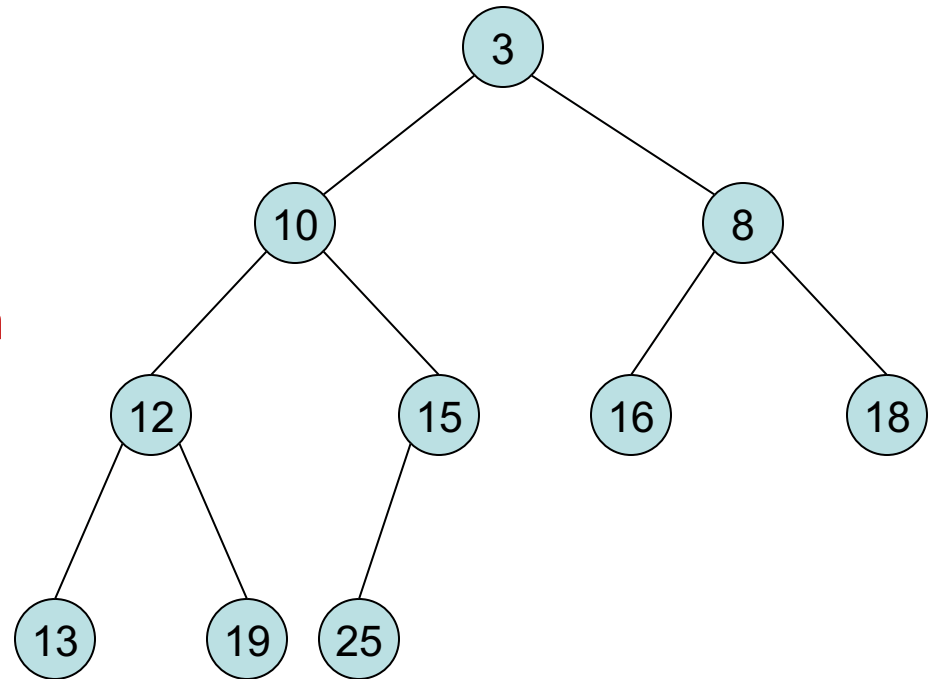
deletemin:

Binary heap

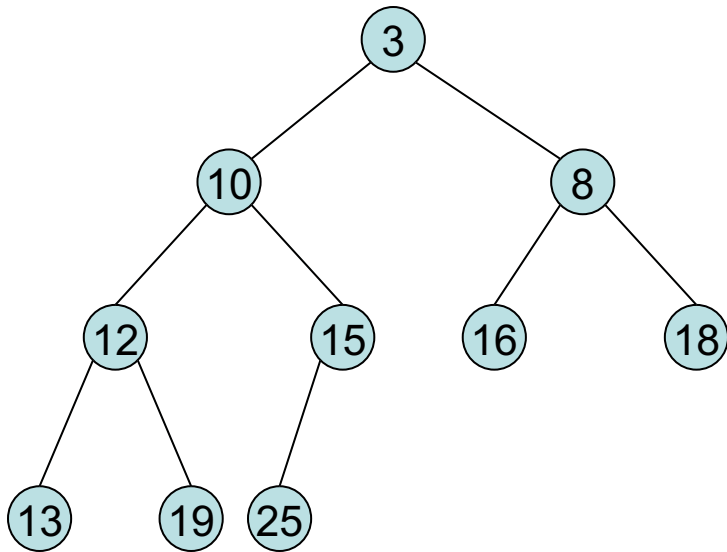
Complete binary tree: filled in row by row, left-to-right

Rule: each node's value is smaller than that of its children

Height $\leq \log_2 n + 1$



Binary heap



insert(7)

decreasekey(19 -> 6)

deletemin

d-ary heap

Same as a binary heap, but with
d children...

height:

insert

deletemin

Running time of Dijkstra's algorithm

	insert, decreasekey	deletemin	V x deletemin + (V+E) x insert
linked list	$O(1)$	$O(V)$	$O(V^2)$
binary heap	$O(\log V)$	$O(\log V)$	$O((V+E) \log V)$
d-ary heap	$O(\log_d V)$	$O(d \log_d V)$	$O((dV + E) \log_d V)$
Fibonacci heap	$O(1)$ amortized	$O(\log V)$	$O(E + V \log V)$

Which is best depends on *sparsity* of graph: ratio E/V (average degree).

Linked list vs. binary heap

Dense graph: $E = \Theta(V^2)$

Linked list is better: $O(V^2)$

Sparse graph: $E = O(V)$

Binary heap is better: $O(V \log V)$

d-ary heap

Best choice $d \approx E/V$

Dense: $O(V^2)$

Sparse: $O(V \log V)$

Intermediate: $E = V^{1+c}$

$O(E/c)$, linear!

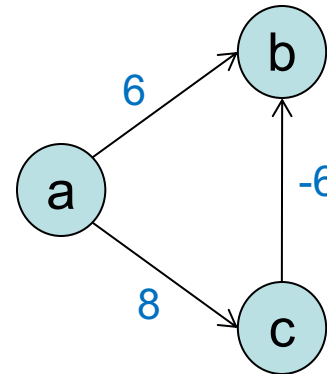
Dijkstra and negative edges

```
procedure dijkstra(G,l,s)
for u in V:
  dist[u] = ∞
dist[s] = 0
H = makequeue(V) // key = dist[]

while H is not empty:
  u = deletemin(H)
  for each edge (u,v) in E:
    if dist[v] > dist[u] + l(u,v):
      dist[v] = dist[u] + l(u,v)
      decreasekey(H,v)
```

Basic principle of Dijkstra's algorithm:
the shortest path to any node only goes through nodes that are closer by.

Not true if negative edges are present!



In Dijkstra's algorithm, dist[] values:

- (i) are never too small
- (ii) get changed only when updating along an edge:

```
procedure update(edge (u,v))
if dist[v] > dist[u] + l(u,v):
  dist[v] = dist[u] + l(u,v)
```