

Data Structures and Algorithms

All Paths Shortest Path (Plus Review)

CS 225

November 6, 2024

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ILLINOIS
URBANA - CHAMPAIGN

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Exam 4 (11/13 — 11/15)

Autograded MC and one coding question

Manually graded short answer prompt

Practice exam on PL

Topics covered can be found on website

Registration started October 31

<https://courses.engr.illinois.edu/cs225/fa2024/exams/>

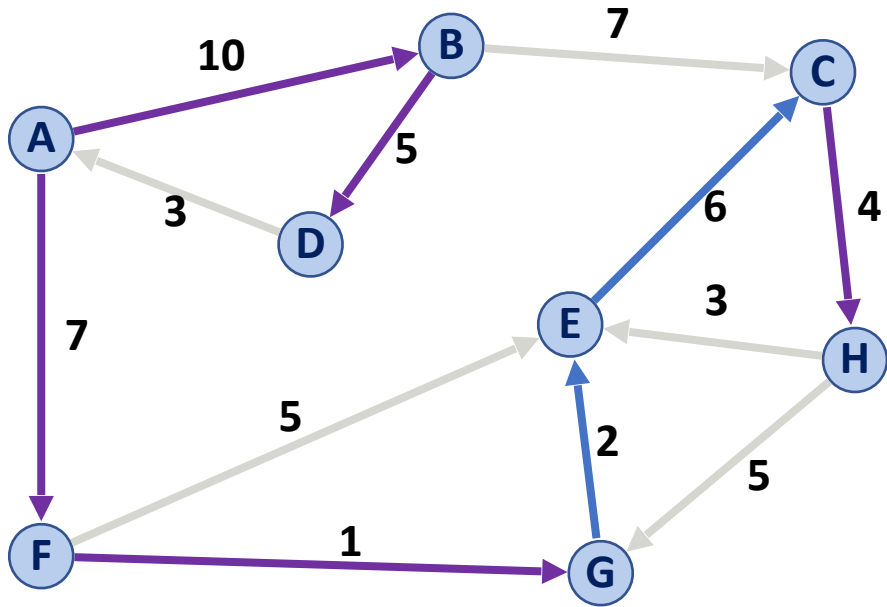
Learning Objectives

Introduce and discuss All-Paths Shortest Path

Review deterministic data structures in CS

An opportunity for Q&A for exam 4

Dijkstra's Algorithm (SSSP)



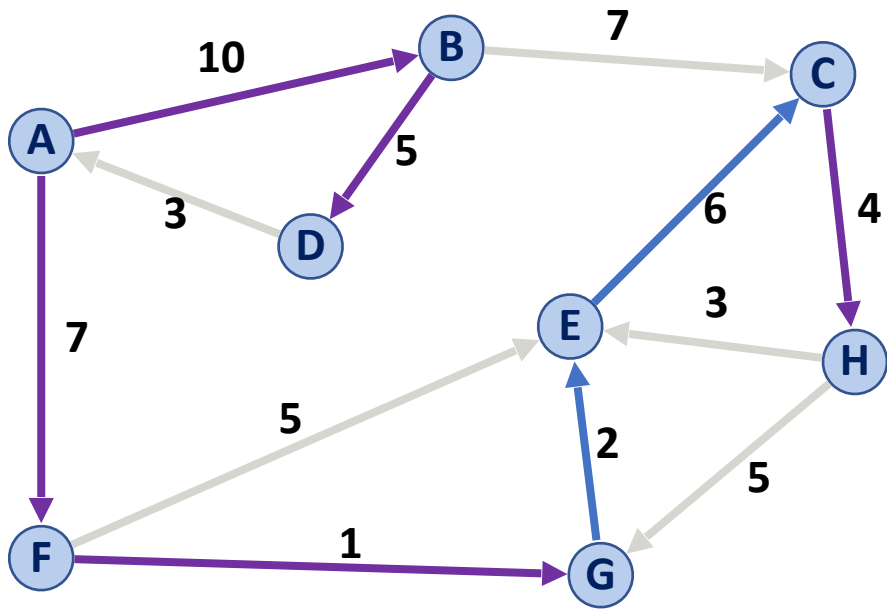
```

DijkstraSSSP(G, s):
6  foreach (Vertex v : G.vertices()):
7    d[v] = +inf
8    p[v] = NULL
9  d[s] = 0
10
11  PriorityQueue Q // min distance, defined by d[v]
12  Q.buildHeap(G.vertices())
13  Graph T          // "labeled set"
14
15  repeat n times:
16    Vertex u = Q.removeMin()
17    T.add(u)
18    foreach (Vertex v : neighbors of u not in T):
19      if cost(u, v) + d[u] < d[v]:
20        d[v] = cost(u, v) + d[u]
21        p[v] = u
  
```

| A | B | C | D | E | F | G | H |
|----|----|----|----|----|---|---|----|
| -- | A | E | B | G | A | F | C |
| 0 | 10 | 16 | 15 | 10 | 7 | 8 | 20 |

Dijkstra's Algorithm (SSSP)

Whats the point of predecessor?



| A | B | C | D | E | F | G | H |
|----|----|----|----|----|---|---|----|
| -- | A | E | B | G | A | F | C |
| 0 | 10 | 16 | 15 | 10 | 7 | 8 | 20 |



Dijkstra's Algorithm (SSSP)

Dijkstra's Algorithm works only on non-negative weights

Optimal implementation:

Fibonacci Heap

If dense, unsorted list ties

Optimal runtime:

Sparse: $O(m + n \log n)$

Dense: $O(n^2)$

```
DijkstraSSSP(G, s):
6  foreach (Vertex v : G):
7      d[v] = +inf
8      p[v] = NULL
9  d[s] = 0
10
11  PriorityQueue Q // min distance, defined by d[v]
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19          if cost(u, v) + d[u] < d[v]:
20              d[v] = cost(u, v) + d[u]
21              p[v] = u
22
23  return T
```

Floyd-Warshall Algorithm

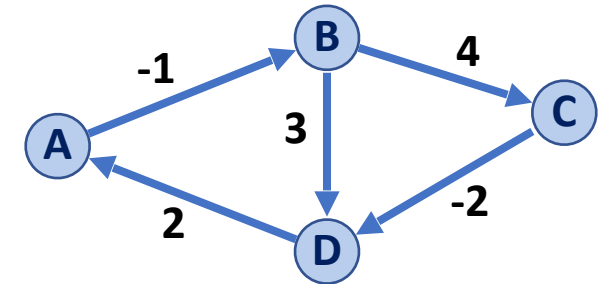
Floyd-Warshall's Algorithm is an alternative to Dijkstra in the presence of **negative-weight edges (not negative weight cycles)**.

```
1 FloydWarshall(G):
2   Let d be a adj. matrix initialized to +inf
3   foreach (Vertex v : G):
4     d[v][v] = 0
5   foreach (Edge (u, v) : G):
6     d[u][v] = cost(u, v)
7
8   foreach (Vertex u : G):
9     foreach (Vertex v : G):
10      foreach (Vertex w : G):
11        if (d[u, v] > d[u, w] + d[w, v])
12          d[u, v] = d[u, w] + d[w, v]
```

Floyd-Warshall Algorithm

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1 FloydWarshall(G):  
2   Let d be a adj. matrix initialized to +inf  
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5   foreach (Edge (u, v) : G):  
6     d[u][v] = cost(u, v)
```

| | A | B | C | D |
|---|---|---|---|---|
| A | | | | |
| B | | | | |
| C | | | | |
| D | | | | |

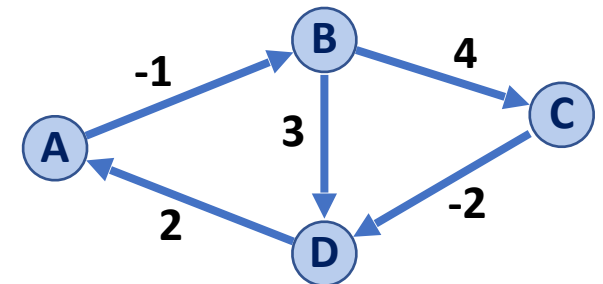


Floyd-Warshall Algorithm

```
8  foreach (Vertex w : G):
9    foreach (Vertex u : G):
10   foreach (Vertex v : G):
11     if (d[u, v] > d[u, w] + d[w, v])
12       d[u, v] = d[u, w] + d[w, v]
```

Let us consider comparisons where $w = A$:

| | A | B | C | D |
|---|----------|----------|----------|----------|
| A | 0 | -1 | ∞ | ∞ |
| B | ∞ | 0 | 4 | 3 |
| C | ∞ | ∞ | 0 | -2 |
| D | 2 | ∞ | ∞ | 0 |



Floyd-Warshall Algorithm

```
8  foreach (Vertex w : G) :
9  foreach (Vertex u : G) :
10  foreach (Vertex v : G) :
11  if (d[u, v] > d[u, w] + d[w, v])
12  d[u, v] = d[u, w] + d[w, v]
```

Let **w** be midpoint

Let **u** be start point

Let **v** be end point

Is our distance shorter now?

Let us consider comparisons where $w = A$:

$u=A, v=A$

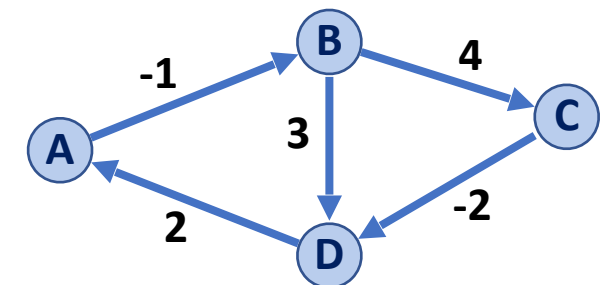


$u=A, v=B$



Don't waste time if $u=w$ or $v=w$!

| | A | B | C | D |
|---|----------|----------|----------|----------|
| A | 0 | -1 | ∞ | ∞ |
| B | ∞ | 0 | 4 | 3 |
| C | ∞ | ∞ | 0 | -2 |
| D | 2 | ∞ | ∞ | 0 |



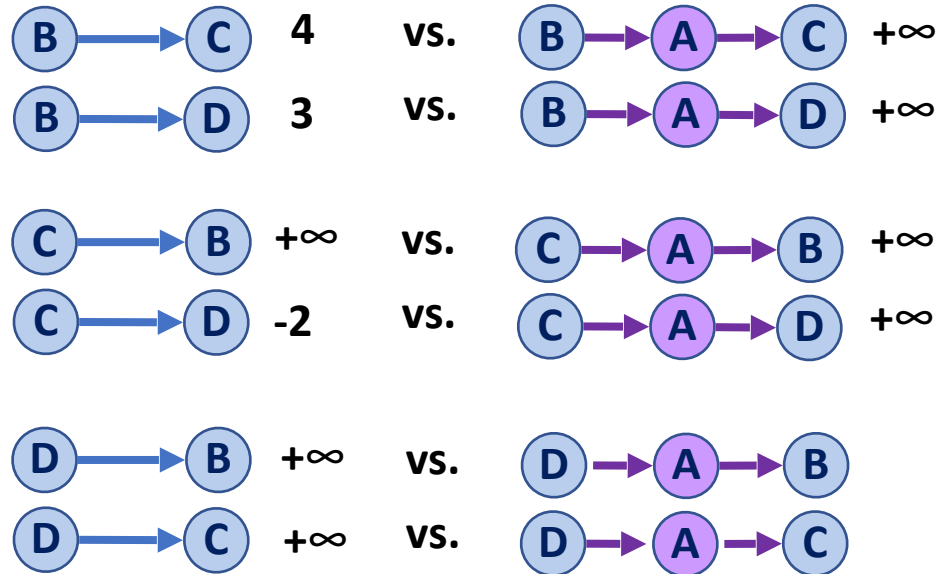
Floyd-Warshall Algorithm

Let **w** be midpoint
 Let **u** be start point
 Let **v** be end point
 Is our distance shorter now?

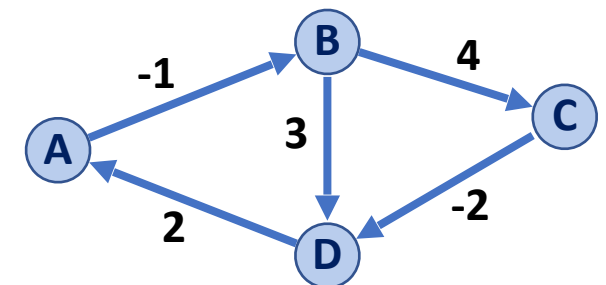
```

8   foreach (Vertex w : G) :
9     foreach (Vertex u : G) :
10    foreach (Vertex v : G) :
11      if (d[u, v] > d[u, w] + d[w, v])
12        d[u, v] = d[u, w] + d[w, v]
    
```

Let us consider $w = A$ (and $u \neq w$ and $v \neq w$):



| | A | B | C | D |
|---|----------|----------|----------|----------|
| A | 0 | -1 | ∞ | ∞ |
| B | ∞ | 0 | 4 | 3 |
| C | ∞ | ∞ | 0 | -2 |
| D | 2 | ∞ | ∞ | 0 |



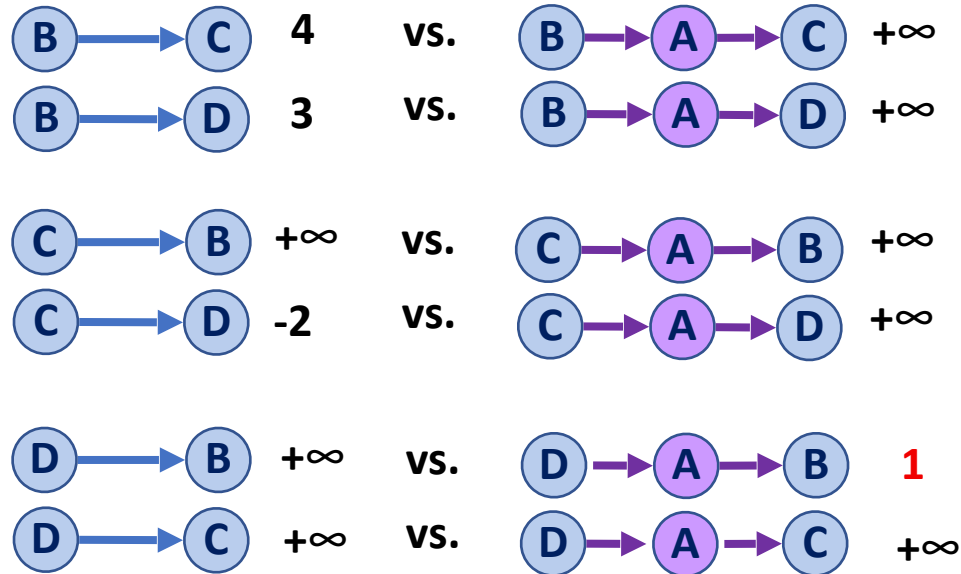
Floyd-Warshall Algorithm

Let **w** be midpoint
 Let **u** be start point
 Let **v** be end point
 Is our distance shorter now?

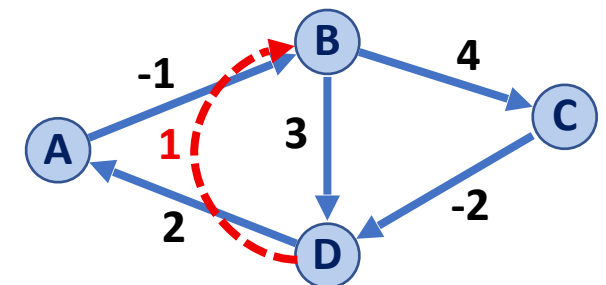
```

8   foreach (Vertex w : G) :
9     foreach (Vertex u : G) :
10    foreach (Vertex v : G) :
11      if (d[u, v] > d[u, w] + d[w, v])
12        d[u, v] = d[u, w] + d[w, v]
    
```

Let us consider $w = A$ (and $u \neq w$ and $v \neq w$):



| | A | B | C | D |
|---|----------|----------|----------|----------|
| A | 0 | -1 | ∞ | ∞ |
| B | ∞ | 0 | 4 | 3 |
| C | ∞ | ∞ | 0 | -2 |
| D | 2 | 1 | ∞ | 0 |



Floyd-Warshall Algorithm

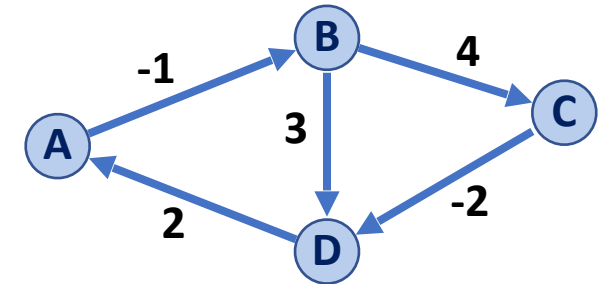
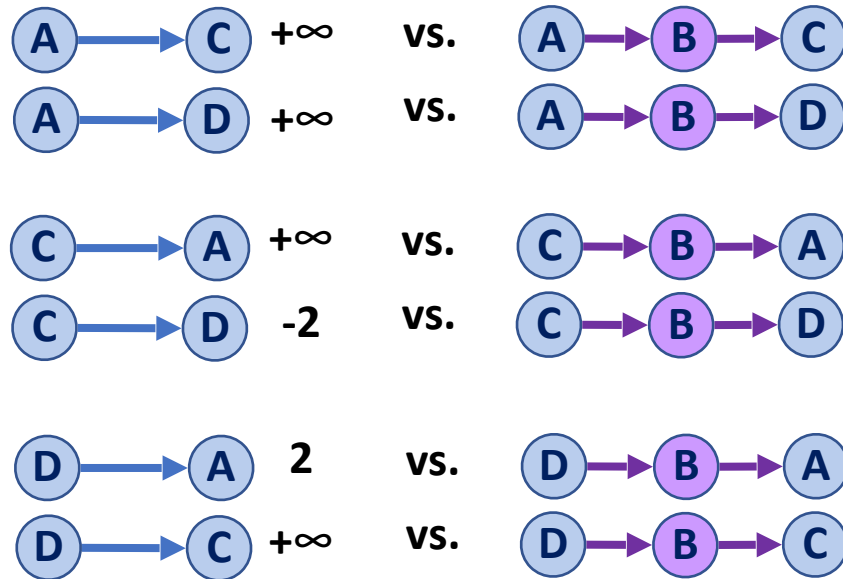
```

8   foreach (Vertex w : G) :
9     foreach (Vertex u : G) :
10    foreach (Vertex v : G) :
11      if (d[u, v] > d[u, w] + d[w, v])
12        d[u, v] = d[u, w] + d[w, v]

```

| | A | B | C | D |
|---|----------|----------|----------|----------|
| A | 0 | -1 | ∞ | ∞ |
| B | ∞ | 0 | 4 | 3 |
| C | ∞ | ∞ | 0 | -2 |
| D | 2 | 1 | ∞ | 0 |

Let us consider $w = B$ (and $u \neq w$ and $v \neq w$):



Floyd-Warshall Algorithm

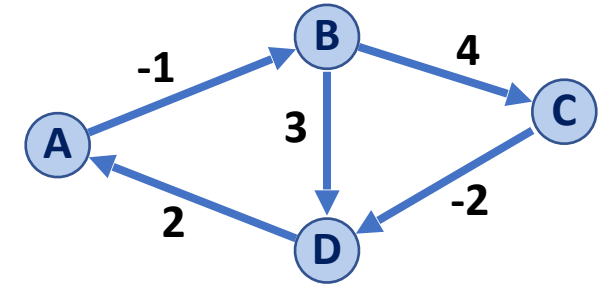
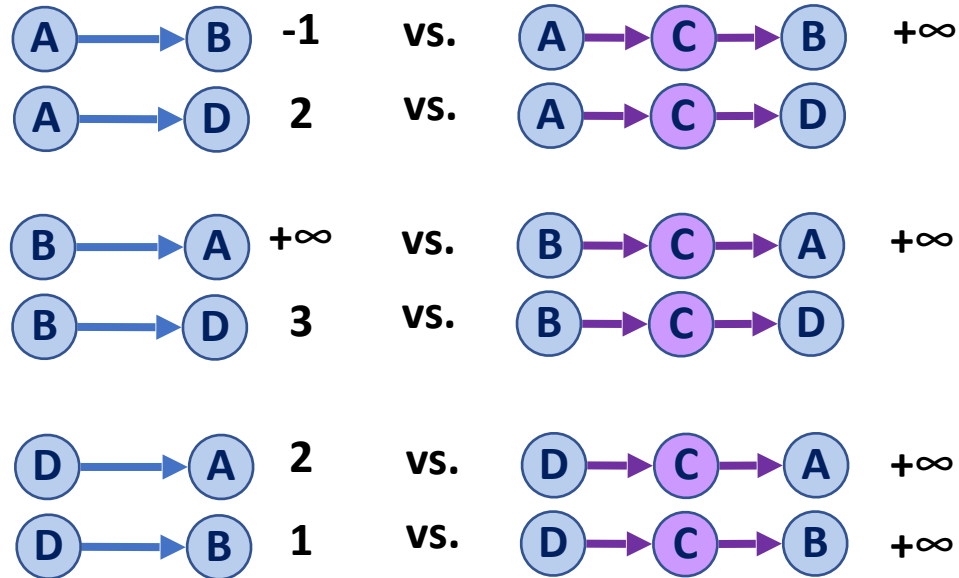
```

8   foreach (Vertex w : G) :
9     foreach (Vertex u : G) :
10    foreach (Vertex v : G) :
11      if (d[u, v] > d[u, w] + d[w, v])
12        d[u, v] = d[u, w] + d[w, v]

```

| | A | B | C | D |
|---|----------|----------|---|----|
| A | 0 | -1 | 3 | 2 |
| B | ∞ | 0 | 4 | 3 |
| C | ∞ | ∞ | 0 | -2 |
| D | 2 | 1 | 5 | 0 |

Let us consider $w = C$ (and $u \neq w$ and $v \neq w$):

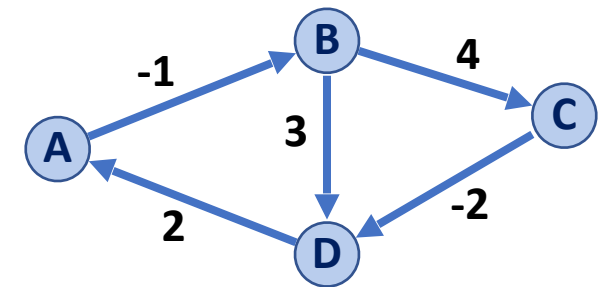


Floyd-Warshall Algorithm



```
1 FloydWarshall(G):
2   Let d be a adj. matrix initialized to +inf
3   foreach (Vertex v : G):
4     d[v][v] = 0
5   foreach (Edge (u, v) : G):
6     d[u][v] = cost(u, v)
7
8   foreach (Vertex u : G):
9     foreach (Vertex v : G):
10      foreach (Vertex w : G):
11        if (d[u, v] > d[u, w] + d[w, v])
12          d[u, v] = d[u, w] + d[w, v]
```

| | A | B | C | D |
|---|---|----|---|----|
| A | 0 | -1 | 3 | 1 |
| B | 5 | 0 | 4 | 2 |
| C | 0 | -1 | 0 | -2 |
| D | 2 | 1 | 5 | 0 |



Floyd-Warshall Algorithm

Running time?

```
FloydWarshall(G) :  
6   Let d be a adj. matrix initialized to +inf  
7   foreach (Vertex v : G) :  
8     d[v][v] = 0  
9   foreach (Edge (u, v) : G) :  
10    d[u][v] = cost(u, v)  
11  
12  foreach (Vertex u : G) :  
13    foreach (Vertex v : G) :  
14      foreach (Vertex w : G) :  
15        if d[u, v] > d[u, w] + d[w, v] :  
16          d[u, v] = d[u, w] + d[w, v]
```

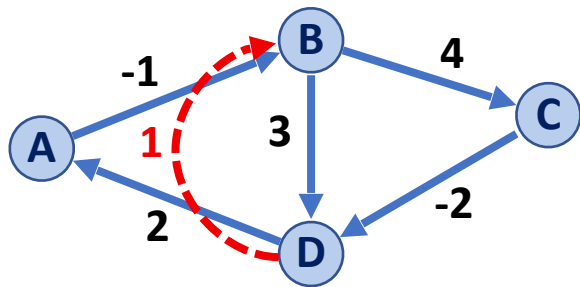

Floyd-Warshall Algorithm

We aren't storing path information! Can we fix this?

```
FloydWarshall(G) :  
6   Let d be a adj. matrix initialized to +inf  
7   foreach (Vertex v : G) :  
8     d[v][v] = 0  
9   foreach (Edge (u, v) : G) :  
10    d[u][v] = cost(u, v)  
11  
12  foreach (Vertex w : G) :  
13    foreach (Vertex u : G) :  
14      foreach (Vertex v : G) :  
15        if (d[u, v] > d[u, w] + d[w, v])  
16          d[u, v] = d[u, w] + d[w, v]
```

Floyd-Warshall Algorithm

```
FloydWarshall(G):  
6   Let d be a adj. matrix initialized to +inf  
7   foreach (Vertex v : G):  
8     d[v][v] = 0  
9     s[v][v] = 0  
10  foreach (Edge (u, v) : G):  
11    d[u][v] = cost(u, v)  
12    s[u][v] = v  
13  
14  foreach (Vertex w : G):  
15    foreach (Vertex u : G):  
16      foreach (Vertex v : G):  
17        if (d[u, v] > d[u, w] + d[w, v])  
18          d[u, v] = d[u, w] + d[w, v]  
19          s[u, v] = s[u, w]
```



| | A | B | C | D |
|---|----------|----------|----------|----------|
| A | 0 | -1 | ∞ | ∞ |
| B | ∞ | 0 | 4 | 3 |
| C | ∞ | ∞ | 0 | -2 |
| D | 2 | 1 | ∞ | 0 |

| | A | B | C | D |
|---|---|---|---|---|
| A | | B | | |
| B | | | C | D |
| C | | | | D |
| D | A | | | |

We have only scratched the surface on graphs!

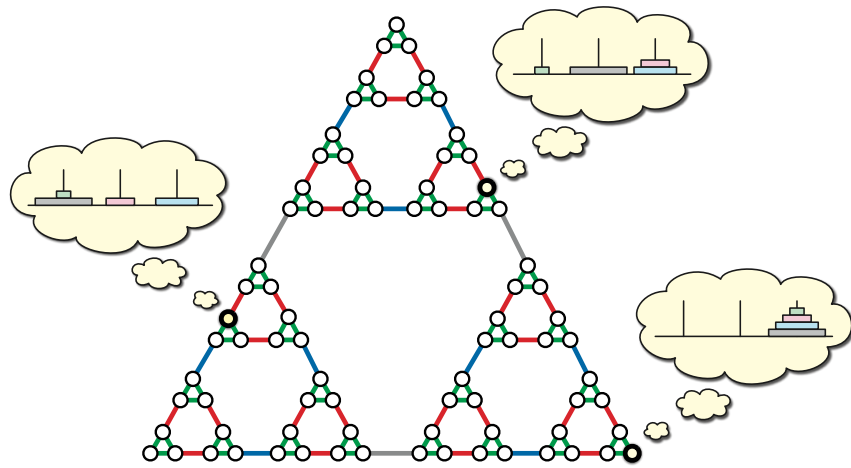


Image from Jeff Erickson Algorithms First Edition

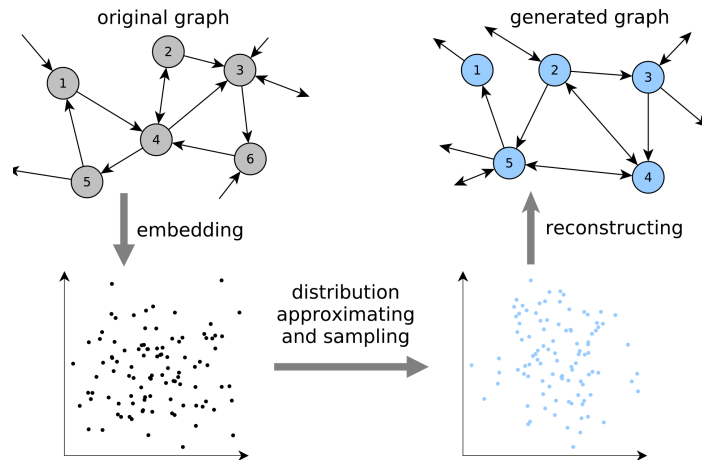
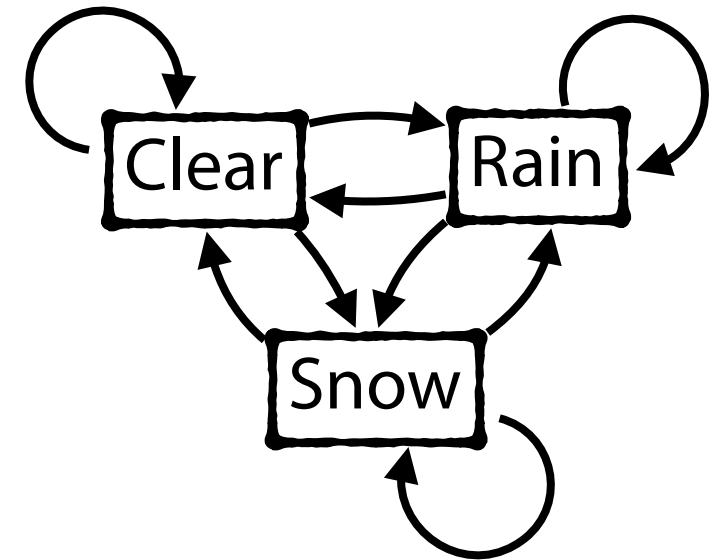


Image from Drobyshevskiy et al. **Random graph modeling: A survey of the concepts.** 2019



$$M = \begin{pmatrix} .5 & .3 & .2 \\ .5 & .4 & .1 \\ .2 & .1 & .7 \end{pmatrix}$$



Lets review what we've seen so far!



Lets review what we've seen so far!

Its arrays all the way down.

Lists



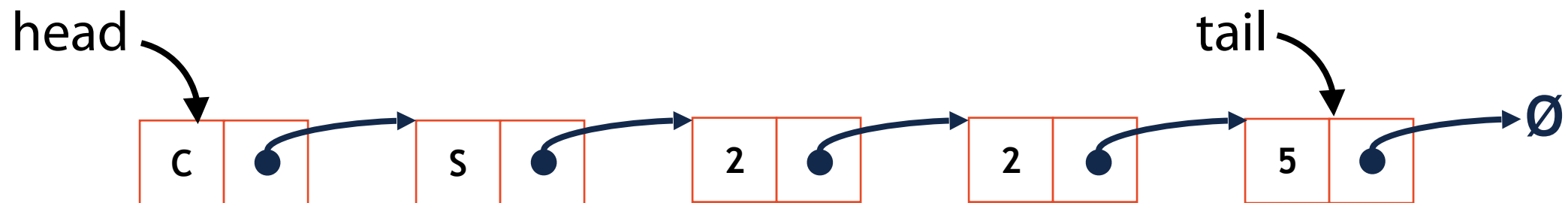
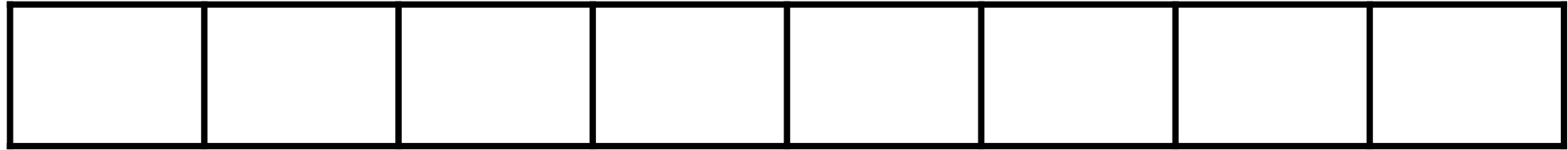
The not-so-secret underlying implementation for many things

| | Singly Linked List | Array |
|-------------------------------------|--------------------|--------|
| Look up arbitrary location | $O(n)$ | $O(1)$ |
| Insert after given element | $O(1)$ | $O(n)$ |
| Remove after given element | $O(1)$ | $O(n)$ |
| Insert at arbitrary location | $O(n)$ | $O(n)$ |
| Remove at arbitrary location | $O(n)$ | $O(n)$ |
| Search for an input value | $O(n)$ | $O(n)$ |

Special Cases:

Stack and Queue

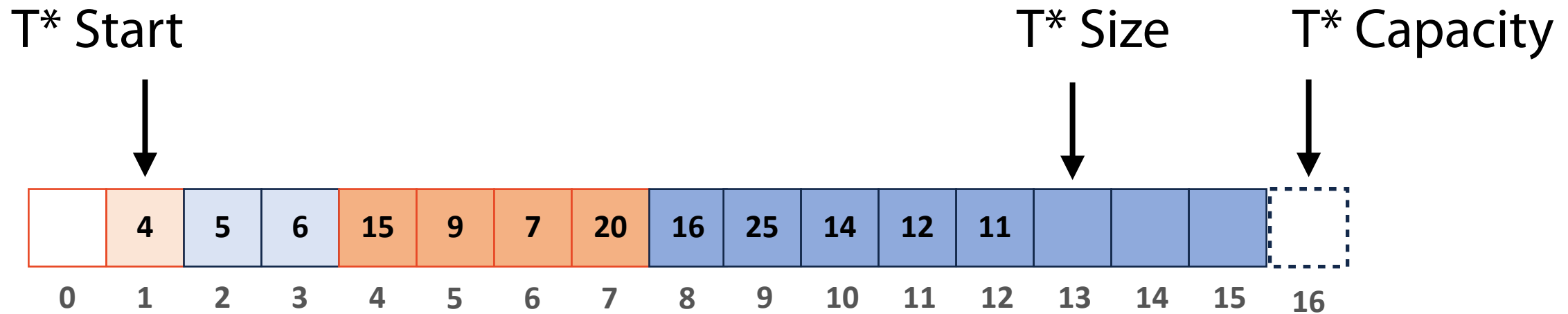
Taking advantage of special cases in lists / arrays



Heap

Taking advantage of special cases in lists / arrays

Array List (Pointer implementation)



↑
size_t Start

↑
size_t Capacity

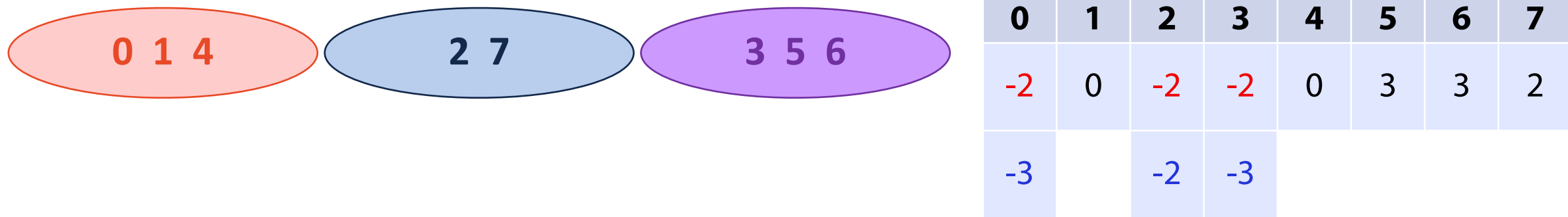
Array List (Index implementation)

↑
size_t Size

Disjoint Set Implementation

Taking advantage of array lookup operations

Store an UpTree as an array, canonical items store **height** / **size**

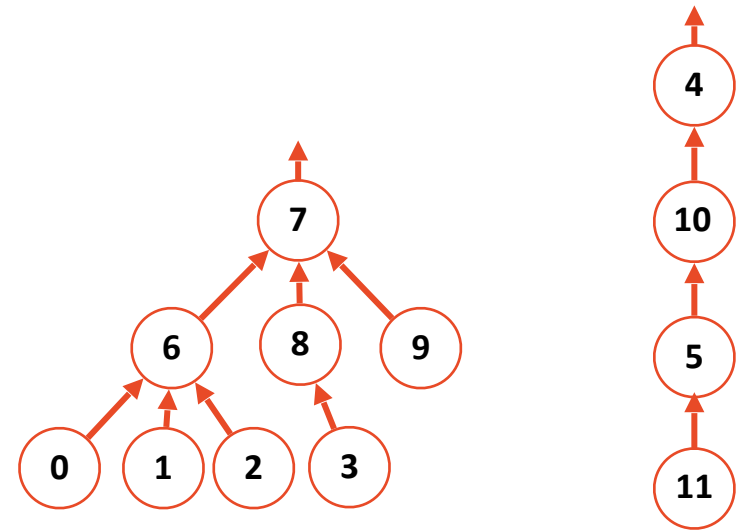


Find(k): Repeatedly look up values until **negative value**

Union(k_1, k_2): Update *smaller* canonical item to point to larger
Update value of remaining canonical item

Disjoint Sets – Smart Union

Minimizing number of $O(1)$ operations



Union by height

| 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 |
|---|---|---|---|----|----|---|---|---|---|----|----|
| 6 | 6 | 6 | 8 | -4 | 10 | 7 | 4 | 7 | 7 | 4 | 5 |

Idea: Keep the height of the tree as small as possible.

Union by size

| 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 |
|---|---|---|---|---|----|---|-----|---|---|----|----|
| 6 | 6 | 6 | 8 | 7 | 10 | 7 | -12 | 7 | 7 | 4 | 5 |

Idea: Minimize the number of nodes that increase in height

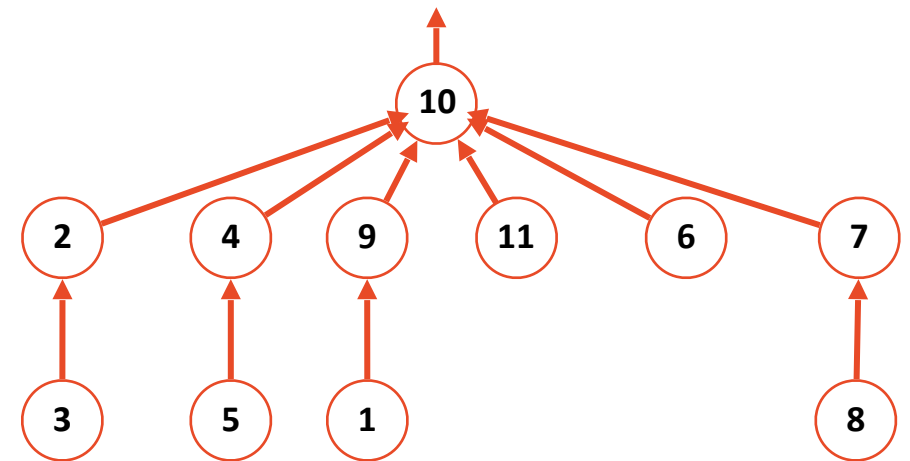
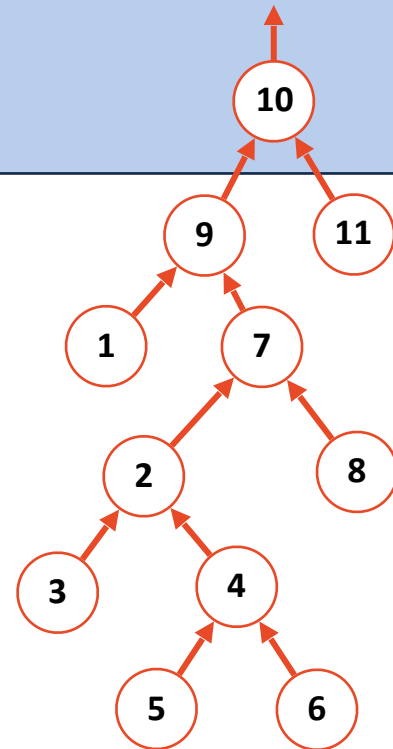
Both guarantee the height of the tree is: $O(\log n)$.

Disjoint Sets Path Compression

Find(6)

Minimizing number of $O(1)$ operations

```
1 int DisjointSets::find(int i) {  
2   if ( s[i] < 0 ) { return i; }  
3   else {  
4     int root = find( s[i] );  
5     s[i] = root;  
6     return root;  
7   }  
8 }
```

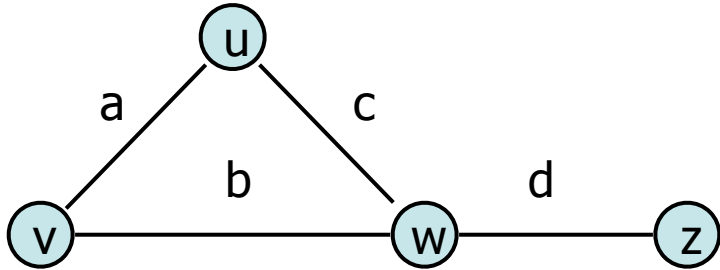


Alternative Not-Actually-A-Proof

Unproven Claim: A disjoint set implemented with smart union and path compression with **m** find calls and **n** items has a worst case running time of **inverse Ackerman**. $[O(m \alpha(n))]$

This grows *very* slowly to the point of being treated a constant in CS.

Graph Implementation: Edge List $|V| = n, |E| = m$

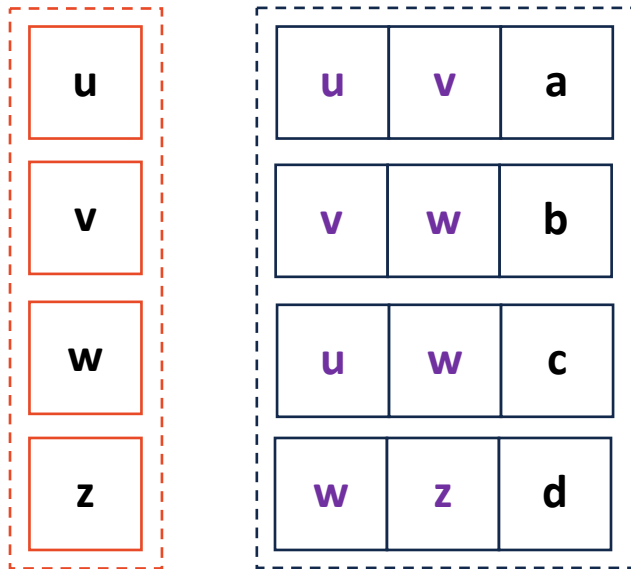


Literally just arrays

$O(1)^*$

insertVertex(K key):

insertEdge(Vertex v1, Vertex v2, K key):



$O(m)$

removeVertex(Vertex v):

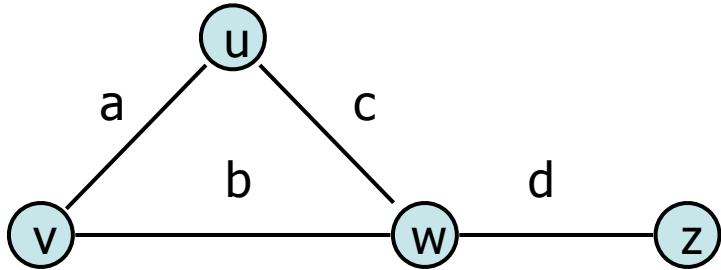
removeEdge(Vertex v1, Vertex v2, K key):

incidentEdges(Vertex v):

areAdjacent(Vertex v1, Vertex v2):

Graph Implementation: Adjacency Matrix

$$|V| = n, |E| = m$$



Literally just a matrix of arrays

$O(1)$

insertEdge(Vertex v1, Vertex v2, K key):

removeEdge(Vertex v1, Vertex v2, K key):

areAdjacent(Vertex v1, Vertex v2):

$O(n)$

incidentEdges(Vertex v):

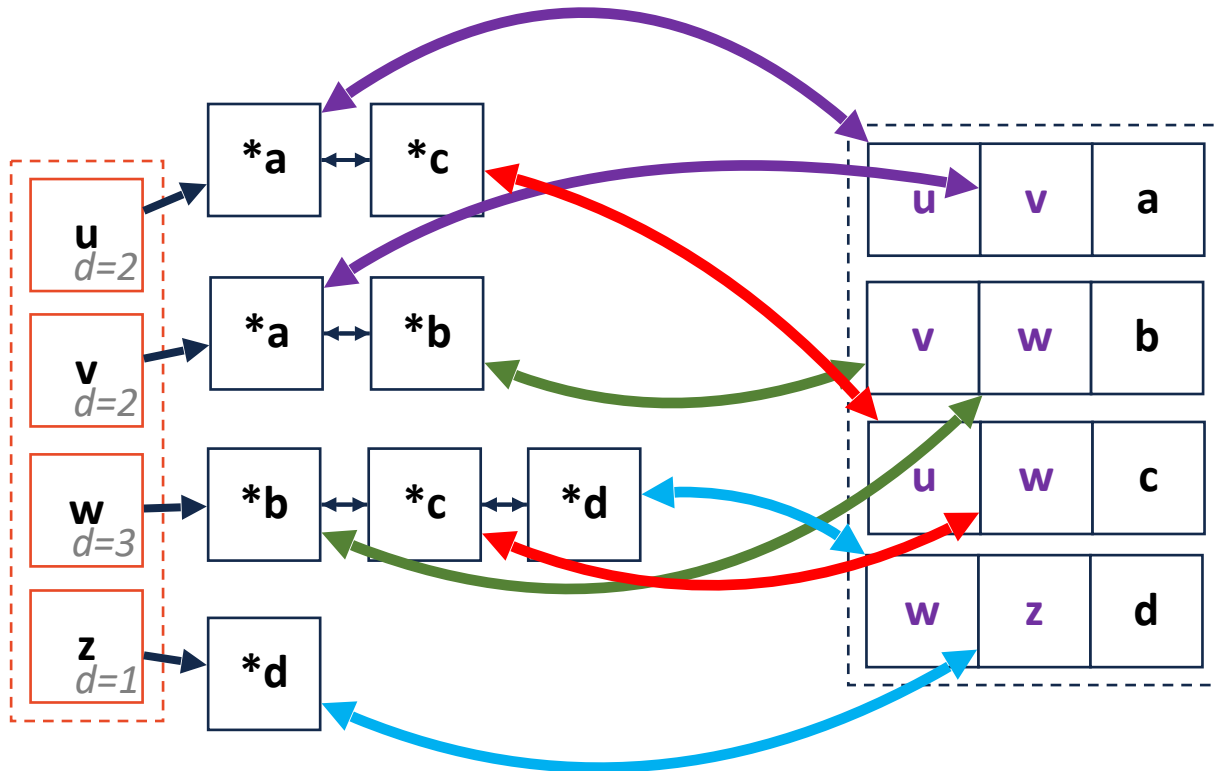
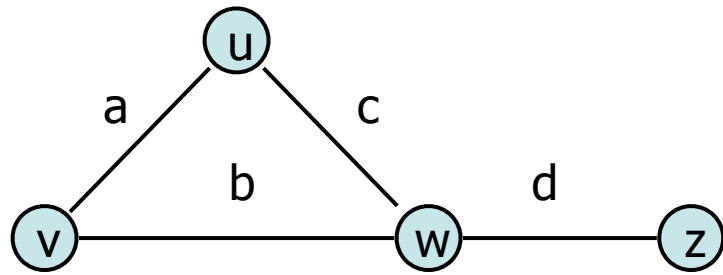
$O(n) \text{---} O(n^2)$

insertVertex(K key):

removeVertex(Vertex v):

| | u | v | w | z |
|---|---|---|---|---|
| u | - | a | c | 0 |
| v | | - | b | 0 |
| w | | | - | d |
| z | | | | - |

Adjacency List



Technically linked lists I guess

| Expressed as $O(f)$ | Adjacency List |
|---------------------|--------------------------------------|
| Space | $n+m$ |
| insertVertex(v) | 1^* |
| removeVertex(v) | $\text{deg}(v)$ |
| insertEdge(u, v) | 1^* |
| removeEdge(u, v) | $\min(\text{deg}(u), \text{deg}(v))$ |
| incidentEdges(v) | $\text{deg}(v)$ |
| areAdjacent(u, v) | $\min(\text{deg}(u), \text{deg}(v))$ |

... And thats most of exam 4

Randomized Algorithms

A **randomized algorithm** is one which uses a source of randomness somewhere in its implementation.

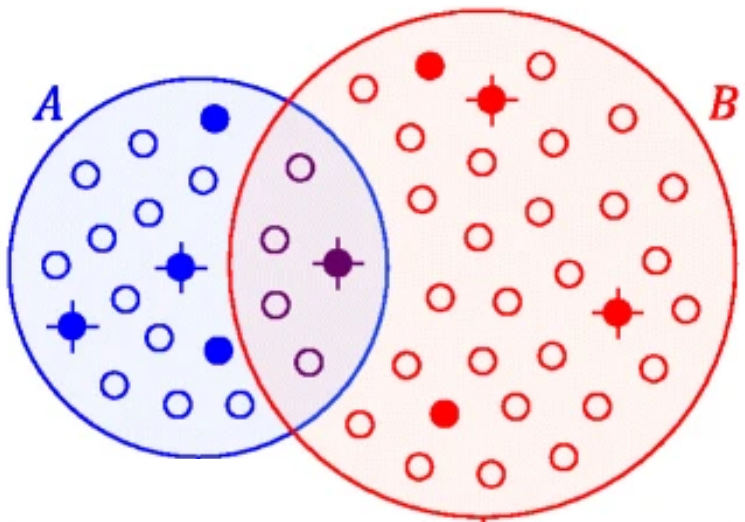
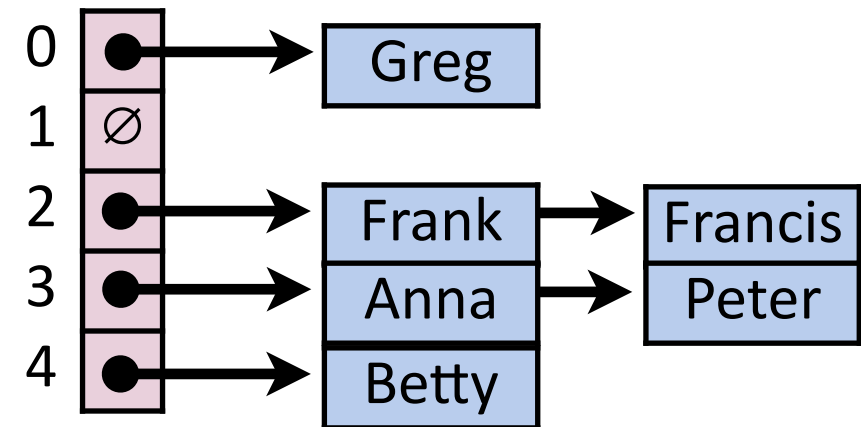
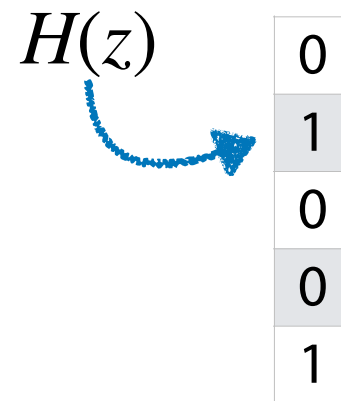


Figure from Ondov et al 2016



| | | | | | | | | | | |
|--------|---|---|---|---|---|---|---|---|---|---|
| $H(x)$ | 0 | 2 | 1 | 0 | 0 | 4 | 0 | 2 | 0 | 6 |
| $H(y)$ | 1 | 0 | 2 | 3 | 1 | 0 | 3 | 4 | 0 | 1 |
| $H(z)$ | 2 | 1 | 0 | 2 | 0 | 1 | 0 | 0 | 7 | 2 |

A faulty list

Imagine you have a list ADT implementation ***except***...

Every time you called **insert**, it would fail 50% of the time.

Quick Primes with Fermat's Primality Test

If p is prime and a is not divisible by p , then $a^{p-1} \equiv 1 \pmod{p}$

But... ***sometimes*** if n is composite and $a^{n-1} \equiv 1 \pmod{n}$

Probabilistic Accuracy: Fermat primality test

| | $a^{p-1} \equiv 1 \pmod{p}$ | $a^{p-1} \not\equiv 1 \pmod{p}$ |
|------------------|-----------------------------|---------------------------------|
| p is prime | | |
| p is not prime | | |

Probabilistic Accuracy: Fermat primality test

Let's assume $\alpha = .5$

First trial: $a = a_0$ and prime test returns 'prime!'

Second trial: $a = a_1$ and prime test returns 'prime!'

Third trial: $a = a_2$ and prime test returns 'not prime!'

Is our number prime?

What is our **false positive** probability? Our **false negative** probability?

Probabilistic Accuracy: Fermat primality test



Summary: Randomized algorithms can also have fixed (or bounded) runtimes at the cost of probabilistic accuracy.

Randomness:

Assumptions: