

Algorithms and Data Structures for Data Science

Graph Implementations 3

CS 277

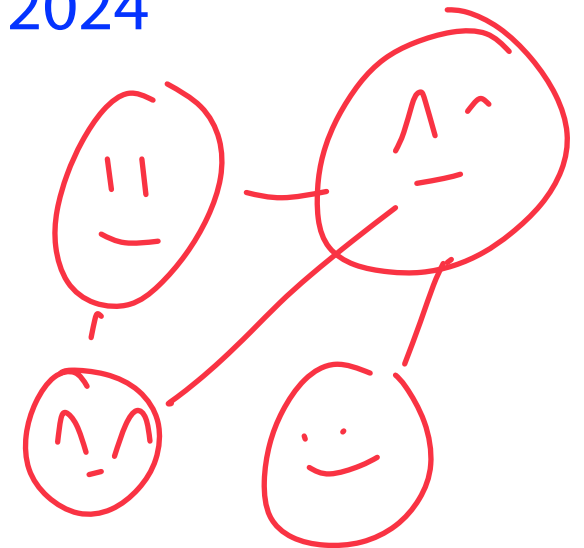
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Learning Objectives

Practice implementing complex data structures (graphs)

Compare and contrast different implementations

Review Big O concepts in the context of graphs

Graph ADT

Find

`getVertices()` — return the list of vertices in a graph

`getEdges(v)` — return the list of edges that touch the vertex v

`areAdjacent(u, v)` — returns a bool based on if an edge from u to v exists

Insert

`insertVertex(v)` — adds a vertex to the graph

`insertEdge(u, v)` — adds an edge to the graph

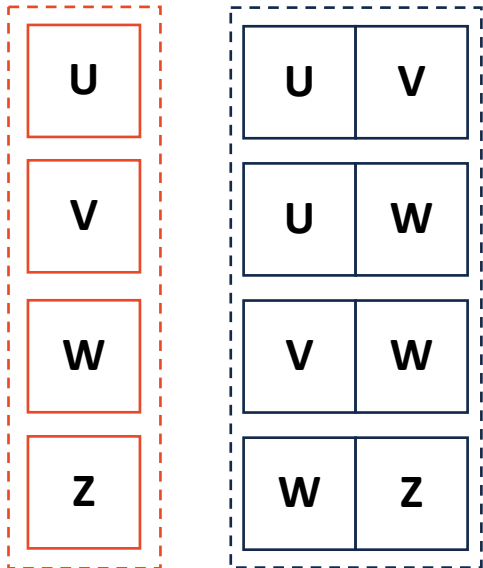
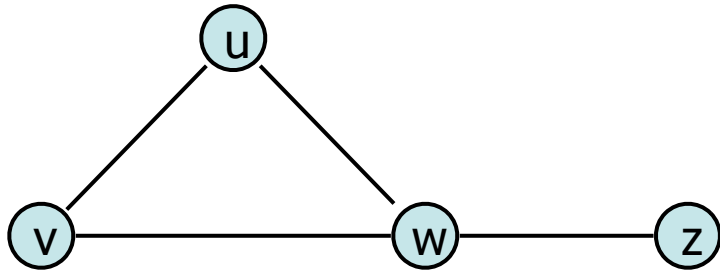
Remove

`removeVertex(v)` — removes a vertex from the graph

`removeEdge(u, v)` — removes an edge from the graph

Graph Implementation: Edge List

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



Very good
at inserting
new items

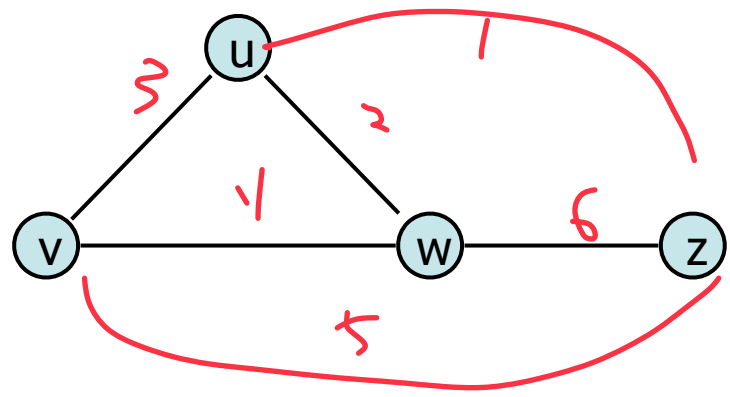
cost
bad at lookup

Expressed as $O(f)$	Edge List
Space	$n+m$
insertVertex(v)	1^*
removeVertex(v)	$n+m$
insertEdge(u, v)	1^*
removeEdge(u, v)	m
getEdges(v)	m
areAdjacent(u, v)	m

Graph Implementation: Adjacency Matrix

good at edge lookup

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



n and m
 if sparse graph
 $\hookrightarrow m \ll n^2$
 \hookrightarrow a lot of 0s
 if dense graph
 $\hookrightarrow m \approx n^2$

u	0
v	1
w	2
z	3

	u	v	w	z
u	0	1	1	0
v	1	0	1	0
w	1	1	0	1
z	0	0	1	0

$O(n)$
 $O(n)$

edge changes

edge lookup

Expressed as O(f)	Adjacency Matrix
Space	n^2
insertVertex(v)	n^*
removeVertex(v)	n^*
insertEdge(u, v)	1
removeEdge(u, v)	1
getEdges(v)	n
areAdjacent(u, v)	1

Graph Implementations

$$O(m+n) \leftarrow O(m)^*$$

I want a graph that uses the least amount of memory possible



↳ Edge List is how most graphs are stored!

I want a graph that has the fastest lookup for specific edges

↳ Adjacency matrix $O(1)$

I want a graph that is efficient for a sparse dataset

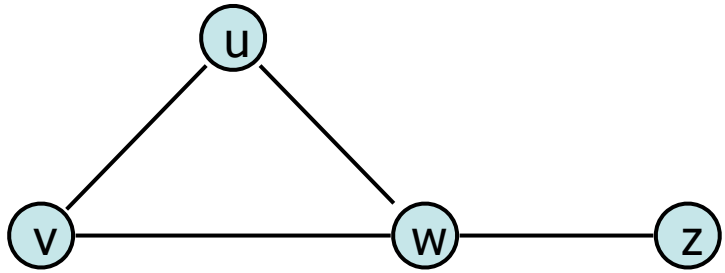
Graph Implementation: Edge List + ?

specific problem

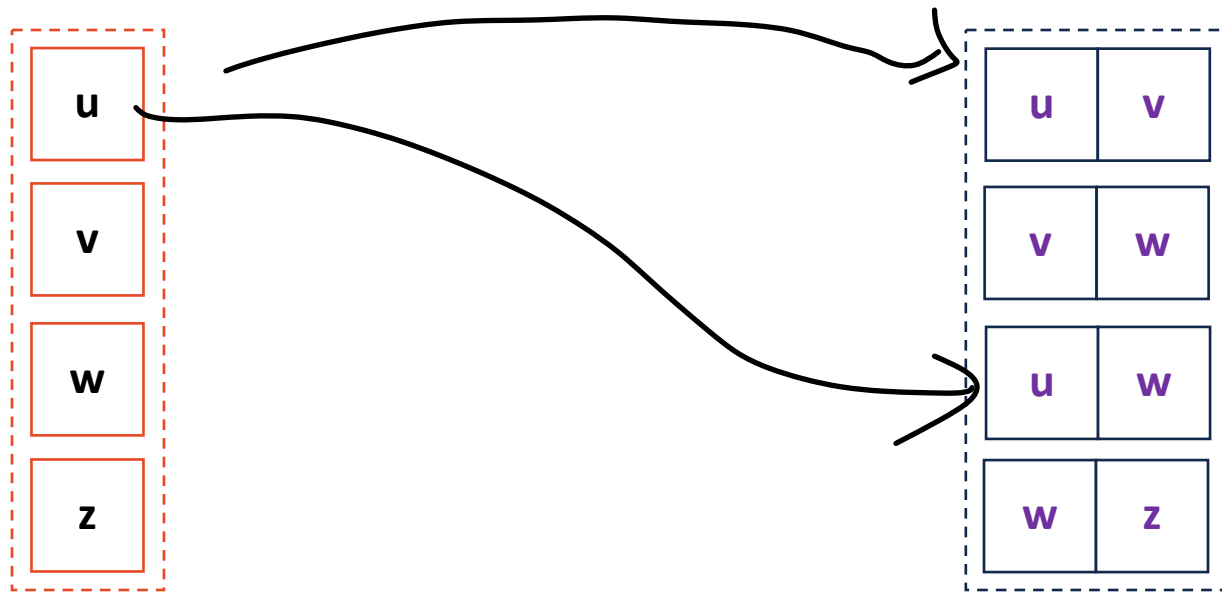
$$|V| = n, |E| = m$$

we want faster

look up for all

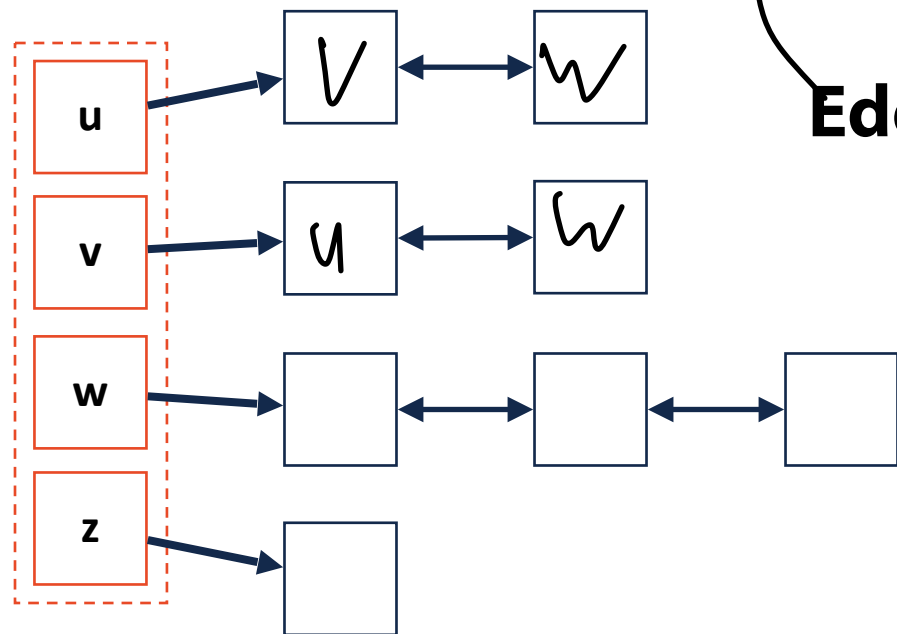
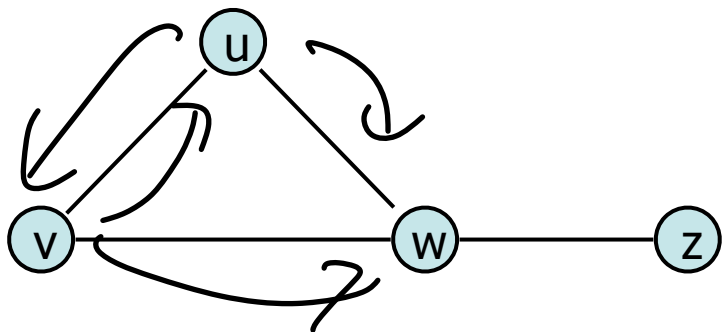


edges connecting to
vertex u



Adjacency List

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



Vertex Storage: Dictionary

↳ Keys are vertex labels

↳ Values are lists

Bidirectional
linked list

Edge Storage:

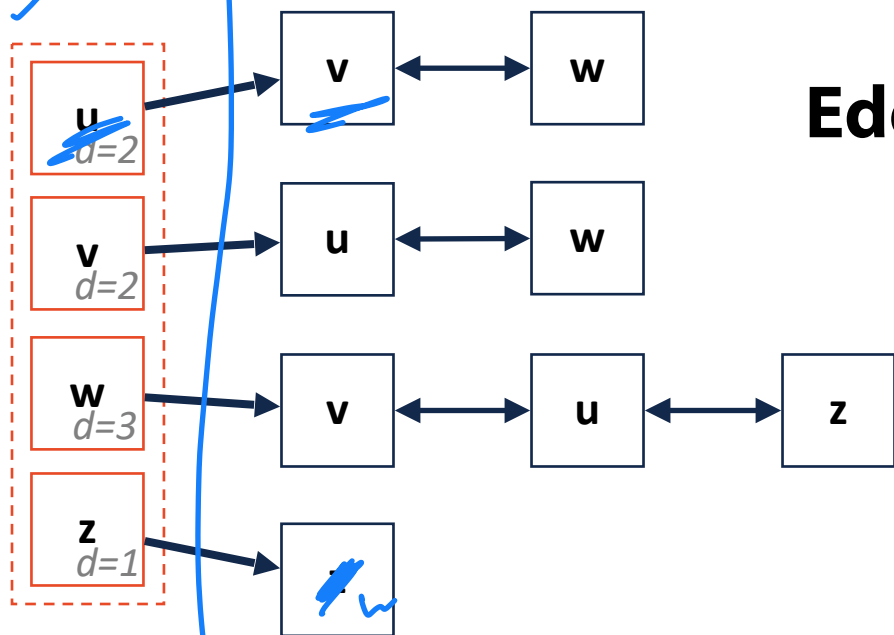
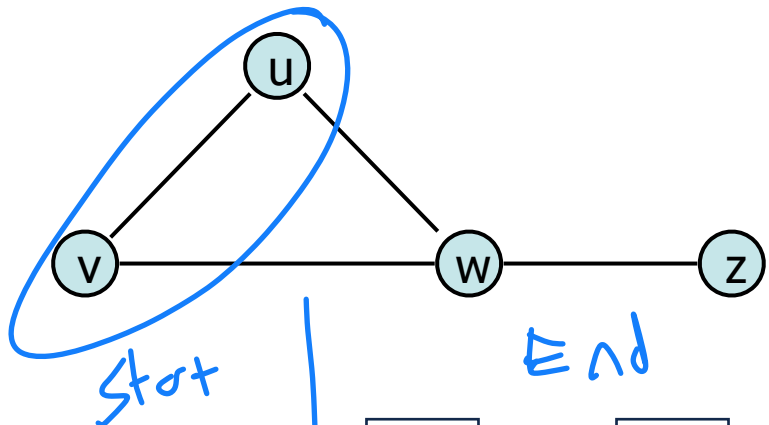
↳ are stored in the list

Adjacency List

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

Dictionary [key = start vertex] = list of end vertices

Vertex Storage: Dictionary

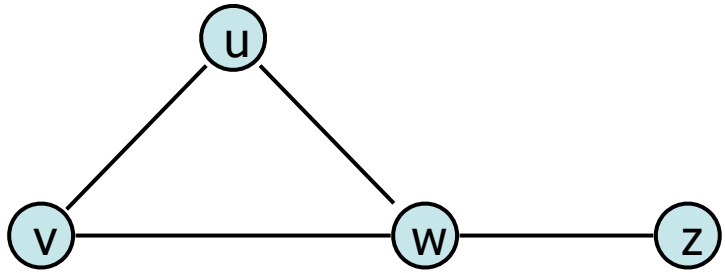


Edge Storage: are in the dictionary

↳ The lists are stored as ~~out going edges~~ end vertex

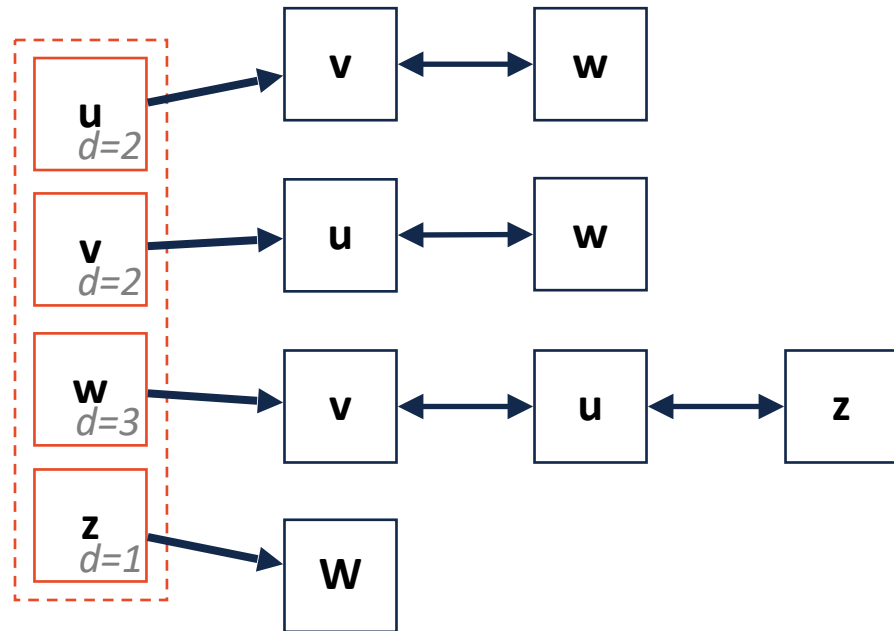
Adjacency List

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



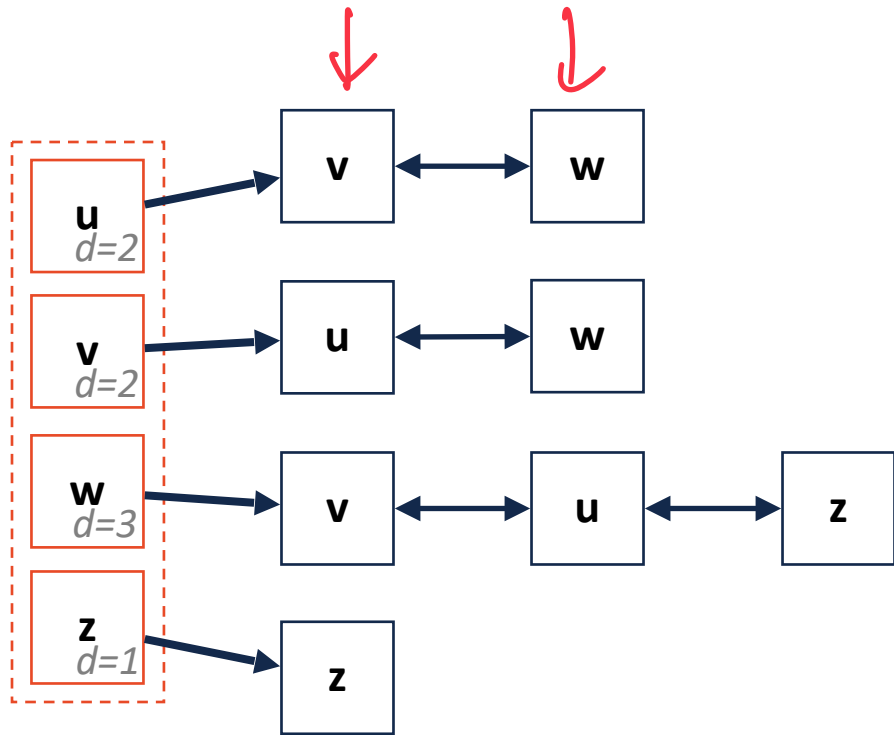
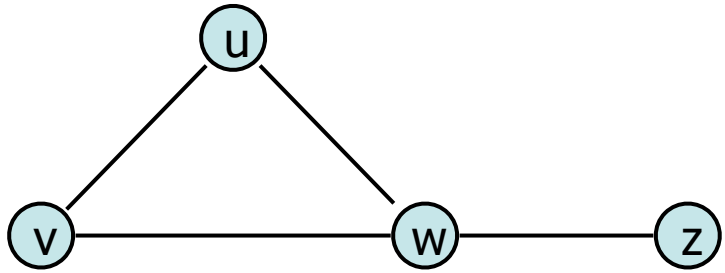
getVertices():

↳ Dictionary.keys()



Adjacency List

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



getEdges(v):

↳ return self.edges[v]

Dictionary

$O(\text{deg}(v))$

$[(u, v), \dots]$

↑
make (u, v) pair

↑
make (v, w) pair

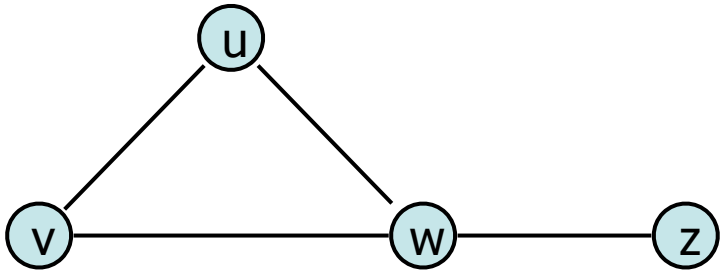
$O(1)$ *

↳ A returning list is fine

Adjacency List



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



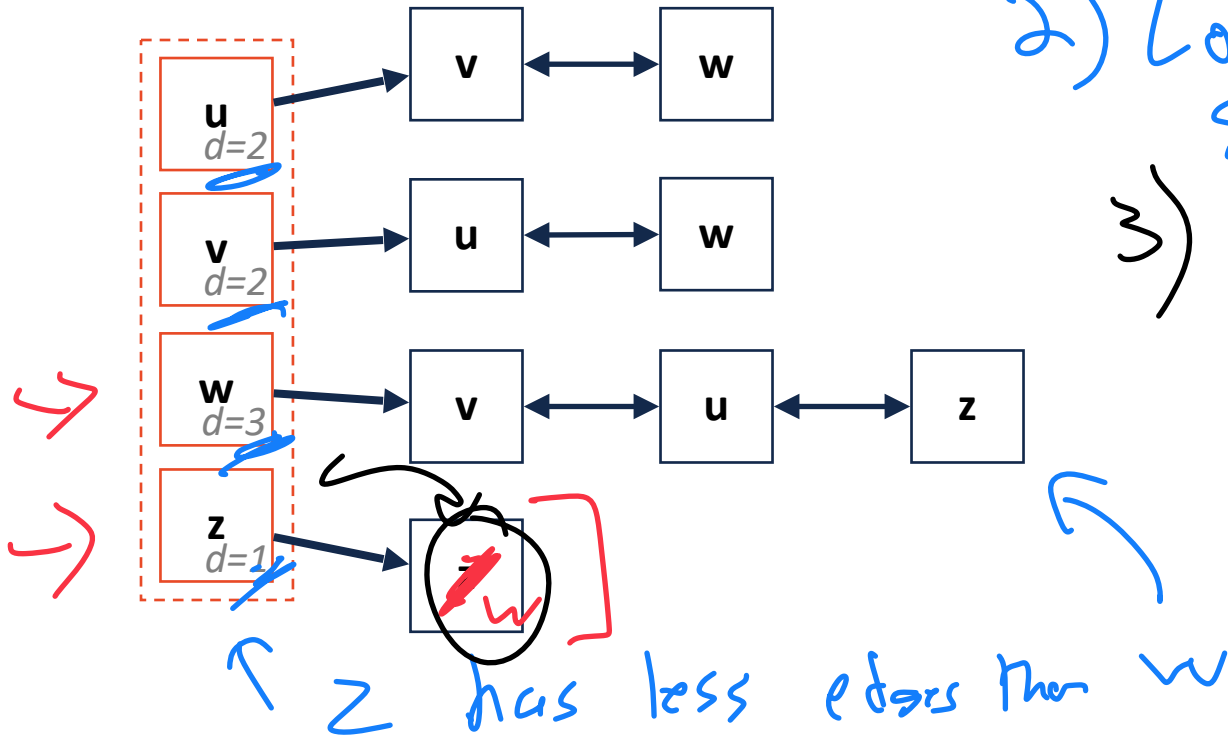
w, z

areAdjacent(u, v):

1) Do vertices exist?
↳ Dictionary.keys()

2) Look up dictionary for key w/
Smaller # of edges (call this v_i)

3) Find v_2 in dictionary $[v_i]$



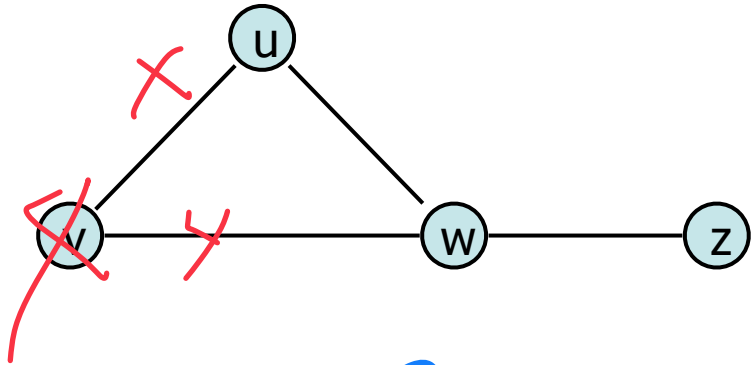
$O(n) \rightarrow O[\text{deg}(v)]$

↑ Dense graph ↑ sparse

Adjacency List

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

(v)

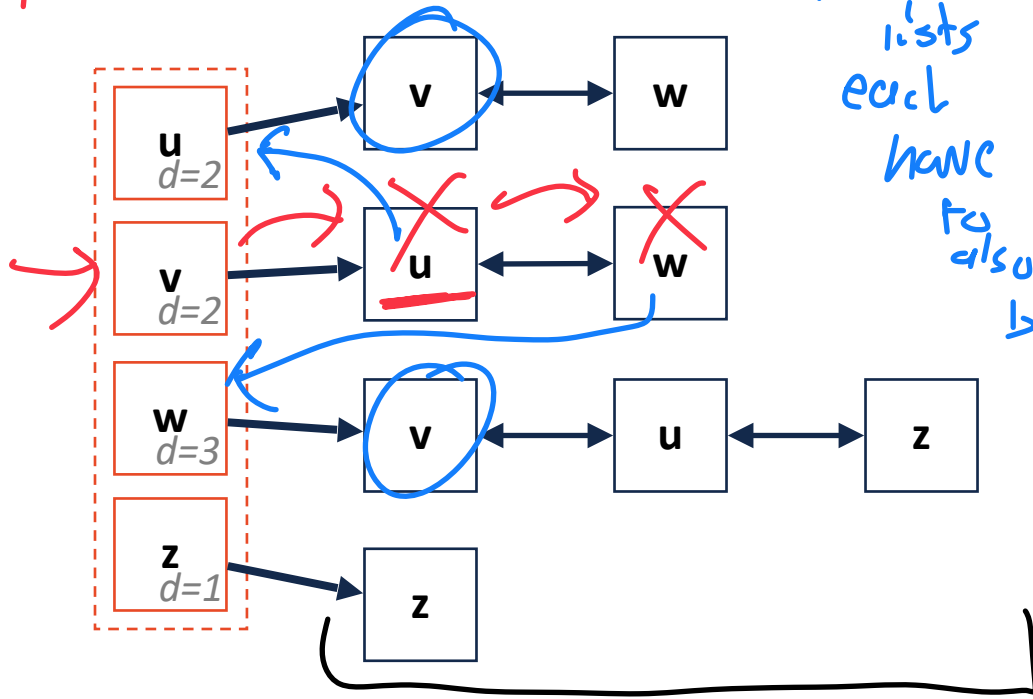


removeVertex(v):

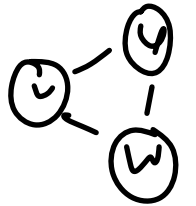
Delete vertex And edges
 1) From dictionary (v)

At most lists each have to also be searched

2) Find matching label in all edges

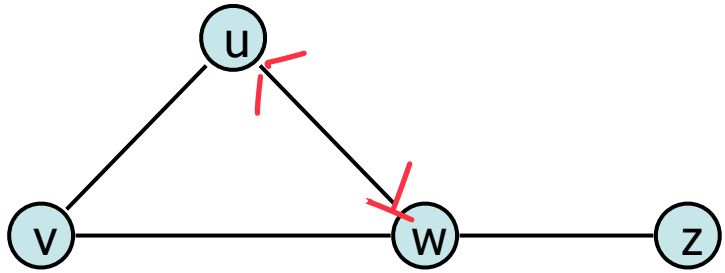


N lists $\rightarrow O(n^2)$
 List of size n



Simple Adjacency List

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

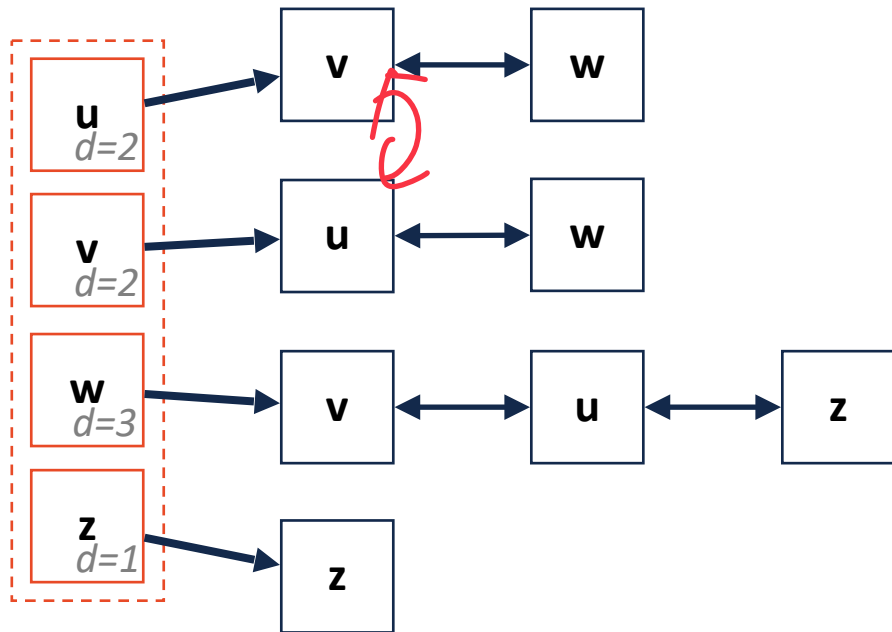


What's wrong with our implementation?

↳ We store every edge twice
↳ But they aren't linked together!

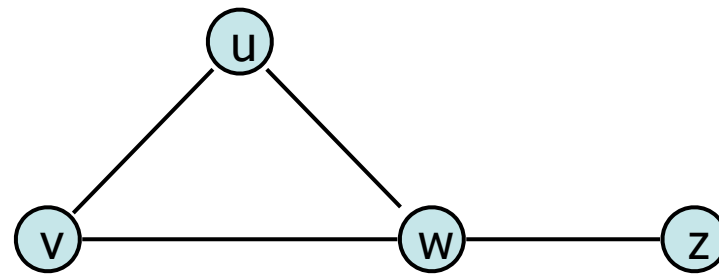
How can we fix it?

↳ Draw arrows connecting
 (u,v) to (v,u)



Adjacency List

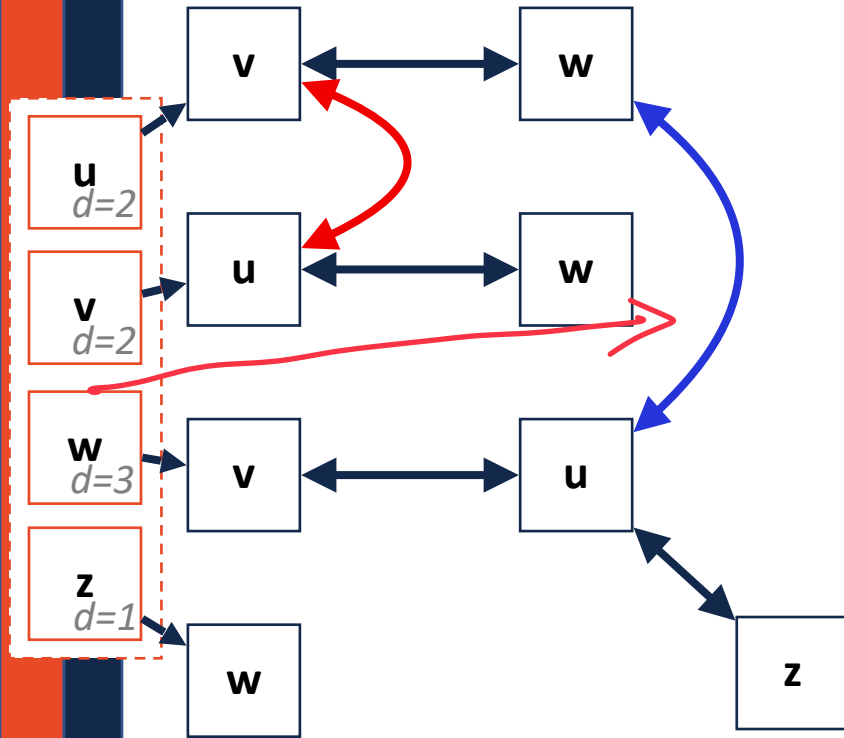
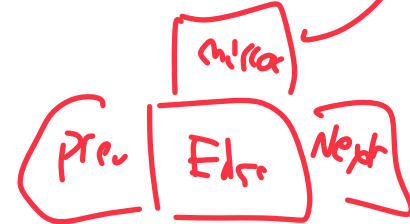
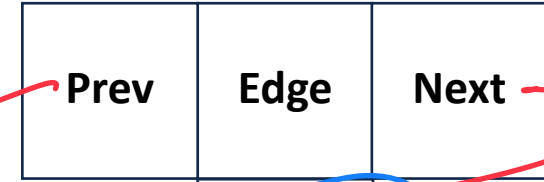
$$|V| = n, |E| = m$$



B; directional list

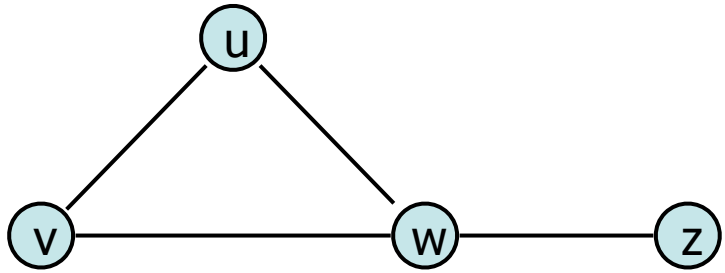


Adj List Node:



This is still a 'lie' but a more accurate one!

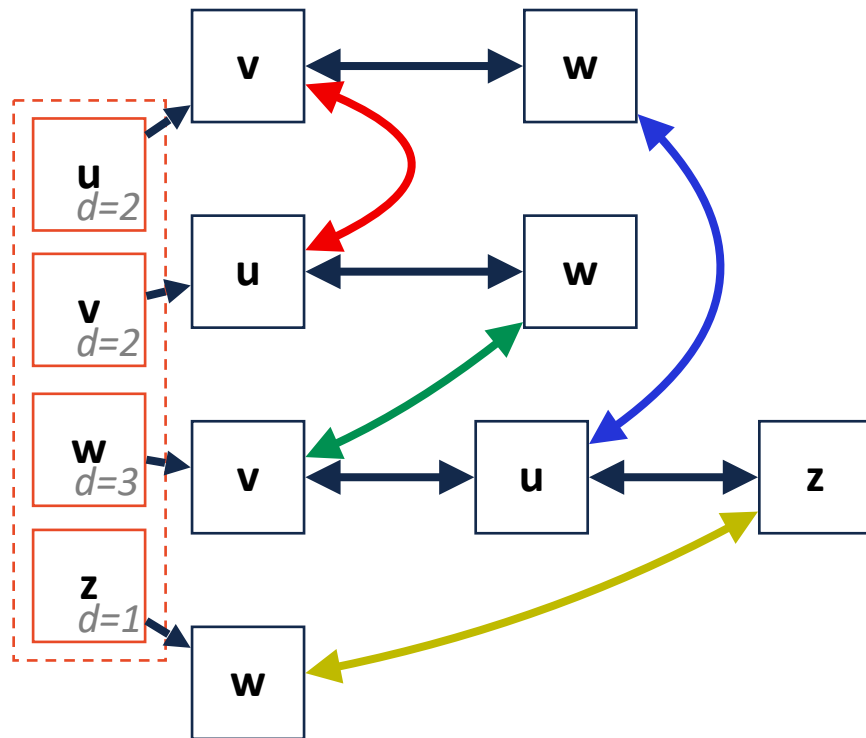
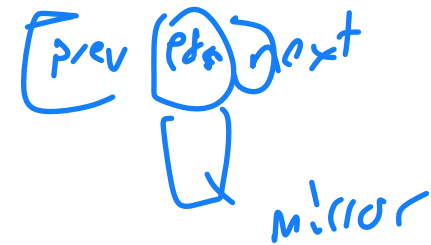
Adjacency List



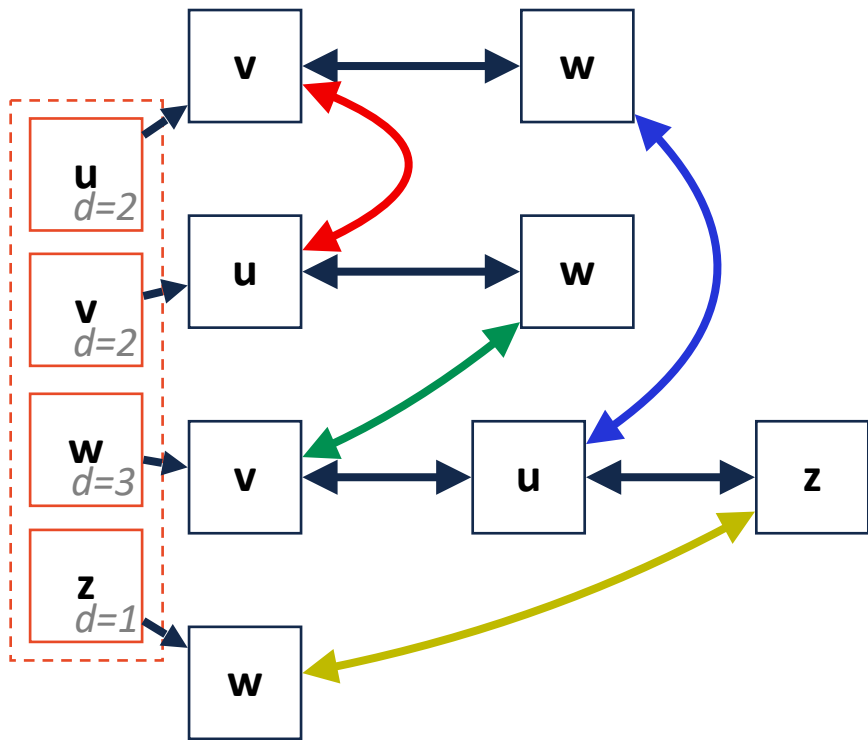
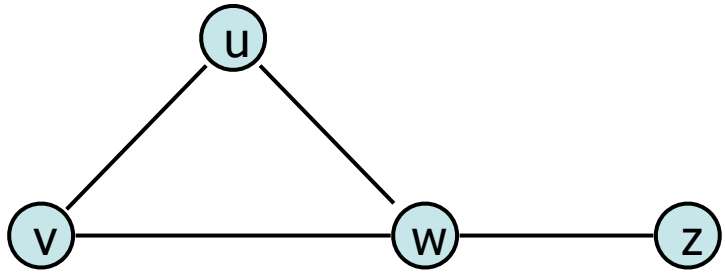
insertVertex(v):

↳ self.edges [v] = []
 = Vertex List()

= Fancy List Node()



Adjacency List



$O(1)^*$

insertEdge(u, v):

lookup $\rightarrow O(1)^*$

1) If u or v is not in dictionary

\hookrightarrow Dictionary.Keys() \rightarrow list of keys

$O(1)^*$

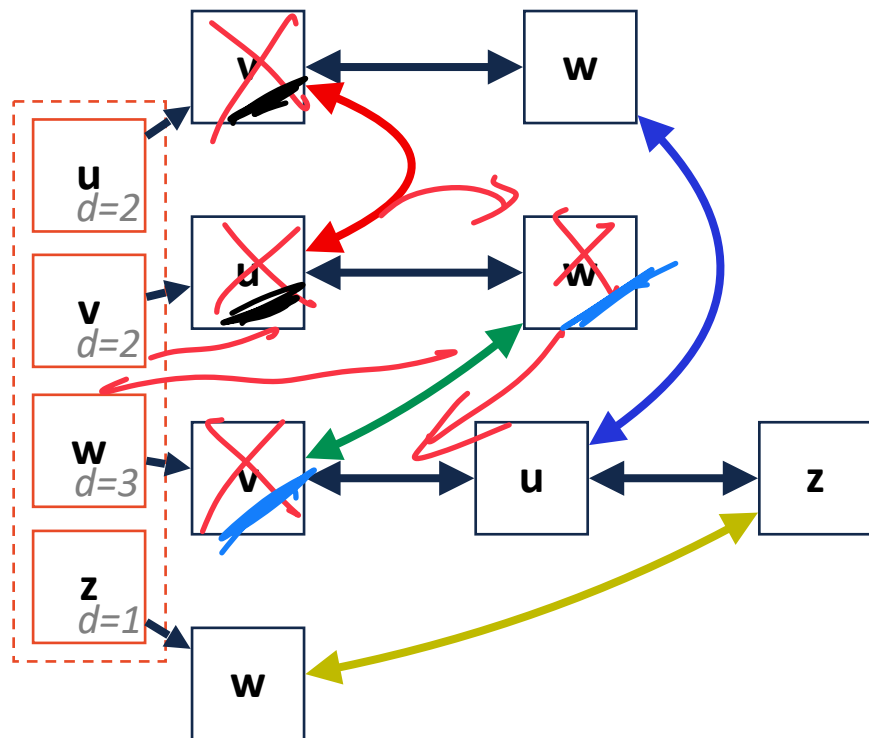
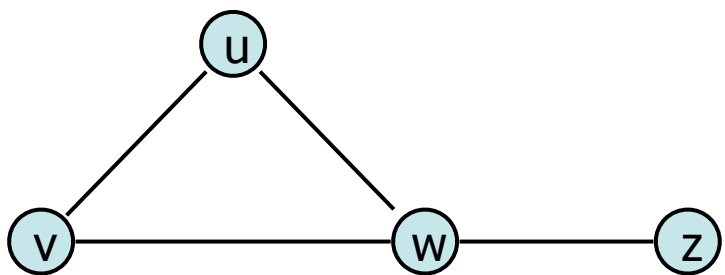
\hookrightarrow add to dictionary (insert vertex)

2) Add edge to dictionary twice

$O(1)^*$
 $O(1)^*$

Dictionary[u].append(v)
Dictionary[v].append(u)

Adjacency List



Vertex $O(1)^*$

~~removeEdge(u, v):~~ →

1) Look up u or v (smaller degree)

2) Walk through list deleting node & node's mirror!

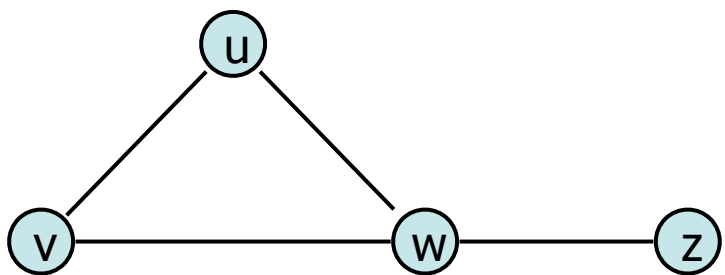
↳ $O(n)$

↳ every node deleted is technically 2 nodes

Overall: $O(n)$

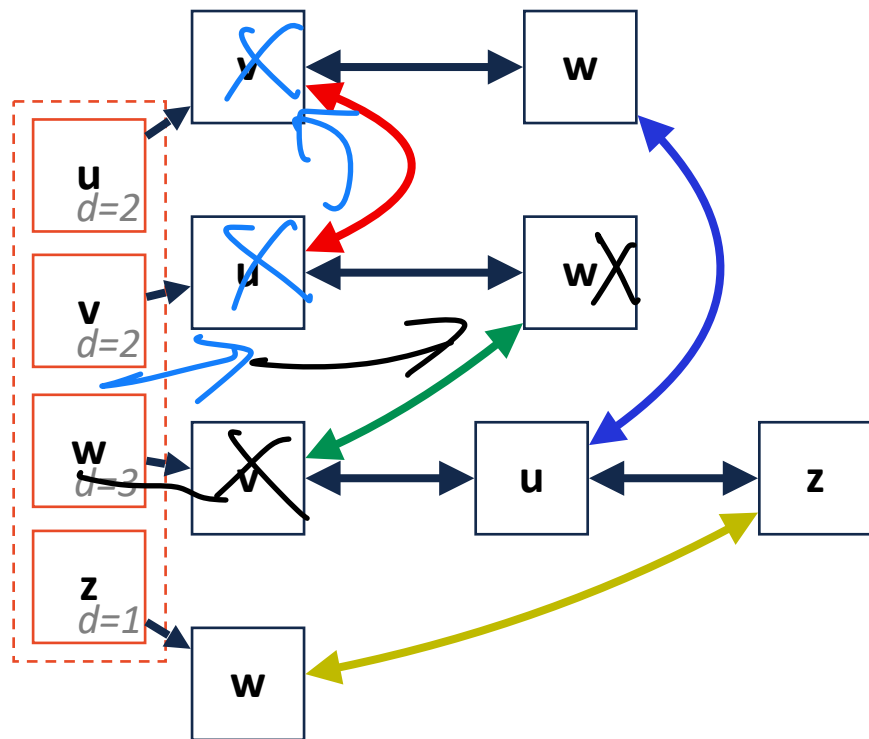
Adjacency List

$O(n)$



removeEdge(u, v):

- 1) Look up ~~u~~ ~~v~~ ✓
- 2) Remove (u, v) and (v, u)

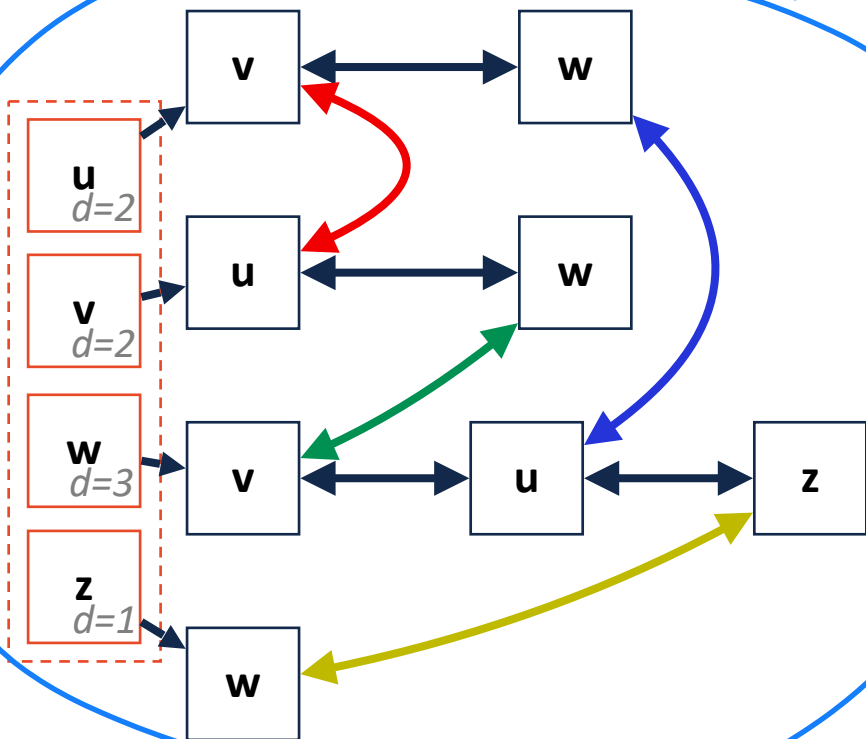
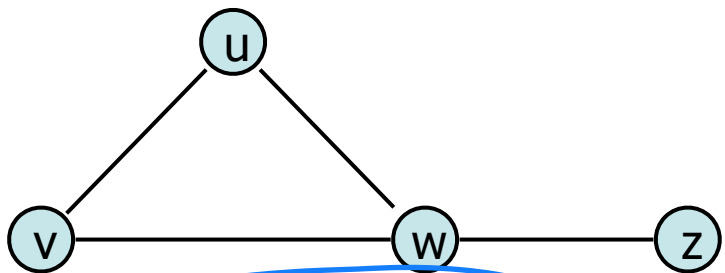


Two implementations

One Naive $v \rightarrow w$ $w \rightarrow v$ $O(n) + O(n)$

One "Smarter" $v \rightarrow w$ $w \rightarrow v$ $O(n)$

Adjacency List



Pros: Good at insert ^{vertex} or edge

- If sparse good at everything!


- Edge lookup - Not as good as adj matrix but... not bad

Cons:

↳ Removal of edges/vertices has to be done twice

↳ Coding it is a pain



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

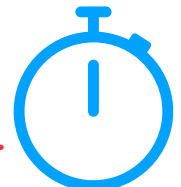
Expressed as O(f)	Edge List	Adjacency Matrix	Adjacency List
Space	$n+m$	n^2	$n+m$
insertVertex(v)	1^*	n^*	1^*
removeVertex(v)	n^2 n	n^*	$deg(v)$
insertEdge(u, v)	1	1	1^*
removeEdge(u, v)	m	1	$\min[deg(u), deg(v)]$
getEdges(v)	m	n	$deg(v)$
areAdjacent(u, v)	m	1	$\min[deg(u), deg(v)]$

Note: $O(deg(v))$
is
 $O(n)$
at
worst

Circle the best

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

 vertices \leftarrow n

 \leftarrow edges \leftarrow m


Expressed as O(f)	Edge List	Adjacency Matrix	Adjacency List
Space	$n+m$	n^2	$n+m$
insertVertex(v)	1^*	n^*	1^*
removeVertex(v)	m^* $n+m$	n^*	$\text{deg}(v)$
insertEdge(u, v)	1	1	1^*
removeEdge(u, v)	m	1	$\min(\text{deg}(u), \text{deg}(v))$
getEdges(v)	m	n	$\text{deg}(v)$
areAdjacent(u, v)	m	1	$\min(\text{deg}(u), \text{deg}(v))$

\leftarrow base speed

Where do we go from here?

Consider: How does our implementation change for weights?
for directed edges?

Consider: How can we implement traversal on graphs?

Graph ADT ↗

Consider: What are some common graph algorithms / uses?

Graphs in Python: NetworkX Package

NetworkX uses concepts from all three implementation methods

A graph can be built from an edge list, adjacency matrix, or adjacency list

A graph can be saved or output as any of the three formats

Many algorithms (and traversals) are built-in.

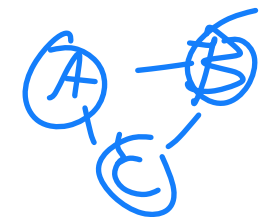
Creating a NetworkX graph

import networkx as nx

```
G = nx.Graph()
```

```
G = nx.Graph(edgeList)
```

← $[(A, B), (A, C), (B, C)]$



```
G.add_node(label, **kwargs)
```

add vertices

```
G.add_edge(v1, v2, **kwargs)
```

add edges

C B A
A B A
B C C
D C

NetworkX Example

```
1 import networkx as nx
2
3 G = nx.Graph()
4
5 G.add_edge("A", "B")
6
7 G.add_edge("B", "C", weight=5)
8
9 G.add_edge("A", "C", anything="Bob", I="was", want="here")
10
11
12 print(G.nodes())
13
14
15 print(G.edges())
16
17
18 print(G.edges(data=True))
19
20
21
22
```

NetworkX Example

```
1 import networkx as nx
2
3 G = nx.Graph()
4
5 G.add_node("A")
6
7 G.add_node("B", name="Bob")
8
9 G.add_node("C", anything="Bob", I="was", want="here")
10
11 print(G.nodes())
12
13 print(G.nodes(data=True))
14
15
16
17
18
19
20
21
22
```

NetworkX Example

```
1 G=nx.random_regular_graph(3, 6)
2
3 nx.draw(G, edge_color='k', width=2, with_labels=True)
4
5 plt.show()
6
7 m = nx.adjacency_matrix(G)
8 print(m.todense())
9
10
11
12
13
14
15 adjL = nx.generate_adjlist(G)
16 for line in adjL:
17     print(line)
18
19
20
21
22
```

Graphs in Python: NetworkX Package

Networkx (and Python packages in general) can do a lot for you!

But they can sometimes make design decisions that don't work for you.

Ex: An adjacency list in NetworkX doesn't duplicate edges!