

The Graph ADT and implementations

The screenshot displays a Google Maps interface with a route calculated between two locations in Ireland. The starting point is 'Western Gateway Building - UCC, University College Cork' and the destination is 'Athlone Bus Station, Southern Station Road, Athlone'. The map shows a blue route line passing through several towns including Limerick, Thurles, and Portlaoise. A sidebar on the left provides route details, including travel time, distance, and toll information. The bottom of the screen shows the Google Maps logo, copyright information, and a scale bar.

Western Gateway Building - UCC, University College Cork

Athlone Bus Station, Southern Station Road, Athlone

Route options

- via M8
2 h 30 min without traffic · [Show traffic](#)
250 km
This route has tolls.
- via M8 and N62
2 h 47 min
- via 205, 1, 2, 6, 9
13 h 9 min

Athlone Bus Station

2 h 47 min
217 km

2 h 46 min
250 km

4 h 7 min
every 60 min

Western Gateway Building - UCC

Google

Map data ©2015 Google

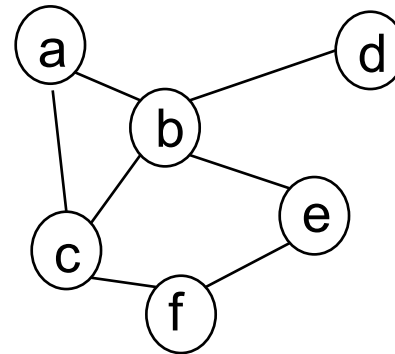
Terms Privacy Report a problem 20 km

Graphs

A graph is an abstract representation of the relationships between a set of objects.

We saw graphs in CS1113:

- social network graphs, call graphs, route map graphs, ...
- graph properties
- shortest path algorithms
- spanning trees



But it was all on paper. We didn't see how to

- implement them efficiently
- implement efficient algorithms for processing them

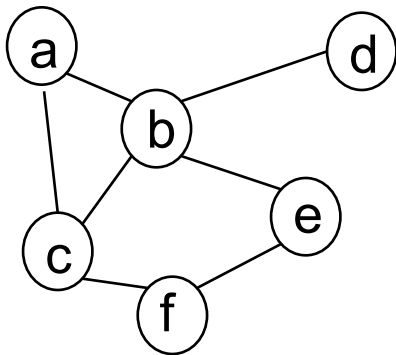
A *simple graph* is a pair (V,E) , where V is a set of *vertices*, and E is a set of *edges*, and each edge is a set $\{x,y\}$, where x and y are vertices in V .

The *degree* of a vertex x is the number of edges that contain x .

Two vertices x and y are *adjacent* if there is an edge $\{x,y\}$.

Edge $\{x,y\}$ is *incident* on x (and incident on y).

The *neighbours* of a vertex x are all other vertices adjacent to x



$$V = \{a,b,c,d,e,f\}$$

$$E = \{ \{a,b\}, \{a,c\}, \{b,c\}, \{b,d\}, \{b,e\}, \{c,f\}, \{e,f\} \}$$

$$\text{degree}(b) = 4$$

$$\text{neighbours}(b) = \{a,c,d,e\}$$

In a *multigraph* E is a bag of edges, and so there may be multiple edges $\{x,y\}$ in E for the same pair of vertices x and y .

In a *directed* graph, each edge is an ordered pair (x,y) .

For a directed graph,

- the *out-degree* of a vertex is the number of edges with x as the first element of the pair
- the *in-degree* of a vertex x as the second element.

A *weighted* graph has a function w from E to some set, defining a *weight* with the edge.

We can also associate *labels* from some set L with either vertices or edges.

Designing the ADTs

We will maintain the graph as a complex data structure, which is composed of vertices and edges.

What methods should a Graph Abstract Data Type offer?

- this is not asking how to implement the data structure
- there are many different ways to implement graphs
- before we start, we need to be clear about what questions we will ask the graph, what commands we will issue, and what we want the data structure to give us in return

E.g. What are your vertices? Is this vertex linked to this other one? What is the weight of that vertex? What is the weight of the edge between those two vertices? How many edges link to that vertex?

Add a new edge to connect these two vertices. Remove that vertex. ...

Vertex and Edge ADTs

Vertex

`element()`: returns the label of the vertex

Edge

`vertices()`: returns the pair of vertices the edge is incident on

`opposite(x)`: if the edge is incident on x , return the other vertex

`element()`: return the label of the edge

(Undirected) Graph ADT

<code>vertices():</code>	return a list of all vertices
<code>edges():</code>	return a list of all edges
<code>num_vertices():</code>	return the number of vertices
<code>num_edges():</code>	return the number of edges
<code>get_edge(x,y):</code>	return the edge between x and y (if it exists)
<code>degree(x):</code>	return the degree of vertex x
<code>get_edges(x):</code>	return a list of all edges incident on x
<code>add_vertex(elt):</code>	add a new vertex with element = elt
<code>add_edge(x, y, elt)</code>	a new edge between x and y, with element elt
<code>remove_vertex(x):</code>	remove vertex and all incident edges
<code>remove_edge(e):</code>	remove edge e

Implementing the ADT

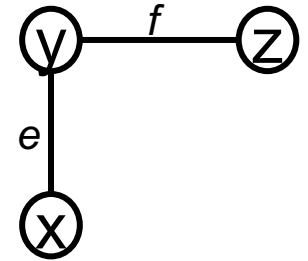
The main operations will be retrieving vertices and edges. Updating will be relatively rare.

The main question is how to store and retrieve the edges. There are 4 main options:

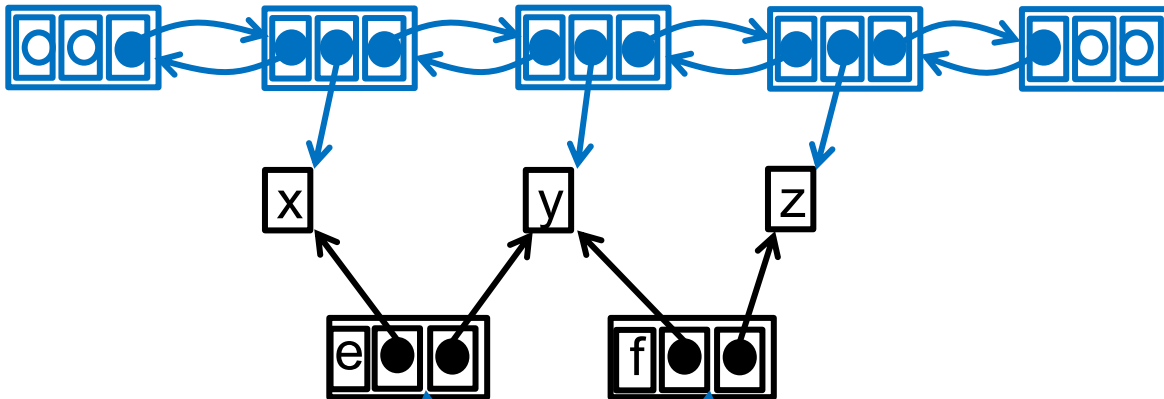
1. a list of edges
2. adjacency list:
 - for each vertex, store a list of the edges incident on it
3. adjacency map:
 - for each vertex, store a map of the edges incident on it, using the other vertex as the key
4. adjacency matrix:
 - maintain a 2D array, where $\text{matrix}[i][j]$ contains a reference to the edge $\{i,j\}$ (ie the edge between the i th and j th vertices)

Edge List

Maintain the vertices and edges in unordered linked lists.



Vertex list



Edge list

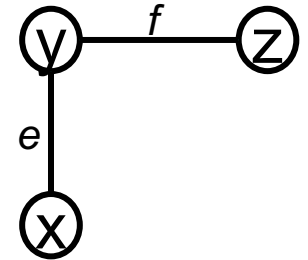


Edge List (improved)

Maintain the vertices and edges in unordered linked lists.

each vertex maintains a reference back to the list elt

each edge maintains a reference back to the list elt



Vertex list



Edge list



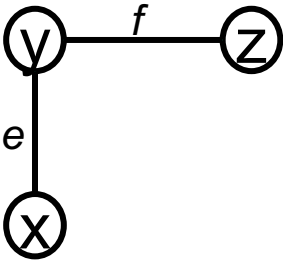
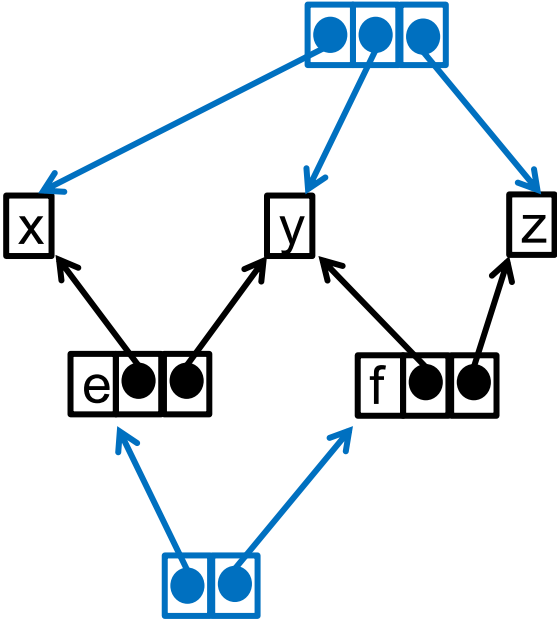
Edge List: complexity

If n is the number of vertices, and m is the number of edges,

Space complexity:	$O(n + m)$
<code>get_edge(x,y):</code>	$O(m)$ – must check each edge
<code>degree(x):</code>	$O(m)$ – must check each edge
<code>get_edges(x):</code>	$O(m)$ – must check each edge
<code>add_vertex(elt):</code>	$O(1)$
<code>add_edge(x, y, elt):</code>	$O(1)$
<code>remove_edge(e):</code>	$O(1)$
<code>remove_vertex(x):</code>	$O(m)$ - must check each edge

Would it make a difference if we sorted the lists?

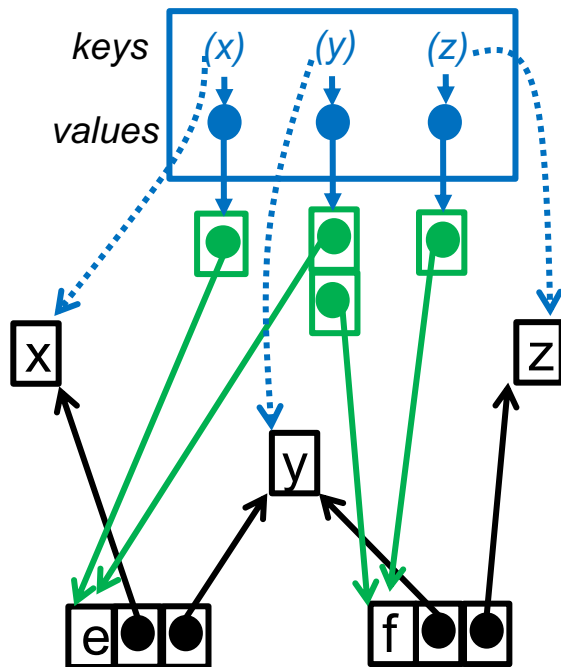
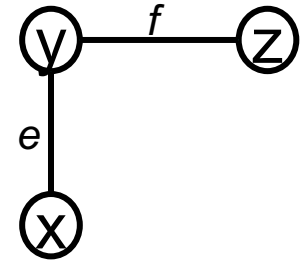
Now going to sketch the implementation
like this, to make the sketch more compact



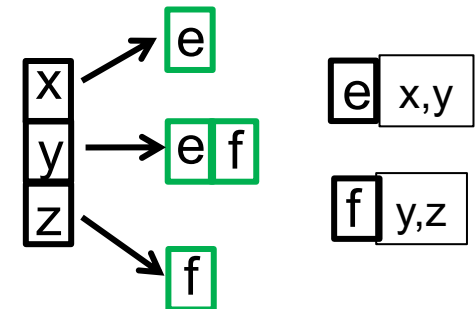
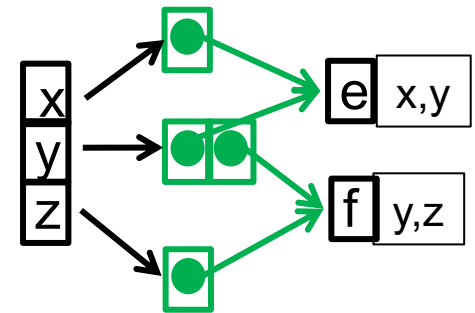
Adjacency List

Maintain a data structure of vertices.

- simplest to use a dictionary
- key is reference to vertex object
- value is reference to list of edges incident on that vertex



This is the simpler sketch to understand, as long as we remember what references we are hiding ...



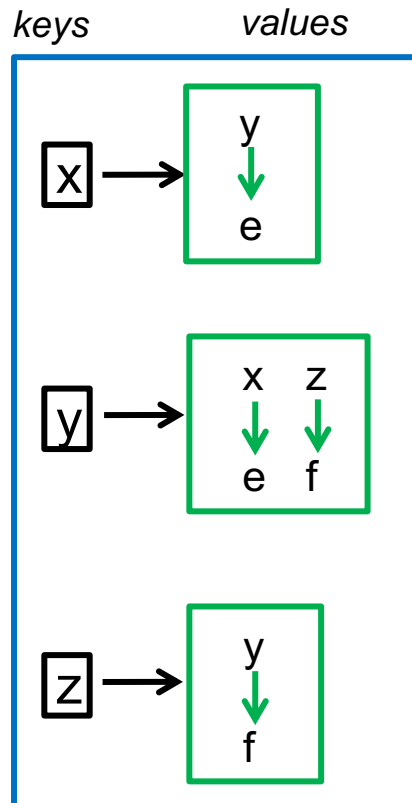
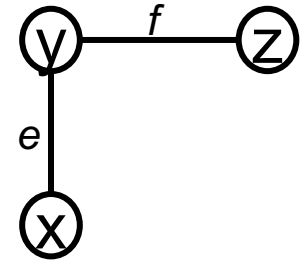
Adjacency List: complexity

If n is the number of vertices, and m is the number of edges,

Space complexity:	$O(n + m)$	
<code>get_edge(x,y):</code>	$O(\min(\text{degree}(x), \text{degree}(y)))$	
<code>degree(x):</code>	$O(1)$	
<code>get_edges(x):</code>	$O(\text{degree}(x))$	(assuming we copy the list, rather than return the actual adjacency list ...)
<code>add_vertex(elt):</code>	$O(1)$	
<code>add_edge(x, y, elt):</code>	$O(1)$	
<code>remove_edge(e):</code>	$O(\text{degree}(x) + \text{degree}(y))$	
<code>remove_vertex(x):</code>	$O(n)$	($O(\text{degree}(x))$ to identify the edges on x , but then removing those edges depends on the degree of the other vertex in each case, since we have to remove the edge from that other vertex's adjacency list ...)

Adjacency map

Maintain a data structure of vertices (e.g. dictionary)
Each vertex (key) maintains (value) a hash-map where
the other vertices are the keys, and the incident edges
are the values.



e x,y

f y,z

Adjacency Map: complexity

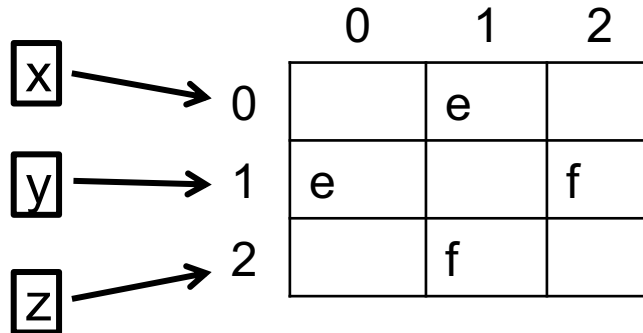
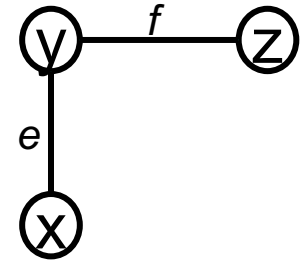
If n is the number of vertices, and m is the number of edges,

Space complexity:	$O(n + m)$	
get_edge(x,y):	$O(1)$ <i>expected</i>	worst case $O(\min(\text{degree}(x), \text{degree}(y)))$
degree(x):	$O(1)$	
get_edges(x):	$O(\text{degree}(x))$	
add_vertex(elt):	$O(1)$	
add_edge(x, y, elt):	$O(1)$ <i>expected</i>	
remove_edge(e):	$O(1)$ <i>expected</i>	
remove_vertex(x):	$O(\text{degree}(x))$	

Adjacency matrix

Associate a unique integer in 0 to $n-1$ with each vertex

Maintain a 2D array, where $\text{cell}[i][j]$ contains a reference to the edge between i and j



e x,y

f y,z

Adjacency matrix: complexity

If n is the number of vertices, and m is the number of edges,

Space complexity:	$O(n^2)$	Wasteful for sparse graphs
get_edge(x,y):	$O(1)$	
degree(x):	$O(n)$	
get_edges(x):	$O(n)$	
add_vertex(elt):	$O(n^2)$	
add_edge(x, y, elt):	$O(1)$	
remove_edge(e):	$O(1)$	
remove_vertex(x):	$O(n^2)$	

Summary

	edge list	adjacency list	adjacency map	adjacency matrix
Space	$O(n + m)$	$O(n + m)$	$O(n + m)$	$O(n^2)$
get_edge	$O(m)$	$O(\min(\deg(x), \deg(y)))$	$O(1)$ <i>expected</i>	$O(1)$
degree	$O(m)$	$O(1)$	$O(1)$	$O(n)$
get_edges(x)	$O(m)$	$O(\deg(x))$	$O(\deg(x))$	$O(n)$
add_vertex	$O(1)$	$O(1)$	$O(1)$	$O(n^2)$
add_edge	$O(1)$	$O(1)$	$O(1)$ <i>expected</i>	$O(1)$
remove_edge	$O(1)$	$O(1)$	$O(1)$ <i>expected</i>	$O(1)$
remove_vertex	$O(m)$	$O(\deg(x))$	$O(\deg(x))$	$O(n^2)$

Adding to the underlying structures may change some of these complexities

Next lecture

Graph traversals