Graphs

#### **Graphs**













## What are Graphs Good for?

- Graphs are a convenient **abstraction** that brings out commonalities between different domains
- Once we understand a problem in term of graphs, we can use **general graph algorithms** to solve it

o no need to reinvent the wheel every time

• Graphs are everywhere





- This is what a social network looked like … in 2005
- vertices are people posting photos
- edges are people following the photo stream of others

#### The FlickrVerse, April 2005



n. 成型

# Lightsout

- Lightsout is a *game* played on boards consisting of *n x n* lights o each light can be either on or off
- We make a *move* by pressing a light, which toggles it and its cardinal neighbors
- **•** From a given configuration, the *goal* of the game is to turn off all light











## Getting Introduced

**E** 

- Figuring out how to get introduced to someone also amounts to **finding a path** between them
	- o Graphs bring out commonalities between different domains

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#### ● On *n x n* lightsout,

o there are *2 n\*n* board configurations  $\triangleright$  each of the  $n^*n$  lights can be either on or off o from any board, we can make *n\*n* moves by pressing any one of the *n\*n* lights

#### ● The graph representing *n x n* lightsout has

- o *2 n\*n* vertices
- o *n\*n \* 2n\*n / 2* edges
	- $\triangleright$  there are  $2^{n^{*}n}$  vertices
	- each has *n x n* neighbors
	- $\triangleright$  but this would count each edge (A,B) twice
		- $\Box$  from A to B and
		- $\Box$  from B to A
		- so we divide by 2



## Models vs. Data Structures

#### • A graph can be

o a conceptual **model** to understand a problem

o a concrete **data structure** to solve it

#### ● For 2x2 lightsout, it is both

- o Conceptually, it brings out the structure of the problem and highlights what it has in common with other problems
- o Concretely, we can traverse a data structure that represents it in search of a path to the solved board
- Turning 6x6 lightsout into a data structure is not practical o each board requires 36 bits
	- o we need over 64GB to represent its 2 <sup>36</sup> vertices
	- o we need over 2TB to represent its 36 \* 2 <sup>36</sup> / 2 edges

That's more memory than most computers have

## Implicit Graphs

 We don't need a graph data structure to solve *n x n* lightsout o from each board we can **algorithmically** generate all boards that can be reached in one move

- o pick one of them and repeat until
	- $\triangleright$  we reach the solved board
	- $\triangleright$  or we reach a previously seen board
		- $\Box$  from it try a different move
- **•** In the process, we are building an **implicit graph** 
	- $\circ$  a small portion of the graph exists in memory at any time
		- $\triangleright$  the boards we have previously seen
			- □ vertices
		- $\triangleright$  the moves we still need to try from them
			- $\Box$  edges

## Explicit Graphs

- For many graphs, there is no algorithmic way to generate their edges
	- $\triangleright$  roads between cities
	- $\triangleright$  social networks
	- $\triangleright$  ...
- We must represent them explicitly as a data structure in memory

• We will now develop a small library for solving problems with these **explicit graphs**

#### **A Graph Interface**

## A Minimal Graph Data Structure

- What we need to represent
	- o graphs themselves
		- type graph\_t
	- o the vertices of a graph
		- $\triangleright$  type vertex
			- $\Box$  we label vertices with the numbers 0, 1, 2, ...
				- consecutive integers starting at 0
			- $\Box$  vertex is defined as unsigned int
	- o the edges of the graph
		- $\triangleright$  we represent an edge as its endpoints
			- *no need for an edge type*

## A Minimal Graph Data Structure

#### • Basic operations on graphs

o graph\_new(n) create a new graph with n vertices

- $\triangleright$  we fix the number of vertices at creation time
	- $\Box$  we cannot add vertices after the fact
- o graph\_size(G) returns the number of vertices in G
- o graph\_hasedge(G, v, w) checks if the graph G contains the edge  $(v,w)$
- o graph\_addedge(G, v, w) adds the edge (v,w) to the graph G o graph\_free(G) disposes of G
- A realistic graph library would provide a much richer set of operations

o we can define most of them on the basis of these five

## A Minimal Graph Interface – I



#### Example

#### • We create this graph as

 $graph_t G = graph_new(5);$ graph\_addedge(G, 0, 1); graph\_addedge(G, 0, 4); graph\_addedge(G, 1, 2); graph\_addedge(G, 1, 4); graph\_addedge(G, 2, 3); graph\_addedge(G, 2, 4);

in any order



#### Then

- graph\_hasedge(G, 3, 2) returns true, but
- $\triangleright$  graph\_hasedge(G, 3, 1) return false
	- $\Box$  there is a path from 3 to 1, but no direct edge

 $\bullet$  It is convenient to handle neighbors explicitly

 $\triangleright$  this is not strictly necessary

 $\triangleright$  but graph algorithms get better complexity if we do so inside the library

#### Abstract type of neighbors o neighbors\_t

• Operations on neighbors

o graph\_get\_neighbors(G, v)

 $\triangleright$  returns the neighbors of vertex v in G

o graph\_hasmore\_neighbors(nbors)

 $\triangleright$  checks if there are additional neighbors

- o graph\_next\_neighbor(nbors)
	- $\triangleright$  returns the next neighbor

This is called an **iterator**

These allow us to iterate through

the neighbors of a vertex

- o graph\_free\_neighbors(nbors)
	- $\triangleright$  dispose of unexamined neighbors

## A Minimal Graph Interface – II



### Example

G

#### • We grab the neighbors of vertex 4 as

neighbors\_t  $n4 = graph$ \_get\_neighbors(G, 4);

n4 contains vertices 0, 1, 2 in some order

vertex  $a = graph\_next\_neighbor(n4);$ 

 $\triangleright$  say a is vertex 1

 $\Box$  it could also be 0 or 2

vertex  $b = graph\_next\_neighbor(n4);$ 

 $\triangleright$  say b is vertex 0

 $\Box$  it cannot be 1 because we already got that neighbor

 $\Box$  but it could be 2

vertex  $c = graph\_next\_neighbor(n4);$ 

 $\triangleright$  c has to be vertex 2

 $\Box$  it cannot be 0 or 1 because we already got those neighbors

graph\_hasmore\_neighbor(n4)

 $\triangleright$  returns false because we have exhausted all the neighbors of 4



#### **Implementing Graphs**

## Implementing Graphs

• How to implement graphs based on what we studied?

- o The main operations are
	- $\triangleright$  adding an edge to the graph
	- $\triangleright$  checking if an edge is contained in the graph
		- □ These are the operations we had for **sets**
	- $\triangleright$  iterating through the neighbors of a vertex
- Implement graphs as
	- o a linked list of edges
	- o a hash set

We could also use AVL trees if we are able to sort the edges

• How much would the operations cost?

## Measuring the Cost of Graph Operations

This is a complete graph

- If a graph has **v** vertices, the number **e** of edges ranges between
	- $\circ$  0, and The graph has no edges
	- $O (v-1)/2$ 
		- $\triangleright$  there is an edge between each of the v vertices and the other v-1 vertices, but we divide by 2 so that we don't double-count edges

#### $\bullet$  So,  $e \in O(v^2)$

o we could do with just v as a cost parameter,

- $\circ$  but many graphs have far fewer than  $v(v-1)/2$  edges
	- using only v would be overly pessimistic
- Use **both** v and e as cost parameters

## Naïve Graph Implementations

 For implementations based on known data structures, the cost of the basic graph operations are



What about iterating through the neighbors of a vertex?

## Naïve Graph Implementations

• Finding the neighbors of a vertex requires going over all the edges

o graph\_get\_neighbors has cost O(e) and O(v) avg

- How many neighbors are there?
	- o at most v-1
		- $\triangleright$  if this vertex has an edge to all other vertices
	- o at most e
		- $\triangleright$  there cannot be more neighbors than edges in the graph
- A vertex has O(min(v,e)) neighbors  $\circ$  iterating through the neighbors costs  $O(min(v,e))$  $\triangleright$  times the cost of the operation being performed



## Naïve Graph Implementations

#### **•** In summary



## Classic Graph Implementations

- Can we do better?
- Two representations of graphs are commonly used o the adjacency matrix representation o the adjacency list representation

"adjacency" is just a fancy word for neighbors

• Both give us better cost … in different situations …

# The Adjacency Matrix Representation

• Represent the graph as a  $v^*v$  matrix of booleans  $\circ$  M[i,j] == true if there is an edge between i and j  $\circ$  M[i,j] == false otherwise M is called the **adjacency matrix**



• Space needed: O(v<sup>2</sup>)





## The Adjacency List Representation

- For each vertex v, keep track of its neighbors in a list
	- o the **adjacency list** of v
- Store the adjacency lists in a vertex-indexed array
- Cost of the operations o graph\_hasedge(G, v, w): O(min(v,e))  $\triangleright$  each vertex has  $O(min(v,e))$  neighbors  $\triangleright$  each adjacency list has length O(min(v,e)) o graph\_addedge(G, v, w): O(1)  $\triangleright$  add v in w's list and w in v's list o graph\_get\_neighbors(G, v): O(1)
	- $\triangleright$  just grab v's adjacency list



0

3

2

4

 $\overline{1}$ 

## The Adjacency List Representation

 *For each vertex v, keep track of its neighbors in a list*

o *the adjacency list of v*

- *Store the adjacency lists in a vertex-indexed array*
- $\bullet$  Space needed:  $O(v + e)$ 
	- o a v-element array
	- o 2e list items
		- $\triangleright$  each edge corresponds to exactly 2 list items
- $\bullet$  O(v + e) is conventionally written O(max(v,e))

Why? Note that  $max(v,e) \le v+e \le 2max(v,e)$ 

0

1  $\leftrightarrow$  4

 $0$   $\rightarrow$  2  $\rightarrow$  4

0

3

2

4

1

1  $\rightarrow$  4  $\rightarrow$  3

0 1 2

2

1

2

3

4

## Adjacency Matrix vs. List



## When to Use What Representation?

- Recall that  $0 \le e \le v(v-1)/2$
- A graph is **dense** if it has lots of edges  $\circ$  e is on the order of  $v^2$
- A graph is **sparse** if it has relatively few edges  $\circ$  e is in  $O(v)$ 
	- $\Box$  at most O(v log v)
	- $\triangleright$  but definitely not  $O(v^2)$
	- o lots of graphs are sparse
		- $\triangleright$  social networks
		- $\triangleright$  roads between cities

 $\triangleright$  ...

## Cost in Dense Graphs

 $\bullet$  We replace e with  $v^2$  and simplify



## Cost in Dense Graphs

- graph\_hasedge is faster with AM
- graph\_get\_neighbors is faster with AL o but we typically iterate through the neighbors after grabbing them
- All other operations are the same
- The space requirements are the same
- For dense graphs

o the two representations have about the same cost o but graph\_hasedge is faster with AM the adjacency matrix representation is preferable

## Cost in Sparse Graphs

• We replace e with **v** and simplify





## Cost in Sparse Graphs

- AL requires **a lot less space**
- graph\_hasedge is faster with AM
- graph\_get\_neighbors is faster with AL
	- o but we typically iterate through the neighbors after grabbing them
- All other operations are the same
- For sparse graphs
	- o AL uses substantially less space
	- o the two representations have about the same cost
	- o but graph\_hasedge is faster with AM

the adjacency list representation is preferable because it doesn't require as much space

#### **Adjacency List Implementation**

# Graph Types

- An adjacency list is just a NULL-terminated linked list of vertices
- The graph data structure consists of
	- o the number v of vertices in the graph
		- $\triangleright$  field size
	- o a v-element array of adjacency lists
		- $\triangleright$  field adjlist





#### Representation Invariants

• The interface defines

typedef unsigned int vertex;

 A vertex is valid if its value is between 0 and the size of the graph



Representation Invariants

• A graph is valid if o it is non-NULL



o the length of the array of adjacency lists is equal to it size

#### **but we can't check this in C**

o each adjacency list is valid

```
bool is_graph(graph *G) {
 if (G == NULL) return false;
 //@assert(G->size == \length(G->adj));
 for (unsigned int i = 0; i < G->size; i++) {
  if (!is_adjlist(G, i, G->adj[i])) return false;
 }
 return true;
}
```
## Representation Invariants

- An adjacency list is valid if o it is NULL-terminated
	- o each vertex is valid
	- o there are not self-edges
	- o every outgoing edge has a corresponding edge coming back in
	- o there are no duplicate edges



```
bool is_adjlist(graph *G, vertex v, adjlist *L) {
 REQUIRES(G != NULL);
 //@requires(G->size == \length(G->adj));
 if (!is_acyclic(L)) return false;
```
while  $(L != NULL)$  {

vertex  $w = L$ ->vert; // w is a neighbor of v

// Neighbors are legal vertices if (!is\_vertex(G, w)) return false;

// No self-edges if  $(v == w)$  return false;

// Every outgoing edge has a corresponding edge coming back to it if (!is\_in\_adjlist(G->adj[w], v)) return false;

// Edges aren't duplicated if (is\_in\_adjlist(L->next, w)) return false;

```
L = L->next;
}
```
return true;

}

#### Basic operations

**• graph\_size returns the stored size**  $\circ$  Cost O(1)

- **•** graph\_new creates an array of empty adjacency lists
	- o calloc makes it convenient
	- o Cost O(v)
		- $\triangleright$  calloc needs to zero out all v positions

unsigned int graph\_size(graph \*G) { REQUIRES(is\_graph(G)); return G->size; }

graph \*graph\_new(unsigned int size) { graph  $*G =$  xmalloc(sizeof(graph));  $G\rightarrow$ size = size;  $G$ ->adj = xcalloc(size, sizeof(adjlist\*)); ENSURES(is\_graph(G)); return G; }



## Freeing a Graph

- graph\_free must free o all adjacency lists  $\circ$  the array
	- o the graph header





## Checking Edges

 graph\_hasedge(G, v, w) does a linear search for w in the adjacency list of v

}

o we could implement it the other way around as well

- Its cost is O(min(v,e)) o the maximum length of an adjacency list
	- o the maximum number of neighbors of a vertex

bool graph\_hasedge(graph \*G, vertex v, vertex w) { REQUIRES(is\_graph(G)); REQUIRES(is\_vertex(G, v) && is\_vertex(G, w));

for (adjlist  $L = G$ ->adj[v]; L != NULL; L = L->next) { if  $(L\text{-}>\text{vert} == w)$  return true; }

return false;



## Adding an Edge

- The preconditions exclude o self-edges
	- o edges already contained in the graph
- graph\_addedge(G, v, w) o adds w as a neighbor of v o and v as a neighbor of w





Constant cost

- We can use the adjacency list of a vertex as a representation of its neighbors
	- o We must be careful however not to modify the graph as we iterate through the neighbors

o Define a struct with a single field

 $\triangleright$  a pointer to the next neighbor to examine



}

- graph\_get\_neighbors(G, v)
	- o creates a neighbors struct
	- o points the next\_neighbor fields to the adjacency list of v
	- o returns this struct

neighbors \*graph\_get\_neighbors(graph \*G, vertex v) { REQUIRES(is\_graph(G) && is\_vertex(G, v));

neighbors \*nbors = xmalloc(sizeof(neighbors)); nbors->next\_neighbor = G->adj[v]; ENSURES(is\_neighbors(nbors)); return nbors;



#### Constant cost

}

#### ● graph\_next\_neighbor

- o returns the next neighbor
- o advances the next\_neighbor field along the adjacency list

It **must not** free that adjacency list node since it is owned by the graph vertex graph\_next\_neighbor(neighbors \*nbors) { REQUIRES(is\_neighbors(nbors)); REQUIRES(graph\_hasmore\_neighbors(nbors));

vertex v = nbors->next\_neighbor->vert; nbors->next\_neighbor = nbors->next\_neighbor->next; return v;



Constant cost

}

}

• graph\_hasmore\_neighbors checks whether the end of the adjacency list has been reached

bool graph\_hasmore\_neighbors(neighbors \*nbors) { REQUIRES(is\_neighbors(nbors)); return nbors->next\_neighbor != NULL;

- graph\_free\_neighbors frees the neighbor header
	- o and **only** the header

It must not free the rest of the adjacency list since it is owned by the graph

void graph\_free\_neighbors(neighbors \*nbors) { REQUIRES(is\_neighbors(nbors)); free(nbors);

Constant time

### Cost Summary



#### **Using the Graph Interface**

# Printing a Graph

• Using the graph interface, write a client function that prints a graph

o for every vertex

 $\triangleright$  print it

 $\triangleright$  print every neighbor of this node

```
void graph_print(graph_t G) {
 for (vertex v = 0; v < graph_size(G); v++) {
  printf("Vertices connected to %u: ", v);
  neighbors_t nbors = graph_get_neighbors(G, v);
  while (graph_hasmore_neighbors(nbors)) {
   vertex w = graph_next_neighbor(nbors);
   printf("\%u,", w);
  }
  graph_free_neighbors(nbors);
  printf("\n");
 }
}
                                    |w is a neighbor of v|
```
typedef unsigned int vertex; typedef struct graph\_header \*graph\_t; graph\_t graph\_new(unsigned int numvert); //@ensures \result != NULL; void graph\_free(graph\_t G); //@requires G != NULL; unsigned int graph\_size(graph\_t G); //@requires G != NULL; bool graph\_hasedge(graph\_t G, vertex v, vertex w); //@requires G != NULL; //@requires  $v <$  qraph size(G) && w < graph size(G); void graph\_addedge(graph\_t G, vertex v, vertex w); //@requires G != NULL; //@requires v < graph\_size(G) && w < graph\_size(G); //@requires  $v = w \&&$  !graph\_hasedge(G, v, w); typedef struct neighbor\_header \*neighbors\_t; neighbors\_t graph\_get\_neighbors(graph\_t G, vertex v); //@requires G != NULL  $88 v <$  graph\_size(G);  $1/10$  ensures \result != NULL: bool graph\_hasmore\_neighbors(neighbors\_t nbors);  $\mathcal{U} \mathcal{Q}$  requires nbors  $!=$  NULL; vertex graph\_next\_neighbor(neighbors\_t nbors);  $\mathcal{U}$ @requires nbors != NULL; //@requires graph\_hasmore\_neighbors(nbors); //@ensures is\_vertex(\result);

**graph.h**

 $\bigcap$ 

5

void graph\_free\_neighbors(neighbors\_t nbors);  $\mathcal{W}$ @requires nbors != NULL;

● We will see other algorithms that follow this pattern

- For a graph with v vertices and e edges
- using a library based on the **adjacency list** representation



 $\bullet$  So the cost of graph print is O(v min(v, e))



- The cost of graph print is  $O(v \ min(v, e))$ o for a graph with v vertices and e edges using adjacency lists
- $\bullet$  Is that right?
	- $\circ$  We assumed every vertex has  $O(min(v,e))$  neighbors
	- o But **overall** graph\_print visits every edge exactly twice
		- $\triangleright$  once from each endpoint
		- $\triangleright$  the body of the inner loop runs 2e times over all iterations of the outer loop
		- $\triangleright$  the entire inner loop costs O(e)



• The entire inner loop costs O(e)



• The actual cost of graph\_print is  $O(v + e)$ o for a graph with v vertices and e edges **using adjacency lists**

- Using the adjacency matrix representation
- $\bullet$  By the same argument, the entire inner loop costs  $O(e)$ o and graph\_free\_neighbors too



• The actual cost of graph\_print is  $O(v^2 + e)$  $\circ$  This is O(v<sup>2</sup>) since  $e \in O(v^2)$  always

## What is the Cost of print\_graph?

- Adjacency list representation:  $O(v + e)$
- Adjacency matrix representation: O(v<sup>2</sup>)



- For a dense graph  $\triangleright$  e  $\in$  O(v<sup>2</sup>) they are the same
- **•** For a sparse graph, AL is better

```
void graph_print(graph_t G) {
 for (vertex v = 0; v < graph_size(G); v++) {
  printf("Vertices connected to %u: ", v);
  neighbors_t nbors = graph_get_neighbors(G, v);
  while (graph_hasmore_neighbors(nbors)) {
   vertex w = graph_next_neighbor(nbors);
   printf("\%u,", w);
  }
  graph_free_neighbors(nbors);
  printf("\n");
 }
}
```