# Graph-theoretic Models

Lesson 5

## objectives

- basic of graphs
- how to implement graphs
- shortest path problem
- search
  - Breadth-first search (implementation 1)
  - Depth-first search (implementation)
  - Breadth-first search (implementation 2)

#### **Computational Models**

- Programs that help us understand the world and solve practical problems
- Saw how we could map the informal problem of choosing what to eat into an optimization problem, and how we could design a program to solve it
- Now want to look at class of models called graphs

#### Who owes whom money ?



#### the full graph could look like this.



#### What's a Graph?

#### Set of nodes (vertices)

- Might have properties associated with them
- Set of edges (arcs) each consisting of a pair of nodes
  - Undirected (graph)
  - Directed (digraph)
    - Source (parent) and destination (child) nodes
  - Unweighted or weighted



#### What's a Graph?

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## Why Graphs?

- To capture useful relationships among entities
  - Rail links between Paris and London
  - How the atoms in a molecule are related to one another
  - Ancestral relationships

#### Trees: An Important Special Case

- A special kind of directed graph in which any pair of nodes is connected by a single path
  - Recall the search trees we used to solve knapsack problem



## Why Graphs Are So Useful

- World is full of networks based on relationship
  - Computer networks
  - Transportation networks
  - Financial networks
  - Sewer or water networks
  - Political networks
  - Criminal networks
  - Social networks
  - etc

## Why Graphs Are So Useful

- We will see that not only do graphs capture relationships in connected networks of elements, they also support inference on those structures
  - Finding sequences of links between elements is there a path from A to B
  - Finding the least expensive path between elements (aka shortest path problem)
  - Partitioning the graph into sets of connected elements (aka graph partition problem)
  - Finding the most efficient way to separate sets of connected elements (aka the min-cut/max-flow problem)

# Graph Theory Saves Me Time Every Day



- Model road system using a digraph
  - Nodes: points where roads end or meet
  - Edges: connections between points
    - Each edge has a weight
      - Expected time to get from source node to destination node for that edge
      - Distance between source and destination nodes
      - Average speed of travel between source and destination nodes
- Solve a graph optimization problem
  - Shortest weighted path between my house and my office







#### First Reported Use of Graph Theory

- Bridges of
  Königsberg
  (1735)
- Possible to take a walk that traverses each of the 7 bridges exactly once?



#### Leonhard Euler's Model

Each island a node

Each bridge an undirected edge

Model abstracts away irrelevant details

- Size of islands
- Length of bridges



# Is there a path that contains each edge exactly once? No!

https://en.wikipedia.org/wiki/Seven\_Bridges\_of\_K%C3%B6nigsberg

## Suppose you want to travel to Golden Gate Bridge from Twin Peaks.



# What's your algorithm to find the path with the fewest steps?



#### can you get there in one step?



#### if no, can you get there in two steps?



#### if no, can you get there in three steps?



#### it takes three steps to get there.



#### summary.

- there are other routes that will get you there
- they are at least three steps long
- the algorithm found that the shortest route to the bridge is three steps long
- this type of problem is called a shortest-path problem

#### Breadth – first search

- we have discussed linear search and binary search
- BFS is a different kind of search algorithm: one that runs on graphs
- it can help answer two types of questions:
  - 1. Is there a path from node A to node B?
  - 2. What is the shortest path from node A to node B?

# Suppose you're the proud owner of a mango farm. Looking for mango seller to sell yours.

- this search is pretty straightforward
- first, make a list of friends to search





now go to each person in the list and check whether that person sells mangoes.



suppose none of your friends are mango sellers. now you have to search through your friends' friends.



each time you search for someone from the list, add all of their friends to the list.



- you not only search your triends, but you search their triends too
- the goal is to find one mago seller in your network
- if Alice isn't a mango seller, you add her friends to the list, too
- that means you'll eventually search her friends and then their friends
- with this algorithm, you'll search your entire network until you come across a mango seller. This algorithm is breadth-first search.

#### relook.

- Question 1: is there a path from node A to node B?
  - Is there a mango seller in your network?
- Question 2: what is the shortest path from node A to node B?
  - Who is the closest mango seller?

#### look deeper to question 2.

• can you find the closest mango seller?



## look deeper.

- we will prefer a first degree connection to a second degree connection, and so on
- we shouldn't search any second degree connections before we make sure we don't have a first degree connection who is a mango seller.
- breadth first search already does this!
  - the way breadth-first search works, the search radiates out from the starting point
  - notice that this only works if we search people in the same order in which they are added.
  - we need to search people in the order that they are added.
    - there's a data structure for this: it's called a **<u>queue</u>**

#### Queues

- same concept as real life
- suppose you and your friend are queuing up at the bus stop. if you are before him in the queue, you get on the bus first.
- two operations in queue
  - enqueue
  - dequeue



#### Queues



- if we enqueue two items to the list, the first item we add will be dequeued before the second item
- we can use this for our search list
- the queue is called a FIFO data structure: First in, First Out.



FIFO (FIRST IN, FIRS
## let's try this simple graph.

START

• Find the length of the shortest path from start to finish.

 Find the length of the shortest path from "cab" to "bat"



## Implementing the graph.

- use a dictionary
  - key, value pair
- we want to map a node to all of its neighbours

graph = {}
graph["you"] = ["alice", "bob", "claire"]





## a bigger graph.



```
graph = {}
graph["you"] = ["alice", "bob", "claire"]
graph["bob"] = ["anuj", "peggy"]
graph["alice"] = ["peggy"]
graph["claire"] = ["thom", "jonny"]
graph["anuj"] = []
graph["peggy"] = []
graph["thom"] = []
```

## directed graph vs undirected graph

- Anuj, Peggy, Thom and Jonny don't have any neighbors.
- they have arrows pointing to them, but no arrows from them to someone else
- this is called a directed graph
- an undirected graph doesn't have any arrows, and both nodes are each other's neighbours.
  - For eg, both of these graphs are equal



## implementing the algorithm



1. KEEP A QUEUE CONTAINING THE PEOPLE TO CHECK



2. POP A PERSON OFF THE QUEUE



3. CHECK IF THIS PERSON IS A MANGO SELLER

#### Note

When updating queues, I use the terms *enqueue* and *dequeue*. You'll also encounter the terms *push* and *pop*. *Push* is almost always the same thing as *enqueue*, and *pop* is almost always the same thing as *dequeue*.



## implementation.

• make a queue to start. in python, use the double ended queue (deque) function for this:

from collections import deque
search\_queue = deque() 
search\_queue += graph["you"] 
Adds all of your neighbors to the search queue

• remember, graph["you"] will give you a list of all your neighbours, like ["alice', "bob", "Claire"]. those all get added to the search queue.



## implementation.

while search\_queue: < While the queue isn't empty ...

person = search\_queue.popleft() < ... grabs the first person off the queue **if** person\_is\_seller(person): < Checks whether the person is a mango seller print person + " is a mango seller!" <----- Yes, they're a mango seller. return True

#### else:

search\_queue += graph[person] < ..... No, they aren't. Add all of this return False the queue was a mango seller.

person's friends to the search queue.

```
def person_is_seller(name):
    return name [-1] == 'm'
```



## in action . .

search.queve += graph [person]

while search-queue:

person = search\_queve.popleft()

if person\_is.seller(person): else: search.queue += graph[person]



...etc...

## when will it ends.

- the algorithm will keep going until either:
  - a mango seller is found, or
  - the queue becomes empty, in which case there is no mango seller.

## potential problem.

- alice and bob share a friend Peggy. So Peggy will be added to the queue twice:
  - once when you add Alice's friends and
  - again when you add Bob's friends
- if you check her twice, you're doing unnecessary, extra work. So once you search a person, you should mark that person as searched and not search them again



TWICE |

• if you don't do this, you could end up in an infinite loop

# Suppose the mango seller graph looked like this

To start, the search queue contains all of your neighbors.



Now you check Peggy. She isn't a mango seller, so you add all of her neighbors to the search queue.



Next, you check yourself. You're not a mango seller, so you add all of your neighbors to the search queue.



And so on. This will be an infinite loop, because the search queue will keep going from you to Peggy.



## final code.

```
def search(name):
    search_queue = deque()
    search_queue += graph[name]
    searched = [] <
                                             which people you've searched before.
   while search_queue:
       person = search_queue.popleft()
       if not person in searched: 
Only search this person if you
                                             haven't already searched them.
           if person_is_seller(person):
               print person + " is a mango seller!"
               return True
           else:
               search_queue += graph[person]
               searched.append(person) < Marks this person as searched
   return False
```

## Implementing and using graphs

#### Building graphs

- Nodes
- Edges
- Stitching together to make graphs

#### Using graphs

- Searching for paths between nodes
- Searching for optimal paths between nodes

## Class Node

```
class Node(object):
    def __init__(self, name):
        """Assumes name is a string"""
        self.name = name
    def getName(self):
        return self.name
    def __str__(self):
        return self.name
```

## **Class Edge**

```
class Edge(object):
    def __init__(self, src, dest):
        """Assumes src and dest are nodes"""
        self.src = src
        self.dest = dest
    def getSource(self):
        return self.src
    def getDestination(self):
        return self.dest
    def __str_(self):
        return self.src.getName() + ' \rightarrow ' \setminus
                + self.dest.getName()
```

## **Common Representations of Digraphs**

#### Digraph is a directed graph

Edges pass in one direction only

#### Adjacency matrix

- Rows: source nodes
- Columns: destination nodes
- Cell[s, d] = 1 if there is an edge from s to d
  - = 0 otherwise
- Note that in digraph, matrix is not symmetric

#### Adjacency list

Associate with each node a list of destination nodes

#### Class Digraph, part 1



#### Class Digraph, part 2

def childrenOf(self, node):
 return self.edges[node]

```
def hasNode(self, node):
    return node in self.edges
```

```
def getNode(self, name):
    for n in self.edges:
        if n.getName() == name:
            return n
        raise NameError(name)
```

```
def __str__(self):
    result = ''
    for src in self.edges:
        for dest in self.edges[src]:
            result = result + src.getName() + '->'\
                 + dest.getName() + '\n'
                 return result[:-1] #omit final newline
```

### Class Graph

```
class Graph(Digraph):
    def addEdge(self, edge):
        Digraph.addEdge(self, edge)
        rev = Edge(edge.getDestination(), edge.getSource())
        Digraph.addEdge(self, rev)
```

Graph does not have directionality associated with an edge

- Edges allow passage in either direction
- Why is Graph a subclass of Digraph?
- Remember the substitution rule?
  - If client code works correctly using an instance of the supertype, it should also work correctly when an instance of the subtype is substituted for the instance of the supertype

 Any program that works with a Digraph will also work with a Graph (but not vice versa)

## A Classic Graph Optimization Problem

#### Shortest path from n1 to n2

- Shortest sequence of edges such that
  - Source node of first edge is n1
  - Destination of last edge is n2
  - For edges, e1 and e2, in the sequence, if e2 follows e1 in the sequence, the source of e2 is the destination of e1

#### Shortest weighted path

Minimize the sum of the weights of the edges in the path

#### Some Shortest Path Problems

Finding a route from one city to another

- Designing communication networks
- Finding a path for a molecule through a chemical labyrinth



•...



#### An Example



## Build the Graph

```
def buildCityGraph(graphType):
   g = graphType()
    for name in ('Boston', 'Providence', 'New York', 'Chicago',
                 'Denver', 'Phoenix', 'Los Angeles'): #Create 7 nodes
        g.addNode(Node(name))
   g.addEdge(Edge(g.getNode('Boston'), g.getNode('Providence')))
   g.addEdge(Edge(g.getNode('Boston'), g.getNode('New York')))
   g.addEdge(Edge(g.getNode('Providence'), g.getNode('Boston')))
   g.addEdge(Edge(g.getNode('Providence'), g.getNode('New York')))
   g.addEdge(Edge(g.getNode('New York'), g.getNode('Chicago')))
   g.addEdge(Edge(g.getNode('Chicago'), g.getNode('Denver')))
    g.addEdge(Edge(g.getNode('Chicago'), g.getNode('Phoenix')))
    g.addEdge(Edge(g.getNode('Denver'), g.getNode('Phoenix')))
    g.addEdge(Edge(g.getNode('Denver'), g.getNode('New York')))
```

g.addEdge(Edge(g.getNode('Los Angeles'), g.getNode('Boston')))

## Finding the Shortest Path

Algorithm 1, depth-first search (DFS)

- Similar to left-first depth-first method of enumerating a search tree (Lecture 2)
- Main difference is that graph might have cycles, so we must keep track of what nodes we have visited to avoid going in infinite loops

Note that we are using divide-and-conquer: if we can find a path from a source to an intermediate node, and a path from the intermediate node to the destination, the combination is a path from source to destination

## Depth First Search

Start at an initial node

Consider all the edges that leave that node, in some order

•Follow the first edge, and check to see if at goal node

If not, repeat the process from new node

## Continue until either find goal node, or run out of options

 When run out of options, backtrack to the previous node and try the next edge, repeating this process

### Depth First Search (DFS)

```
returning to this point in the
                                                        recursion to try next node
def DFS(graph, start, end, path, shortest, toPrint = False):
    path = path + [start]
                                                               Note how will explore
   if toPrint:
                                                                all paths through first
        print('Current DFS path:', printPath(path))
    if start == end:
                                                                  node, before ...
        return path
    for node in graph.childrenOf(start):
        if node not in path: #avoid cycles
            if shortest == None or len(path) < len(shortest):
                newPath = DFS(graph, node, end, path, shortest, toPrint)
                if newPath != None:
                    shortest = newPath
        elif toPrint:
            print('Already visited', node)
    return shortest
```

```
def shortestPath(graph, start, end, toPrint = False):
    return DFS(graph, start, end, [], None, toPrint)
```

## Test DFS

#### An Example



#### Output (Chicago to Boston)



#### Output (Boston to Phoenix)

Current DFS path: Boston

Current DFS path: Boston->Providence

Already visited Boston

Current DFS path: Boston->Providence->New York

Current DFS path: Boston->Providence->New York->Chicago

Current DFS path: Boston->Providence->New York->Chicago->Denver

Current DFS path: Boston->Providence->New York->Chicago->Denver->Phoenix Found path

Already visited New York

Current DFS path: Boston->Providence->New York->Chicago->Phoenix Found a shorter path

Current DFS path: Boston->New York

Current DFS path: Boston->New York->Chicago

Current DFS path: Boston->New York->Chicago->Denver

Current DFS path: Boston->New York->Chicago->Denver->Phoenix Found a "shorter" path Already visited New York

Current DFS path: Boston->New York->Chicago->Phoenix Found a shorter path

Shortest path from Boston to Phoenix is Boston->New York->Chicago->Denver->Phoenix

#### **Breadth First Search**

Start at an initial node

- Consider all the edges that leave that node, in some order
  - •Follow the first edge, and check to see if at goal node
- If not, try the next edge from the current node
  - Continue until either find goal node, or run out of options
    - When run out of edge options, move to next node at same distance from start, and repeat
    - When run out of node options, move to next level in the graph (all nodes one step further from start), and repeat

## Algorithm 2: Breadth-first Search (BFS)

```
def BFS(graph, start, end, toPrint = False):
    initPath = [start]
    pathQueue = [initPath]
    while len(pathQueue) != 0:
        #Get and remove oldest element in pathQueue
        tmpPath = pathQueue.pop(0)
        if toPrint:
            print('Current BFS path:', printPath(tmpPath))
        lastNode = tmpPath[-1]
        if lastNode == end:
            return tmpPath 🔶
        for nextNode in graph.childrenOf(lastNode):
            if nextNode not in tmpPath:
                newPath = tmpPath + [nextNode]
                pathQueue.append(newPath)
    return None
```

## Output (Boston to Phoenix)

Current BFS path: Boston

Current BFS path: Boston->Providence

Current BFS path: Boston->New York

Current BFS path: Boston->Providence->New York

Current BFS path: Boston->New York->Chicago

Current BFS path: Boston->Providence->New York->Chicago

Current BFS path: Boston->New York->Chicago->Denver

Current BFS path: Boston->New York->Chicago->Phoenix

Shortest path from Boston to Phoenix is Boston->New York->Chicago->Phoenix
#### Output (Boston to Pheonix)



# What About a Weighted Shortest Path

- Want to minimize the sum of the weights of the edges, not the number of edges
- DFS can be easily modified to do this
- BFS cannot, since shortest weighted path may have more than the minimum number of hops

### Recap

Graphs are cool

- Best way to create a model of many things
  - Capture relationships among objects
- Many important problems can be posed as graph optimization problems we already know how to solve
- Depth-first and breadth-first search are important algorithms
  - Can be used to solve many problems

## terms to find out more.

- weighted graph
- cycles
- tree
- complete graph
- clique
- bipartite graph

• Dijkstra's algorithm

### Finger Exercises

Modify the DFS algorithm to find a path that minimizes the sum of the weights. Assume that all weights are positive integers.

Consider a digraph with weighted edges. Is the first path found by BFS guaranteed to minimize the sum of the weights of the edges?