

# Project 10: Graphs Part 2

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**Due: Thursday, April 21st @ 10:00 pm**

*This is not a team project, do not copy someone else's work.*

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## Assignment Overview

### What is A\* Search?

Instead of searching the "next closest" vertex as is done in Dijkstra's algorithm, [A\\* Search](#) (A-Star Search) picks the vertex which is "next closest to the goal" by weighting vertices more cleverly.

Recall that in Dijkstra's algorithm, vertices are stored in a priority queue with a priority key equal to the current shortest path to that vertex. If we denote the current shortest path to a vertex  $V$  by  $g(v)$ , then on each iteration of Dijkstra's algorithm, we search on the vertex with  $\min(g(v))$ .

A\* search takes the same approach to selecting the next vertex, but instead of setting the priority key of a vertex equal to  $g(v)$  and selecting  $\min(g(v))$ , it uses the value  $f(v)$  and selects the vertex with  $\min(f(v))$  where

$$f(v) = g(v) + h(v)$$

$$= \text{current\_shortest\_path\_to\_v} + \text{estimated\_distance\_between\_v\_and\_target}$$

In English, Please...

A\* Search prioritizes vertices  $V$  to search based on the value  $f(v)$ , which is the sum of

- $g(v)$ , or the current shortest known path to vertex  $V$ , and
- $h(v)$ , which is the estimated (Euclidean or Taxicab) distance between the vertex  $V$  and the target vertex you're searching for

The result is that A\* prioritizes vertices to search that (1) are *close to the origin along a known path* AND which (2) are *in the right direction towards the target*. Vertices with a small  $g(v)$  are *close to the origin along a known path* and vertices with a small  $h(v)$  are *in the right direction towards the target*<sup>\*\*</sup>,<sup>\*\*</sup> so we pick vertices with the smallest sum of these two values.

[We strongly recommend you watch this video to build your intuition behind A\\* Search!](#)

[A\* is extremely versatile. Here we use Euclidean and Taxicab distances to prioritize certain directions of search, but note that any [metric](#)  $h(v, \text{target})$  could be used should the need arise. See [here](#) for more information on situations where different metrics may be practical.] For additional information on Graphs,

Shortest path and A\* please go to [D2L-Week-10-11](#)



## Assignment Notes

- A plotting function is provided to help you visualize the progression of various search algorithms
  - Be sure to read the specs explaining **plot()**
  - If you don't want to use it, just comment out the related import statements and **plot()** function
- Python allows representation of the value infinity using **float('inf')**
- No negative edge weights will ever be added to the graph
  - All edge weights are numeric values greater than or equal to zero
- Time complexities are specified in terms of  $V$  and  $E$ , where  $V$  represents the number of vertices in the graph and  $E$  represents the number of edges in a graph
  - Recall that  $E$  is bounded above by  $V^2$ ; a graph has  $E = V^2$  edges if and only if every vertex is connected to every other vertex
- Recall that **list.insert(0, element)** and **list.pop(0)** are both  $O(N)$  calls on a Python list
  - Recall that python's 'lists' are not lists in the more common sense of the word: linked lists. They are dynamically managed tuples, stored in memory as contiguous arrays of pointers to elements elsewhere in memory. This allows indexing into a 'list' in constant time. The downside of this, however, is that adding to a python 'list' at a specific index,  $i$ , requires shifting the pointer to every element past  $i$  by one in the underlying array: a linear operation.
  - Be careful when implementing **dijkstra**, **astar**, and the Application Problem to ensure you do not break time complexity by popping or inserting from the front of a list when reconstructing a path
  - Instead of inserting into / popping from the front of the list, simply append to or pop from the end, then reverse the list *once* at the end
    - If you have  $N$  calls to **list.insert(0, element)**, that is  $O(N^2)$
    - If you instead have  $N$  calls to **list.append(element)**, followed by a single call to **list.reverse()**, that is  $O(N)$
    - Both methods will result in the same list being constructed, but the second is far more efficient

## Assignment Specifications

class Vertex

Represents a vertex object, the building block of a graph.

**DO NOT MODIFY the following attributes/functions**

- **Attributes**

- **id**: A string used to uniquely identify a vertex
- **adj**: A dictionary of type **{other\_id : number}** which represents the connections of a vertex to other vertices; the existence of an entry with key **other\_id** indicates connection from this vertex to the vertex with id **other\_id** by an edge with weight **number**
  - Note that as of Python 3.7, [insertion ordering](#) in normal dictionaries is guaranteed and ensures traversals will select the next neighbor to visit deterministically
- **visited**: A boolean flag used in search algorithms to indicate that the vertex has been visited
- **x**: The x-position of a vertex (used in assignment) (defaults to zero)
- **y**: The y-position of a vertex (used in assignment) (defaults to zero)

- **\_\_init\_\_(self, idx: str, x: float=0, y: float=0) -> None:**

- Constructs a Vertex object

- **\_\_eq\_\_(self, other: Vertex) -> bool:**

- Compares this vertex for equality with another vertex

- **\_\_repr\_\_(self) -> str:**

- Represents the vertex as a string for debugging

- **\_\_str\_\_(self) -> str:**

- Represents the vertex as a string for debugging

- **\_\_hash\_\_(self) -> int:**

- Allows the vertex to be hashed into a set; used in unit tests

**USE the following function however you'd like**

- **deg(self) -> int:**

- Returns the number of outgoing edges from this vertex; i.e., the outgoing degree of this vertex
- *Time Complexity:  $O(1)$*
- *Space Complexity:  $O(1)$*

- **get\_outgoing\_edges(self) -> Set[Tuple[str, float]]:**

- Returns a **set** of tuples representing outgoing edges from this vertex
- Edges are represented as tuples (**other\_id**, **weight**) where
  - **other\_id** is the unique string id of the destination vertex
  - **weight** is the weight of the edge connecting this vertex to the other vertex
- Returns an empty set if this vertex has no outgoing edges
- *Time Complexity:  $O(degV)$*
- *Space Complexity:  $O(degV)$*

- **euclidean\_distance(self, other: Vertex) -> float:**

- Returns the [euclidean distance](#) [based on two-dimensional coordinates] between this vertex and vertex **other**
- Used in AStar
- *Time Complexity:  $O(1)$*
- *Space Complexity:  $O(1)$*

- **taxicab\_distance(self, other: Vertex) -> float:**

- Returns the [taxicab distance](#) [based on two-dimensional coordinates] between this vertex and vertex **other**

- Used in AStar
- *Time Complexity:  $O(1)$*
- *Space Complexity:  $O(1)$*

## class Graph

Represents a graph object

**DO NOT MODIFY the following attributes/functions**

- **Attributes**

- **size:** The number of vertices in the graph
- **vertices:** A dictionary of type **{id : Vertex}** storing the vertices of the graph, where **id** represents the unique string id of a **Vertex** object
  - Note that as of Python 3.7, [insertion ordering](#) in normal dictionaries is guaranteed and ensures **get\_edges(self)** and **get\_vertices(self)** will return deterministically ordered lists
- **plot\_show:** If true, calls to **plot()** display a rendering of the graph in matplotlib; if false, all calls to **plot()** are ignored (see **plot()** below)
- **plot\_delay:** Length of delay in **plot()** (see **plot()** below)

- **\_\_init\_\_(self, plt\_show: bool=False) -> None:**

- Constructs a Graph object
- Sets **self.plot\_show** to False by default

- **\_\_eq\_\_(self, other: Graph) -> bool:**

- Compares this graph for equality with another graph

- **\_\_repr\_\_(self) -> str:**

- Represents the graph as a string for debugging

- **\_\_str\_\_(self) -> str:**

- Represents the graph as a string for debugging

- **add\_to\_graph(self, start\_id: str, dest\_id: str=None, weight: float=0) -> float:**

- Adds a vertex / vertices / edge to the graph
  - Adds a vertex with id **start\_id** to the graph if no such vertex exists
  - Adds a vertex with id **dest\_id** to the graph if no such vertex exists and **dest\_id** is not None
  - Adds an edge with weight **weight** if **dest\_id** is not None
- If a vertex with id **start\_id** or **dest\_id** already exists in the graph, this function will not overwrite that vertex with a new one
- If an edge already exists from vertex with id **start\_id** to vertex with id **dest\_id**, this function will overwrite the weight of that edge

- **matrix2graph(self, matrix: Matrix) -> None:**

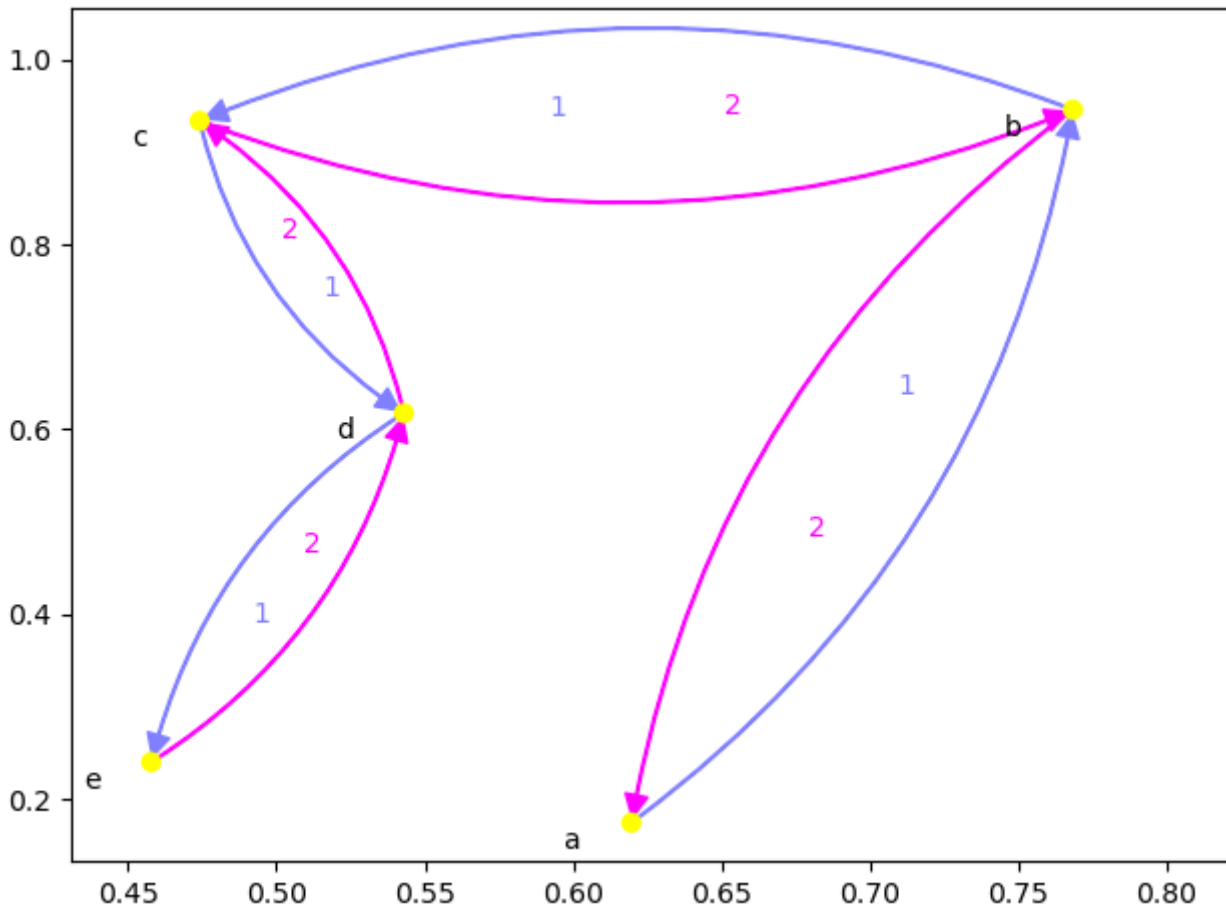
- Constructs a graph from a given adjacency matrix representation

- **matrix** is guaranteed to be a square 2D list (i.e. list of lists where # rows = # columns), of size **[V+1] x [V+1]**
  - **matrix[0][0]** is None
  - The first row and first column of **matrix** hold string ids of vertices to be added to the graph and are symmetric, i.e. **matrix[i][0] = matrix[0][i]** for  $i = 1, \dots, n$
  - **matrix[i][j]** is None if no edge exists from the vertex **matrix[i][0]** to the vertex **matrix[0][j]**
  - **matrix[i][j]** is a **number** if an edge exists from the vertex **matrix[i][0]** to the vertex **matrix[0][j]** with weight **number**
- **graph2matrix(self) -> None:**
  - Constructs and returns an adjacency matrix from a graph
  - The output matches the format of matrices described in **matrix2graph**
  - If the graph is empty, returns **None**
- **graph2csv(self, filepath: str) -> None:**
  - Encodes the graph (if non-empty) in a csv file at the given location
- **reset\_vertices(self) -> None:**
  - Resets visited flags to False of all vertices within the graph
  - Used in unit tests to reset graphs between tests
  - Time Complexity:  $O(V)$
  - Space Complexity:  $O(V)$
- **get\_vertex\_by\_id(self, v\_id: str) -> Vertex:**
  - Returns the vertex object with id **v\_id** if it exists in the graph
  - Returns None if no vertex with unique id **v\_id** exists
  - Time Complexity:  $O(1)$
  - Space Complexity:  $O(1)$
- **get\_all\_vertices(self) -> Set[Vertex]:**
  - Returns a set of all Vertex objects held in the graph
  - Returns an empty set if no vertices are held in the graph
  - Time Complexity:  $O(V)$
  - Space Complexity:  $O(V)$
- **get\_edge\_by\_ids(self, begin\_id: str, end\_id: str) -> Tuple[str, str, float]:**
  - Returns the edge connecting the vertex with id **begin\_id** to the vertex with id **end\_id** in a tuple of the form (**begin\_id**, **end\_id**, **weight**)
  - If the edge or either of the associated vertices does not exist in the graph, returns None
  - Time Complexity:  $O(1)$
  - Space Complexity:  $O(1)$
- **get\_all\_edges(self) -> Set[Tuple[str, str, float]]:**

- Returns a set of tuples representing all edges within the graph
  - Edges are represented as tuples (begin\_id, end\_id, weight) where
    - begin\_id is the unique string id of the starting vertex
    - end\_id is the unique string id of the starting vertex
    - weight is the weight of the edge connecting the starting vertex to the destination vertex
  - Returns an empty set if the graph is empty
  - Time Complexity:  $O(V+E)$
  - Space Complexity:  $O(E)$
- **build\_path(self, back\_edges: Dict[str, str], begin\_id: str, end\_id: str) -> Tuple[List[str], float]:**
    - Given a dictionary of back-edges (a mapping of vertex id to predecessor vertex id), reconstructs the path from start\_id to end\_id and computes the total distance
    - Returns tuple of the form ([path], distance) where [path] is a list of vertex id strings beginning with begin\_id, terminating with end\_id, and including the ids of all intermediate vertices connecting the two distance is the sum of the weights of the edges along the [path] traveled
    - Handle the cases where no path exists from vertex with id begin\_id to vertex with end\_id or if one of the vertices does not exist
    - Time Complexity:  $O(V)$
    - Space Complexity:  $O(V)$

***USE the following function however you'd like***

- **plot(self) -> None:**
  - Renders a visual representation of the graph using matplotlib and displays graphic in PyCharm
    - [Follow this tutorial to install matplotlib and numpy if you do not have them](#), or follow the tooltip auto-suggested by PyCharm
  - Provided for use in debugging
  - If you call this in your searches and **\*\*self.plot\_show** is true, the search process will be animated in successive plot renderings (with time between frames controlled by **self.plot\_delay**)
  - Not tested in any testcases
    - All testcase graphs are constructed with **self.plot\_show** set to False
  - If vertices have (x,y) coordinates specified, they will be plotted at those locations
  - If vertices do not have (x,y) coordinates specified, they will be plotted at a random point on the unit circle
  - To install the necessary packages (matplotlib and numpy), follow the auto-suggestions provided by PyCharm
  - Vertices and edges are labeled; edges are color-coded by weight
    - If a bi-directional edge exists between vertices, two color-coded weights will be displayed



- **reset\_vertices(self) -> None:**
  - Resets visited flags of all vertices within the graph
  - Used in unit tests to reset graph between searches
  - *Time Complexity:*  $O(V)$
  - *Space Complexity:*  $O(V)$
- **get\_vertex(self, vertex\_id: str) -> Vertex:**
  - Returns the Vertex object with id **vertex\_id** if it exists in the graph
  - Returns None if no vertex with unique id **vertex\_id** exists
  - *Time Complexity:*  $O(1)$
  - *Space Complexity:*  $O(1)$
- **get\_vertices(self) -> Set[Vertex]:**
  - Returns a **set** of all Vertex objects held in the graph
  - Returns an empty set if no vertices are held in the graph
  - *Time Complexity:*  $O(V)$
  - *Space Complexity:*  $O(V)$
- **get\_edge(self, start\_id: str, dest\_id: str) -> Tuple[str, str, float]:**
  - Returns the edge connecting the vertex with id **start\_id** to the vertex with id **dest\_id** in a tuple of the form **(start\_id, dest\_id, weight)**
  - If the edge or either of the associated vertices does not exist in the graph, returns **None**
  - *Time Complexity:*  $O(1)$
  - *Space Complexity:*  $O(1)$
- **get\_edges(self) -> Set[Tuple[str, str, float]]:**
  - Returns a **set** of tuples representing all edges within the graph
  - Edges are represented as tuples **(start\_id, other\_id, weight)** where

- **start\_id** is the unique string id of the starting vertex
- **other\_id** is the unique string id of the destination vertex
- **weight** is the weight of the edge connecting the starting vertex to the destination vertex
- Returns an empty set if the graph is empty
- *Time Complexity*:  $O(V+E)$
- *Space Complexity*:  $O(E)$

### **IMPLEMENT the following functions**

- **dijkstra(self, begin\_id: str, end\_id: str) -> Tuple[List[str], float]:**
  - Perform a dijkstra search beginning at vertex with id **begin\_id** and terminating at vertex with id **end\_id**
  - As you explore from each vertex, iterate over neighbors using **vertex.adj** (not `vertex.get_edges()`) to ensure neighbors are visited in proper order
  - Returns tuple of the form **([path], distance)** where
    - **[path]** is a list of vertex id strings beginning with **begin\_id**, terminating with **end\_id**, and including the ids of all intermediate vertices connecting the two
    - **distance** is the sum of the weights of the edges along the **[path]** traveled
  - If no path exists from vertex with id **begin\_id** to vertex with **end\_id** or if one of the vertices does not exist, returns tuple **([],0)**
  - Guaranteed that **begin\_id != end\_id** (since that would be a trivial path)
  - Because our adjacency maps use [insertion ordering](#), neighbors will be visited in a deterministic order, and thus you do not need to worry about the order in which you visit neighbor vertices at the same depth
  - Use the given PriorityQueue class to simplify priority key updates in a search priority queue
  - *Time Complexity*:  $O((E + V) \log V)$
  - *Space Complexity*:  $O(V)$
- **a\_star(self, begin\_id: str, end\_id: str, metric: Callable[[Vertex, Vertex], float]) -> Tuple[List[str], float]**
  - Perform an A\* search beginning at vertex with id **begin\_id** and terminating at vertex with id **end\_id**
  - As you explore from each vertex, iterate over neighbors using **vertex.adj** (not `vertex.get_edges()`) to ensure neighbors are visited in proper order
  - Use the **metric** parameter to estimate  $h(v)$ , the remaining distance, at each vertex
    - This is a callable parameter and will always be **Vertex.euclidean\_distance** or **Vertex.taxicab\_distance**
  - Returns tuple of the form **([path], distance)** where
    - **[path]** is a list of vertex id strings beginning with **begin\_id**, terminating with **end\_id**, and including the ids of all intermediate vertices connecting the two
    - **distance** is the sum of the weights of the edges along the **[path]** traveled
  - If no path exists from vertex with id **begin\_id** to vertex with **end\_id** or if one of the vertices does not exist, returns tuple **([],0)**
  - Guaranteed that **begin\_id != end\_id** (since that would be a trivial path)
  - Recall that vertices are prioritized in increasing order of  **$f(v) = g(v) + h(v)$** , where
    - **$g(v)$**  is the shortest known path to this vertex
    - **$h(v)$**  is the Euclidean distance from  $V$  to the target vertex
  - Use the given PriorityQueue class to simplify priority key updates in a search priority queue



- **Implementations of this function which do not utilize the heuristic metric will not receive any credit, visible and hidden testcases included.**
  - **Do not simply implement Dijkstra's Algorithm**
- *Time Complexity:  $O((E + V)\log(V))$*
- *Space Complexity:  $O(V)$*

To simplify the operation of updating a priority key in your search priority queue, use the following given class.

class PriorityQueue

*DO NOT MODIFY the following attributes/functions*

- **Attributes**
  - **data:** Underlying data list of priority queue
  - **locator:** Dictionary to locate vertices within the priority queue
  - **counter:** Used to break ties between vertices
- **\_\_init\_\_(self)**
  - Constructs an AStarPriorityQueue object
- **\_\_repr\_\_(self)**
  - Represents the priority queue as a string for debugging
- **\_\_str\_\_(self)**
  - Represents the priority queue as a string for debugging
- **empty(self)**
  - Returns boolean indicating whether priority queue is empty
- **push(self, priority, vertex)**
  - Push the **vertex** object onto the priority queue with a given **priority** key
  - This priority queue has been specially designed to hold Vertex objects as values ranked by priority keys; be sure you push Vertex objects and NOT vertex id strings onto the queue
- **pop(self)**
  - Visit, remove and return the Vertex object with the highest priority (i.e. lowest priority key) as a **(priority, vertex)** tuple
- **update(self, new\_priority, vertex)**
  - Update the priority of the **vertex** object in the queue to have a **new\_priority**

## Application Problem

After finishing your school year for the summer, you decide to take a vacation to the lovely toll-ridden city of Chicago - modeled by a road network with **v vertices and e edges**. After cleaning out your dorm room, you find you have **m EZ-pass coupons** to spend that cover the cost of **half of any toll segment**, i.e., you have m EZ-pass coupons which effectively set the toll of a road segment to HALF.

Your goal is to devise an algorithm to find a route from point **A** to point **B** on the toll roads of Chicago for the cheapest possible overall toll for your vacation.

- **tollways\_algorithm(self, start\_id: str, target\_id: str, coupons: int) -> Tuple[float, float]**
  - Perform an toll algorithm beginning at vertex with id **start\_id** and terminating at vertex with id **target\_id**
  - Assume the distance between paths is the cost of going from one to another

- As you explore from each vertex, iterate over neighbors using **vertex.adj** (not `vertex.get_edges()`) to ensure neighbors are visited in proper order
- The **coupon** parameter is the maximum number of coupons
  - This parameter will always be an integer
- Returns tuple of the form **(cost, coupons)** where
  - **cost** is the cost of the best path beginning with **start\_id**, terminating with **target\_id**
  - When halving coupon cost use [integer division](#)
  - `float('inf')` should be used for positive infinity
  - **coupons** is the sum of the number of coupons used along the best path traveled
- If no path exists from vertex with id **start\_id** to vertex with **target\_id** or if one of the vertices does not exist, returns tuple **(None, None)**
- Guaranteed that **start\_id != target\_id** (since that would be a trivial path)
- Use the given TollWayPriorityQueue class to simplify priority key updates in a search priority queue
- *Time Complexity:  $O((C*V+E)*\log(V*C))$ ,  $C$  is the number of coupons*
- *Space Complexity:  $O(V*C)$ ,  $C$  is the number of coupons*
- You can only use 1 coupon per edge
  - consider A -- 100 -- B -- 2 -- C example with 2 coupons, you can't apply both coupons to the 100 edge, you would have to apply 1 coupon to each, so the best you could do would be cost 51
- **HINTS:**
  - Use priority queue and dictionary similar to Dijkstra's but storing {cost, coupons used to reach the node, node} rather than just {distance, node}
  - When constructing the dictionary and priority queue think of a matrix  $V$  vertices rows by  $C$  columns starting at 0 coupons, similar to [Cartesian Product](#)

To simplify the operation of updating a priority key in your search priority queue, use the following given class.

class TollWayPriorityQueue:

*DO NOT MODIFY the following attributes/functions*

- **Attributes**
  - **data:** Underlying data list of priority queue
  - **locator:** Dictionary to locate vertices within the priority queue
  - **counter:** Used to break ties between vertices
- **\_\_init\_\_(self)**
  - Constructs an TollWayPriorityQueue object
- **\_\_repr\_\_(self)**
  - Represents the priority queue as a string for debugging
- **\_\_str\_\_(self)**
  - Represents the priority queue as a string for debugging
- **empty(self)**
  - Returns boolean indicating whether priority queue is empty
- **push(self, priority, vertex)**
  - Push the **vertex** object onto the priority queue with a given **priority** key

- This priority queue has been specially designed to hold Vertex objects as values ranked by priority keys; be sure you push Vertex objects and NOT vertex id strings onto the queue
- **pop(self)**
  - Visit, remove and return the Vertex object with the highest priority (i.e. lowest priority key) as a **(priority, vertex)** tuple
- **update(self, new\_priority: float, new\_coupon: int, vertex: Vertex)**
  - Update the priority of the **vertex** object in the queue to have a **new\_priority**

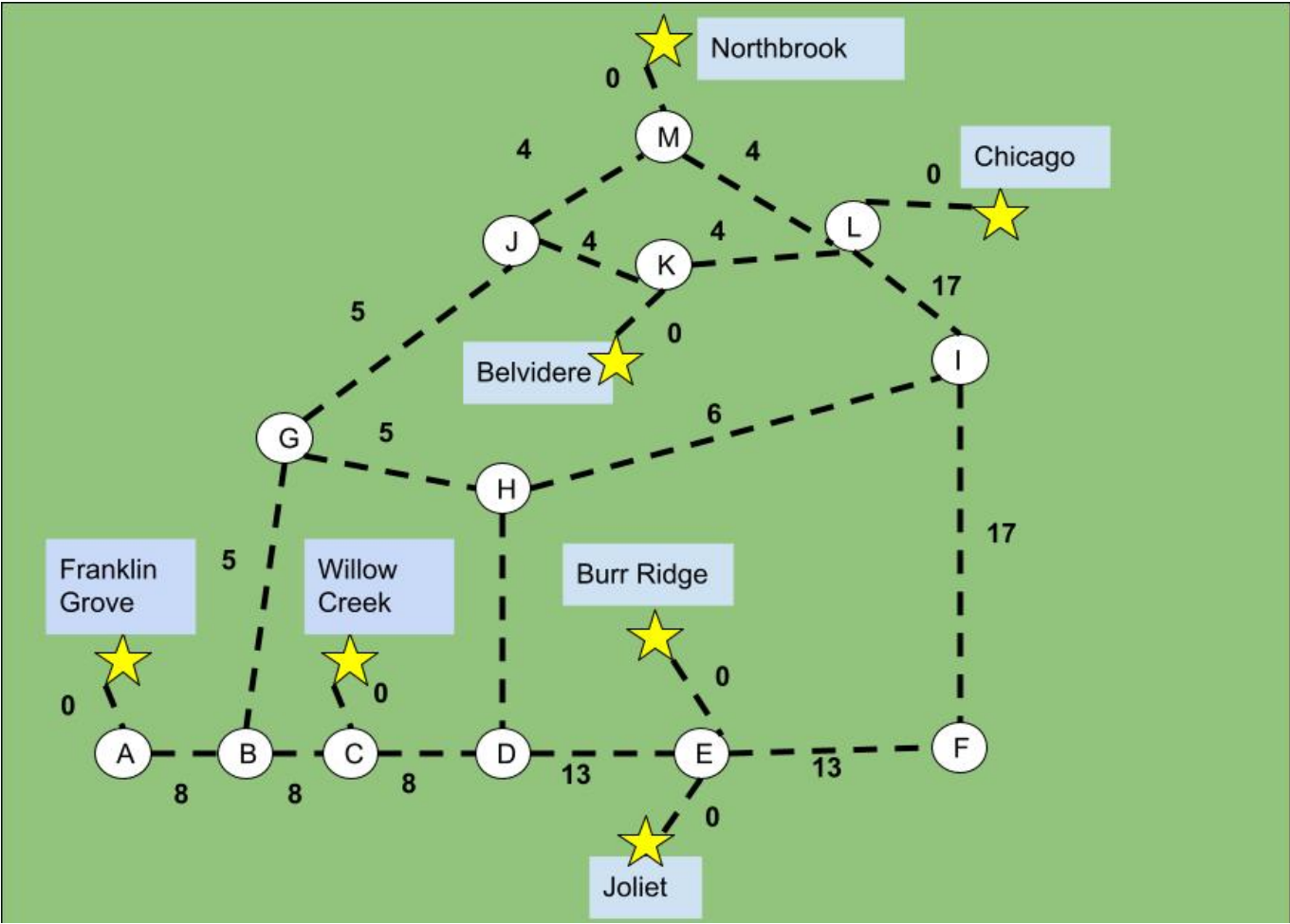
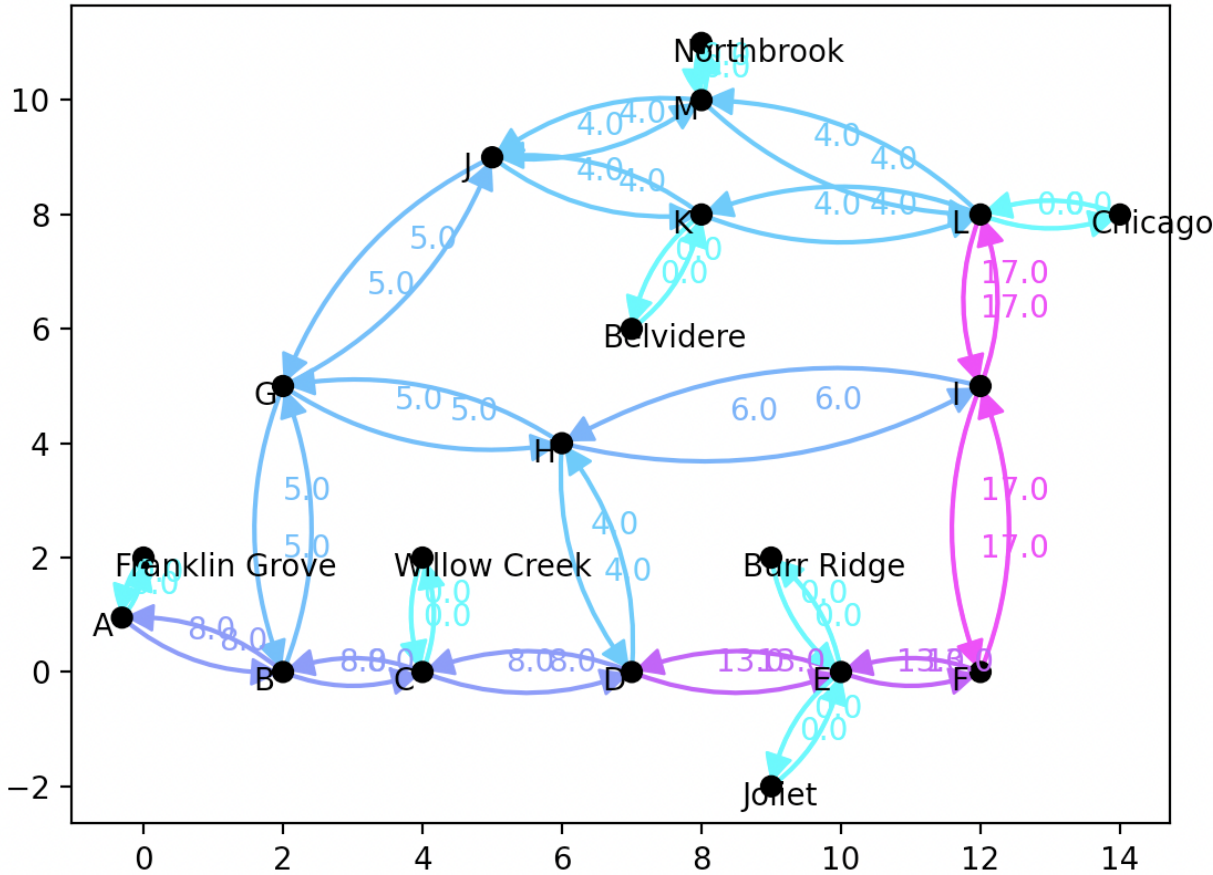
## Examples

Use the map below of the Chicago Tollways,

Letters A-M represent toll booths, and the stars are cities you are visiting.



and below is a modified and simplified version of the map represented as a graph.



Note: maps/graphs may not be accurate to actual tollway maps.

## Examples for a\_star

- **a\_star('Franklin Grove', 'Northbrook')** would return
  - (['Franklin', 'A', 'B', 'G', 'J', 'M', 'Northbrook'], 22)
  - This finds the same optimal path as Dijkstra
- **a\_star('Franklin Grove', 'Joliet')** would return
  - (['Franklin Grove', 'A', 'B', 'G', 'H', 'D', 'E', 'Joliet'], 35)
  - This finds the optimal path (note that BFS would find a sub-optimal path of path length 37)
- **a\_star('Northbrook', 'Belvidere')** would return
  - (['Northbrook', 'M', 'J', 'K', 'Belvidere'], 8)
  - Although two equally-optimal paths exist, A\* chooses the one that's closer to the straight line connecting the Cafeteria to the Med Bay

## Examples for tollways\_algorithim

- For Test 3: **tollways\_algorithm('A', 'C', 1)** would return
  - (3,1) aka (cost, coupon)
  - We only can spend one coupon. It is applied on the path A->C which costs 6. The coupon halves the cost,  $6/2 = 3$ .
- For Test 3: **tollways\_algorithm('A', 'D', 1)** would return
  - (5,1)
  - The path A->B costs 2 then B->D cost 6. We can only use one coupon so applied to the higher cost toll. Therefore  $2 (A \rightarrow B) + 3 (B \rightarrow D \text{ with coupon}) = 5$ .
  - The following is the steps of the data structure updating through Test 3

#1 INITIAL	0 coupons	1 coupons
A	inf	inf
B	inf	inf
C	inf	inf
D	inf	inf
E	inf	inf

We create this matrix format in a dictionary `dp['A'] = [inf,inf,inf]` the dictionary key represents the row and the columns index represent the number of coupons.

We create a priority queue that holds this information as well by pushing in the format `[cost, coupons, node]` e.g. `[inf, 0, A]`.... `[inf,1,E]`

#2 START	0 coupons	1 coupons
A	0	0
B	inf	inf
C	inf	inf
D	inf	inf
E	inf	inf

Initialize the starting vertex in PQ and dictionary to start at 0  
E.g. For start at A... `dp['A'] = [0]*(coupons+1) = [0, 0]`

#3 A TO B	0 coupons	1 coupons
A	0	0
B	2	1
C	inf	inf
D	inf	inf
E	inf	inf

Perform a dijkstra's search and update PQ & dictionary along the way

#4 A TO C	0 coupons	1 coupons
A	0	0
B	2	1
C	6	3
D	inf	inf
E	inf	inf

LOOK AT NEIGHBORS AHEAD

Next step involves If the number of coupons is within the limit of coupons AND the **neighbor** vertex cost indexed at coupon+1 is greater than the cost of the path to  $\text{vertex\_to\_neighbor\_path} // 2 + \text{vertex\_path\_cost}$  update neighbor vertex cost in the dictionary and update the PQ.

#5 PATH to D(based on PQ ordering)	0 coupons	1 coupons
A	0	0
B	2	1
C	6	3
D	inf	7
E	inf	inf

#6 PATH TO D	0 coupons	1 coupons
A	0	0
B	2	1
C	6	3
D	8	UPDATE VALUE! 5 < 7 SO 5
E	inf	9

#7 PATH TO E	0 coupons	1 coupons
A	0	0
B	2	1
C	6	3
D	8	5
E	10	6

#8 PATH TO E	0 coupons	1 coupons
A	0	0
B	2	1
C	6	3
D	8	5
E	9	6

`tollway(A,B, 1)`: look at the minimum cost for row B, use your dictionary.  
The minimum cost is \$1, and is at column 1 which means one coupon is used.  
`tollway(A,C,1)`: look at the minimum cost for row C  
The minimum cost is \$3, and is at column 1 which means one coupon is used.  
`tollway(A,D,1)`: look at the minimum cost for row D  
The minimum cost is \$4, and is at column 1 which means one coupon is used.  
`tollway(A,E,1)`: look at the minimum cost for row E  
The minimum cost is \$6, and is at column 1 which means one coupon

o

- For Test 4: **tollways\_algorithm('A', 'C', 2)** would return
  - We have a direct path from A->C that costs 4. Only need one coupon.
- For Test 4: **tollways\_algorithm('A', 'E', 2)** would return
  - (3,2) aka (cost, coupon)
  - We only can spend two coupons. The path is from A->B (2) and B->E (5).  $1(A \rightarrow B) + (5(B \rightarrow E) // 2)$  (integer division - no remainder) =  $1 + 2 = 3$
- For Test 7: **tollways\_algorithm('Franklin Grove', 'Chicago', 5)** City by Coupon Number Data Table for debugging

	0	1	2	3	4	5
M	22	18	15	12	10	10
Burr Ridge	35	28	24	21	18	16
Chicago	26	22	19	16	14	12
C	16	12	8	8	8	8
E	35	28	24	21	18	16
B	8	4	4	8	8	12
I	24	20	17	14	11	11
H	18	14	11	8	8	12
Willow Creek	16	12	8	8	8	8
L	26	22	19	16	14	12
Franklin Grove	0	0	0	0	0	0
J	18	14	11	8	8	12
G	13	9	6	6	10	10
Northbrook	22	18	15	12	10	10
Belvidere	22	18	15	12	10	10
F	41	32	28	25	22	19
A	0	0	8	8	12	12
K	22	18	15	12	10	10
D	22	18	15	12	10	10
Joliet	35	28	24	21	18	16

## Writing Application Problem

- **Complete the questions below - copy and paste them into a word doc - and include this pdf document as you are submitting your Project10 on Mimir.**
- Use project10 comparison.py file to answer questions
- Matplotlib installation resources: [how-to-install-matplotlib-in-python](#) and [install python packages on PyCharm](#)
- **Writing Questions**

- Question 1 (2pts): Run the first test (test1) from comparison.py file and include the figure produced here.
- Question 2 (2pts): The time complexity of both breadth-first and depth-first graph searches is  $O(|V| + |E|)$ , where  $V$  is the set of vertices of the graph, and  $E$  is the set of edges. In the figure produced above, the "connectivity factor" of a graph is the probability that given any two vertices, say  $v_1$  and  $v_2$ , there exists an edge from  $v_1$  to  $v_2$ . Note that in a fully-connected graph, where every vertex is connected to every other vertex, we have  $E=V^2$ . Why in the figure above does BFS/DFS performance appear linear for small connectivity factors, but not so far larger factors?
- Question 3 (2pts): Run the second test (test2) and include the figure produced here.
- Question 4 (2pts): Notice that AStar search notably outperforms Dijkstras algorithm on the graphs provided. In one or two sentences, informally explain why. What does AStar take advantage of?
- Question 5 (2pts): Suppose you were to call the AStar search as follows: What issues(s) may this cause?

```
graph.a_star(vertex_1, vertex_2, metric = lambda v1, v2 :  
random.rand())
```

## Submission

### Deliverables:

Be sure to upload the following deliverables in a .zip folder to Mimir by 10:00 pm EST, on Thursday, April 21st.

Your .zip folder can contain other files (for example, description.md and tests.py), but must include (at least) the following:

```
| - Project10.zip  
  | - Project10/  
    | - feedback.xml      (for project feedback, make sure to fill this out)  
    | - __init__.py       (for proper testcase loading)  
    | - solution.py       (contains your solution source code)  
    | - project10.pdf     (contains your written solution)
```

## Grading

The following 100-point rubric will be used to determine your grade on Project 10

- Policies
  - **Making all of these policies bold or italic would get too visually fatiguing but read them all because they're important!**



- Using a different algorithm than the one specified for some function will result in the loss of all automated and manual points for that function.
- You will not receive any points on this project if you use Python's built-in sorting functions or sorting functions imported from any library.
- You will not receive any points on the project if you use any list-reversing function such as `reversed`, `list.reverse`, or a homemade alternative to these. You must sort the lists in ascending or descending order directly.
- Tests (70)
  - Test feedback.xml: \_\_/3
  - Test coding standard: \_\_/3
  - **dijkstra**: \_\_/14
    - **dijkstra\_basic**: \_\_/7
    - **dijkstra\_comprehensive**: \_\_/7
  - **a\_star**: \_\_/20
    - **a\_star\_basic**: \_\_/10
    - **a\_star\_comprehensive**: \_\_/10
  - **tollways\_algorithm**: \_\_/20
    - **application\_basic**: \_\_/10
    - **application\_comprehensive**: \_\_/10
- Manual (30)
  - Time and space complexity points are **all-or-nothing** for each function. If you fail to meet time or space complexity in a given function, you do not receive manual points for that function. (30)
  - Manual points for each function require passing all tests for that function, except for the comprehensive tests.
  - Use of non-local in any function will result with loss of manual points for that function.
  - Time & Space Complexity (30)
    - M1 - **dijkstra** : \_\_/10
      - *Time Complexity*:  $O((E + V)\log(V))$
      - *Space Complexity*:  $O(V)$
    - M2 - **a\_star**: \_\_/10
      - *Time Complexity*:  $O((E + V)\log(V))$
      - *Space Complexity*:  $O(V)$
    - M3 - **tollways\_algorithm**: \_\_/10
      - *Time Complexity*:  $O((C*V+E)*\log(V*C))$ , *C is the number of coupons*
      - *Space Complexity*:  $O(V*C)$ , *C is the number of coupons*
    - M4 - **WRITING ASSIGNMENT**: \_\_/10

Congratulations on Finishing your Last Project!

- [What time is it?!](#)