

Sources:

Problem 2

1. <https://bradfieldcs.com/algos/graphs/dijkstras-algorithm/>

Problem 3

2. <https://stackoverflow.com/questions/13159337/why-doesnt-dijkstras-algorithm-work-for-negative-weight-edges>

1. Graph Traversal

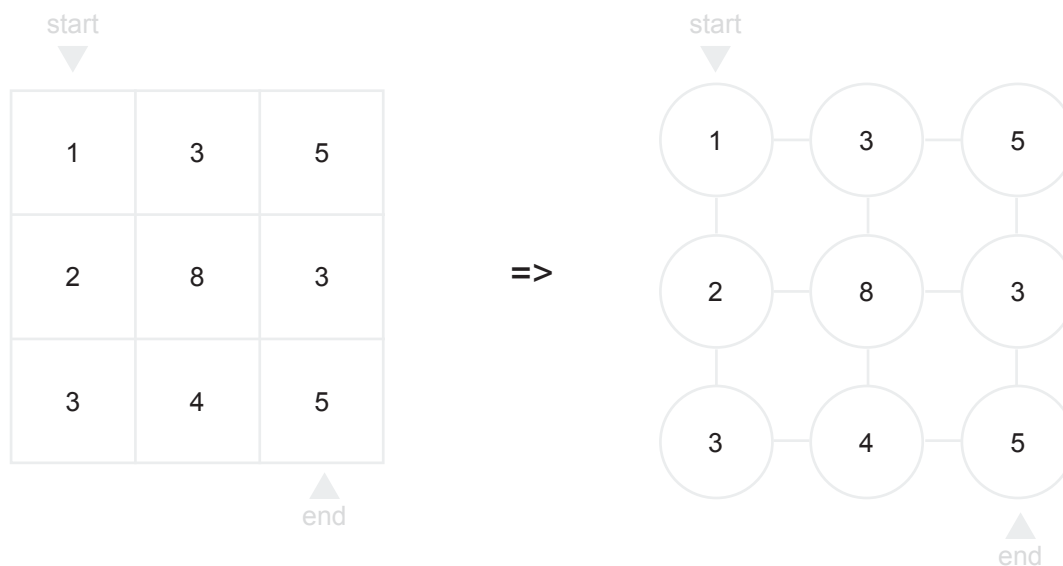
BFS: A, B, D, C, F, G, E

DFS: A, B, F, C, D, E, G

2. Apply BFS/DFS to solve a problem

a. description

Using the example provided in the assignment, we can translate the 2d puzzle[m][n] into an undirected graph:



Evaluating at the puzzle as an undirected graph, it looks a lot like the Dijkstra algorithm. The difference is that instead of looking at the cumulative weights of a path, we want to only consider the minimum effort of a path.

The relation of the Dijkstra of the algorithm is (where we are considering $x \rightarrow y$) :

$$\text{dist}[y] = \min\{ \text{dist}[y] , \text{weight}[x,y] + \text{dist}[x] \}$$

Our modified version to consider the minimum effort of the path:

$$\text{effort}[y] = \min\{ \text{effort}[y] , \text{effort}[x] \}$$

Since the *minEffort(puzzle)* algorithm is very similar to the Dijkstra algorithm, I adapted the *calculate_distances(graph, starting_vertex)* provided in the lecture. This algorithm uses a BFS approach, so a priority queue implemented as a minimum heap is used.

Since *puzzle[][]* is a 2d array, a nested for-loop is used to convert the puzzle to a graph. For each vertex in the graph, it's neighbours are computed and stored as cell indices with their corresponding efforts.

b. pseudocode

```

minEffort(puzzle[m][n]):

    # convert 2d array puzzle[m][n] into graph for calculate_min_effort()
    starting_vertex = "r" + str(0) + "-" + "c" + str(0)
    end_vertex = "r" + str(m-1) + "-" + "c" + str(n-1)
    puzzle_graph = {}

    for m in range(len(puzzle)):
        for n in range(len(puzzle[0])):
            current_cell = "r" + str(m) + "-" + "c" + str(n)
            puzzle_graph[current_cell] = {}
            # if not top row
            if m != 0:
                neighbor_cell = "r" + str(m-1) + "-" + "c" + str(n)
                puzzle_graph[current_cell][neighbor_cell] = abs(puzzle[m][n] - puzzle[m-1][n])
            # if not left column
            if n != 0:
                neighbor_cell = "r" + str(m) + "-" + "c" + str(n-1)
                puzzle_graph[current_cell][neighbor_cell] = abs(puzzle[m][n] - puzzle[m][n-1])
            # if not bottom row
            if m != (len(puzzle)-1):
                neighbor_cell = "r" + str(m+1) + "-" + "c" + str(n)
                puzzle_graph[current_cell][neighbor_cell] = abs(puzzle[m][n] - puzzle[m+1][n])
            # if not rightmost column
            if n != (len(puzzle[0])-1):
                neighbor_cell = "r" + str(m) + "-" + "c" + str(n+1)
                puzzle_graph[current_cell][neighbor_cell] = abs(puzzle[m][n] - puzzle[m][n+1])

    return calculate_min_effort(puzzle_graph, starting_vertex, end_vertex)

```

```

calculate_min_effort (graph{}, starting_vertex, end_vertex) :

    efforts = {vertex: float('infinity') for vertex in graph}
    efforts[starting_vertex] = 0

    # set up priority queue with starting vertex
    pq = [(0, starting_vertex)]
    while len(pq) > 0:
        current_effort, current_vertex = pop (effort, vertex) from heap pq

        if current_effort > efforts[current_vertex]:
            continue

        for neighbor, weight in graph[current_vertex].items():
            effort = weight

            if effort < efforts[neighbor]:
                efforts[neighbor] = effort
                push (effort, neighbor) onto heap pq

    return efforts[end_vertex]

```

c. time complexity

Let V equal the number of vertices in the `puzzle[m][n]` which is $m \cdot n$

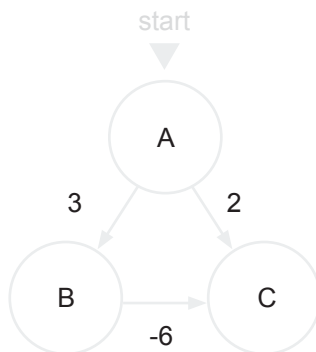
Let E equal the number of edges in the `puzzle[m][n]` which is $2 \cdot (m \cdot n)$, since the puzzle is an undirected graph

For `calculate_min_effort()` the while loop is executed E times for each edge and the inner for loop will be executed V times for each vertex. Push and pop from pq take $\log V$ time complexity. So `calculate_min_effort()` will have $O((V+E)\log V)$ time complexity.

For `minEffort()`, we need to convert the 2d array `puzzle[m][n]` into a graph for `calculate_min_effort()`. This takes $O(V+E)$, since for each vertex we need to calculate the weight of each neighbour and add it to the graph, which there are E in total.

Since $O((V+E)\log V) > O(V+E)$, the time complexity will be $O((V+E)\log V)$

3. Analyze Dijkstra with Negative Edges



Vertex	Cost
A	0
B	∞
C	∞

Step 1.

We will first compare the length of the edges (A, B, 3) and (A, C, 2) and update our path length in the table.

Vertex	Path length
A	0
B	3
C	2

Step 2.

We have reached the end vertex C, with a cost of 2. However, if we had explored B, we would have found that the total path length would be $3 + -6 = -3$. $-3 < 2$, so the solution provided with the Dijkstra algorithm is incorrect.

Vertex	Path length
A	0
B	-3
C	2

We have an incorrect solution because the algorithm only 'visits' a vertex once, and at a vertex we follow the relation: (where we are considering $x \rightarrow y$) : $\text{dist}[y] = \min\{ \text{dist}[y] , \text{weight}[x,y] + \text{dist}[x] \}$

We assume that this relation provides us with an optimal solution for a vertex and then move on to the next vertex. However, the inductive step does not hold if there is a negative weight, because the Dijkstra algorithm is greedy by design. At each step we assume that by choosing a locally optimum value we will reach a globally optimum solution. If the next step holds a better locally optimum value than the previous step because there is a negative weight, then the problem does not have an optimal substructure, and thus won't provide a globally optimum solution.

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HW6 : Graph Algorithms

Extra credit

BFS: *A, B, C, D, E, G, F, I, H, J*

DFS: *A, B, C, D, E, G, F, H, I, J*