## Midterm Exam

## 1 Generating random walks and random spanning trees on a graph

Download the file rw.py and the file  $test_rw.py$ . The file rw.py contains a class Graph (with methods to complete). An object g of type Graph has an attribute g.n that gives the number of vertices (the n vertices have distinct labels in [0..n-1]), and an attribute g.L that is a list of lists, such that g.L[i] gives the list of neighbours of vertex i (in whatever order), see Figure 1 for an example.

$$g = 0$$

$$g.n = 4$$

$$g.adj = [[1, 2, 3], [0, 2], [0, 3, 1], [0, 2]]$$

FIGURE 1 – Left: a graph g. Right: the encoding of g as an instance of the class **Graph**. For instance the neighbours of vertex 1 are 0 and 2, so that g.L[1] = [0, 2].

Question 1 (1 pts). Complete the method random\_neighbour(self,i) that has to return one of the neighbours of vertex i at random (i.e., if i has d neighbours, then each neighbour has probability 1/d of being returned). To test your method, uncomment the line containing test\_random\_neighbour() at the end of test\_rw.py, and then run that file.

For  $k \ge 0$ , a random walk of length k on a graph g (starting from vertex 0) is a list  $[v_0, v_1, \ldots, v_k]$  of vertices such that  $v_0 = 0$ , and for each  $i \in [1..k]$ ,  $v_i$  is a random neighbour of  $v_{i-1}$ .

Question 2 (2 pts). Complete the method  $random_walk(self,k)$  that has to return a random walk of length k starting from 0 (the returned list should have length k+1 and start with 0). To test your method, uncomment the line containing  $test_random_walk()$  at the end of  $test_rw.py$ , and then run that file.

When drawing a random walk  $v_0, v_1, v_2, ...$ , the *cover time* is the first time where we have visited (at least once) each of the n vertices.

Question 3 (3 pts). Complete the method random\_walk\_till\_covered(self) that draws a random walk starting at 0, stops at the cover time, and returns the walk seen so far (the returned list should start with 0, contain every entry in [0..n-1] at least once, and the last entry of the list should occur just once in the list). To test your method, uncomment the line containing test\_cover\_walk() at the end of test\_rw.py, and then run that file.

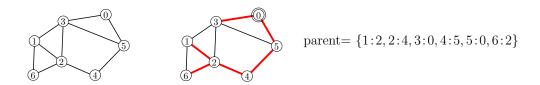


FIGURE 2 – Left: a graph g. Right: a spanning tree T of g (whose edges are bold red), and the encoding of T as a dictionary parent. For instance the path from 2 to 0 in T visits (after leaving 2) successively 4, 5 and 0, hence the parent of 2 is 4, so that parent [2]=4.

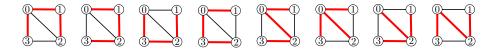


FIGURE 3 – The 8 spanning trees of the graph of Figure 1.

For g a graph, a spanning tree T of g is a set of edges of g that forms a tree and covers all vertices of g. Precisely, there is no cycle of edges from T, and for every vertex  $i \in [1..n-1]$  there is a (necessarily unique) path of edges of T from i to 0. The next vertex after i along this path is called the parent of i. We will store a spanning tree T as a dictionary parent where the set of keys is [1..n], and for each  $i \in [1..n]$ , parent [i] gives the label of the parent of i, see Figure 2 for an example.

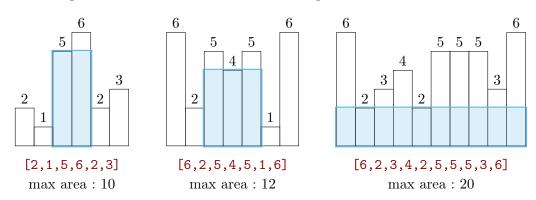
We let  $\mathcal{ST}(g)$  be the set of spanning trees of g (see Figure 3 that shows the set  $\mathcal{ST}(g)$  of 8 spanning trees of the graph g of Figure 1). We would like to have a method that returns a random element in  $\mathcal{ST}(g)$  (for instance for the graph of Figure 1, each of its 8 spanning trees shown in Figure 3 should have probability 1/8 of being chosen). As it turns out, one can use a random walk till cover time to do that. Precisely, if W is a random walk of g (starting at 0) till cover time, for each  $i \in [1..n-1]$  we let p(i) be the label of the vertex preceding the first occurrence of i in W. Then it can be shown (admitted here) that the tree formed by the n-1 edges  $\{i,p(i)\}$  is a random spanning tree of g (each element of  $\mathcal{ST}(g)$  has the same probability of being chosen). It is easy to see that p(i) is the parent of i in this tree (indeed the path from i to 0 visits p(i), then p(p(i)), etc.).

Question 4 (2 pts). Complete the method random\_spanning\_tree(self) that has to return a dictionary parent associated to a random spanning tree of self (your code should not call the method random\_walk\_till\_covered, but it should have a quite similar structure). To test your method, uncomment the line containing test\_random\_tree() at the end of test\_rw.py, and then run that file.

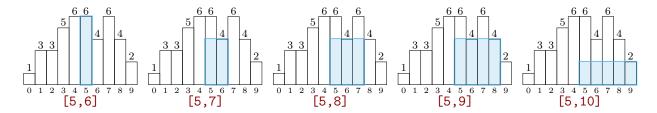
## 2 Largest Rectangle in Histogram

Guidelines. Download rect\_hist.py and rect\_hist\_test.py. Write your solution in rect\_hist.py. To test your code, run the file rect\_hist\_test.py. You don't have to justify the complexity of your programs.

We are given a histogram, a sequence of positive values, and we want to find the area of the largest rectangle in the histogram. In this exercise, n will always refer to the length of the histogram. The largest such rectangle is illustrated below on several examples:



At first, we consider the following simpler problem: given a position i, find the area of the largest rectangle starting at position i. The example below illustrates all possible such rectangles on a histogram with i = 5.



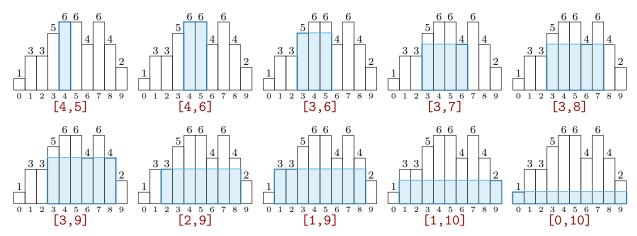
Question 5 (2 pts). Write an algorithm rect\_from\_left(hist, i) that given an array hist representing the histogram and a position i, returns the area of the largest rectangle in the histogram that starts at position i. Make sure your algorithm has complexity O(n). Test your implementation using rect\_hist\_test.py.

Question 6 (1 pts). Write a "brute force" algorithm rect\_hist\_brute(hist) that given an array hist representing the histogram returns the area of the largest rectangle in the histogram, using rect\_from\_left. Make sure your algorithm has complexity  $O(n^2)$ . Test your implementation using rect\_hist\_test.py.

Now we would like to design a divide and conquer algorithm to solve this problem more efficiently. Consider a histogram H and a sub-histogram H[i:j] with i < j, and let  $m = \lfloor \frac{i+j}{2} \rfloor$ . Then there are three possibilities for a rectangle spanning H[l:r] of largest area:

- it is entirely within H[i:m],
- it is entirely within H[m+1:j],
- it is within H[i:j] and contains m, that is  $l \leq m < r$ .

We now focus on the latter case: starting from a rectangle containing only A[m], we repeatedly expand this rectangle to the left or the right until it fills all of A[i:j]. We then take the maximum area of all rectangles is this sequence. This process is illustrated on an example below where H=[1,3,3,5,6,6,4,6,4,2] and m=4. We have represented the different steps of the expansion process and the rectangle spanning H[1:r].



Formally, given a sub-histogram H[i:j] and a rectangle spanning H[l:r] (with  $[l:r] \subset [i:j]$ ), an *expansion step* consists in expanding the rectangle to the left or the right (we only expand in one direction):

- if l = i then expand to the right (r' = r + 1),
- if r = j then expand to the left (l' = l 1),
- if H[l-1] > H[r] then expand to the left,
- otherwise expand to the right.

Question 7 (3 pts). Write a function expand\_rect(hist, i, j, l, r, h) that implements one expansion step and returns a tuple 1',r',h'. In the input, h is assumed to be the minimum height in hist[1:r]. In the output, 1' and r' are such that [1',r'] is the updated range, and h' has to be the minimum height in hist[1':r'] Your algorithm should have complexity O(1). Test your implementation using rect\_hist\_test.py.

Question 8 (3 pts). Write a function best\_from\_middle(hist, i, j, m) that implements the expansion process by repeatedly calling expand\_rect starting from l=m and r=m+1, and until l==i and r==j. It must return the area of the largest rectangle in the sequence. Your algorithm should have complexity O(j-i). Test your implementation using rect\_hist\_test.py.

Question 9 (3 pts). Write a recursive function rect\_hist\_dac\_aux(hist, i, j) that implements the divide-and-conquer approach outlined at the beginning and uses best\_from\_middle. It must return the area of the largest rectangle in the sequence, or -math.inf when i==j. Your algorithm should have complexity  $O(n \log(n))$ . Test your implementation using rect\_hist\_test.py.