Homework 1 - Algorithm Design

Sapienza University of Rome

Marco Costa, 1691388

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Problem

Solution

Let's define a matrix M $(k \times n)$, in which we have the expected value opening the i^{th} box having j as current best reward $(i \in [0, k-1], j \in [0, n])$

Algorithm 1 Populate the matrix

1: **for**
$$i$$
 in $(0, ..., k-1)$ **do**

2: **for**
$$j$$
 in $(0, ..., n)$ **do**

3:
$$M[i,j] \leftarrow -\sum_{h=1}^{i} c_h + \frac{j}{n+1} (j-1) + \frac{1}{n+1} \sum_{h=j}^{n} h$$

Now, for each box we can calculate the "a priori" expected value as follow:

$$ar{\mathbf{v_k}} = \sum\limits_{i=0}^{n} \mathbf{P_{k-1}}(i) \mathbf{v_k}(i)$$

where $\mathbf{P_{k-1}}(\mathbf{i})$ is the probability of having reward i at least in one of the k-1 boxes, while $\mathbf{v_k}(\mathbf{i})$ is the expected value opening the k^{th} box having as current best reward i.

$$\mathbf{P}_{k-1}(i) = (\tfrac{1}{n+1})(\tfrac{i+1}{n+1})^{k-2}(k-1), \qquad \mathbf{v}_k(i) = \mathbf{M}[k,i]$$

So we have:

$$\bar{\mathbf{v_k}} = \sum_{i=0}^n (\tfrac{1}{n+1}) (\tfrac{i+1}{n+1})^{k-2} (k-1) \mathbf{M}[k,i] = \tfrac{k-1}{n+1} \sum_{i=0}^n (\tfrac{i+1}{n+1})^{k-2} \mathbf{M}[k,i]$$

Now we have to calculate $\bar{\mathbf{v}_k}$ for each box and return the max:

Algorithm 2 Find expected optimal reward

- 1: $k_{max} \leftarrow 0$
- $2: v[k] \leftarrow \emptyset$
- 3: **for** i in (0,...,k-1) **do**
- for j in (0, ..., n) do $v[i] += \frac{k-1}{n+1} (\frac{i+1}{n+1})^{k-2} M[i, j]$ 5:
- if $v[i] > v[k_{max}]$ then 6:
- $k_{max} \leftarrow i$
- 8: return $v[k_{max}]$

First problem

Given a weighted tree T with n nodes, find the complete graph G of minimum weight such that $T \subseteq G$ and T is the unique minimum spanning tree of G.

Solution

Idea: Insert edges that are not in the tree so as to obtain the complete graph. These edges must have a greater weight than those of the tree, so that T is the only MST of G.

Algorithm 3 Find complete graph

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1: for u in V do
2: for v in V do
3: e \leftarrow (u, v)
4: if e not in E then
5: weight \leftarrow max(u.getHeaviestEdge().getWeight(), v.getHeaviestEdge().getWeight())
6: e.setWeight(weight + 1)
7: G.addEdge(e)
```

Cost: $\mathbf{O}(|E|)$, but $|E| = \frac{|V|(|V|-1)}{2}$ and |V| = n, so $|E| = \frac{n^2 - n}{2}$ and the cost is $\mathbf{O}(n^2)$

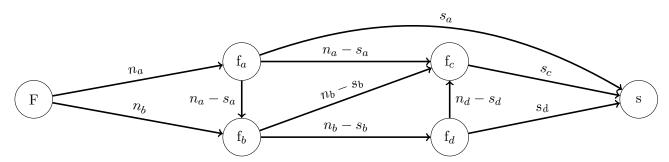
Second problem

Solution

Hint: Since *Kruskal*'s algorithm uses the *Union-Find* structures for representing the cuts, we use this structures to solve the problem.

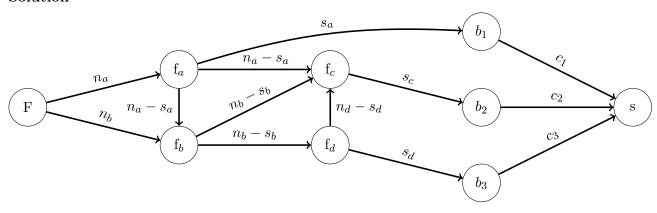
First Problem

Solution



Second Problem

Solution



Problem

Solution