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Algorithm & Data Structures

Solutions to Sheet P8

AS 16

Solution P8.1 *Wind Turbines.*

We will use dynamic programming for this task: for every $i \in \{0, \dots, n\}$, let m_i be the best solution only using the positions $\{d_1, \dots, d_i\}$, so $m_0 = 0$ and we are interested in m_n .

Now there are two ways the best solution on positions $\{d_1, \dots, d_i\}$ may look: either it uses the turbine at d_i or not. If it does use the turbine at d_i , then the value is e_i plus the best solution from turbines before d_i far enough from d_i , which is captured by some m_z . If it does not use the turbine at d_i , then we have just $m_i = m_{i-1}$. Taken together, we have

$$m_i = \max\{e_i + m_z, m_{i-1}\},$$

where z is maximal such that $d_z \leq d_i - D$ (or 0 if already $d_1 > d_i - D$). Now we can compute the values m_i from m_1 to m_n and we just have to find the right value of z in every step.

One way to find z is to look at all values $0, \dots, i - 1$ in every step, comparing d_z to $d_i - D$, but that would give us an $\mathcal{O}(n^2)$ time algorithm, too slow for the tests. We can improve this with a binary search on $d_1, \dots, i - 1$ to find the last value before $d_i - D$ (or 0 if there isn't such), and get an $\mathcal{O}(n \log n)$ time solution.

However, the fastest way to find z happens to be also the simplest one: Observe that the value of z for step i has to be at least the value of z for step $i - 1$. So we can remember the old value of z we used in the step computing m_{i-1} and in the step computing m_i we increase this z by 1 in a `while` loop until we hit $d_{z+1} > d_i - D$.

This gives an $\mathcal{O}(n)$ time solution to the problem: It may happen that we need to increase z many times (checking $d_{z+1} > d_i - D$ every time) in a single step from d_{i-1} to d_i but since z starts at 0 and never goes above $n - 1$, we can only do this at most n times during the whole algorithm, so the runtime consumed by this `while` loop is $\mathcal{O}(n)$ during the whole program.

Solution programs

On the lecture website, you can find a solution running in time $\mathcal{O}(n)$ that uses the last approach. The solution source contains further comments on the implementation.

There is also a second solution illustrating a different dynamic programming approach: let t_i be the best solution from turbines at d_1, \dots, d_i that *does* use d_i . When computing t_i , we let $t_i = e_i + \max\{0, t_1, t_2, \dots, t_z\}$ where z is as above. The presented solution looks at all such t_j every step, obtaining an $\mathcal{O}(n^2)$ time solution. However, even in this case this could be easily sped up to $\mathcal{O}(n)$ by remembering not only the last z , but also the previous maximum of $\{0, t_1, t_2, \dots, t_z\}$ and updating it along with increasing z . We leave the details to the reader.

Data

Every judge test also contained 1-3 small additional special cases, e.g. with only 1 position, with all positions too close (so you can only build one), and with all the positions far enough (so you can build all of them).

judge1 $n = 100$, $D = 1\,000\,000$, all positions in conflict.

judge2 $n = 100$, $D = 1$, no positions in conflict.

judge3 $n = 100$, $D = 200\,000$, every position in conflict with ca $2/5$ of others.

judge4 $n = 1\,000$, $D = 50\,000$, random positions from $0 \dots 1\,000\,000$.

judge5 $n = 100\,000$, $D = 5\,000$, random positions from $0 \dots 1\,000\,000$. A quadratic program should fail this.