

(a) To do this, we build a graph H as follows. H has the same nodes as all the G_i , and it consists precisely of those edges that occur in every one of G_0, \dots, G_b . In this graph H , we simply perform breadth-first search to find the shortest path from s to t (if any s - t path exists).

Note that this idea, generalized to any sequence of graphs G_i, \dots, G_j , will be useful in part (b).

(b) We are given graphs G_0, \dots, G_b . While trying to find the last path P_b , we have several choices. If G_b contains P_{b-1} , then we may use P_{b-1} , adding $l(P_{b-1})$ to the cost function (but not adding the cost of change K .) Another option is to use the shortest s - t path, call it S_b , in G_b . This adds $l(S_b)$ and the cost of change K to the cost function. However, we may want to make sure that in G_{b-1} we use a path that is also available in G_b so we can avoid the change penalty K . This effect of G_b on the earlier part of the solution is hard to anticipate in a greedy-type algorithm, so we'll use dynamic programming.

We will use subproblems $Opt(i)$ to denote minimum cost of the solution for graphs G_0, \dots, G_i .

To compute $Opt(n)$ it seems most useful to think about where the last changeover occurs. Say the last changeover is between graphs G_i and G_{i+1} . This means that we use the path P in graphs G_{i+1}, \dots, G_b , hence the edges of P must be in every one of these graphs.

Let $G(i, j)$ for any $0 \leq i \leq j \leq b$ denote the graph consisting of the edges that are common in G_i, \dots, G_j ; and let $\ell(i, j)$ be the length of the shortest path from s to t in this graph (where $\ell(i, j) = \infty$ if no such path exists).

If the last change occurs between graphs G_i and G_{i+1} then we get that $Opt(b) = Opt(i) + (b - i)\ell(i + 1, b) + K$. We have to deal separately with the special case when there are no changes at all. In that case $Opt(b) = (b + 1)\ell(0, b)$.

So we get argued that $Opt(b)$ can be expressed via the following recurrence:

$$Opt(b) = \min[(b + 1)\ell(0, b), \min_{1 \leq i < b} Opt(i) + (b - i)\ell(i + 1, b) + K].$$

Our algorithm will first compute all $G(i, j)$ graphs and $\ell(i, j)$ values for all $1 \leq i \leq j \leq b$. There are $O(b^2)$ such pairs and to compute one such subgraph can take $O(n^2b)$ time, as there are up to $O(n^2)$ edges to consider in each of at most b graphs. We can compute the shortest path in each graph in linear time via BFS. This is a total of $O(n^2b^3)$ time, polynomial but really slow. We can speed things up a bit to $O(b^2n^2)$ by computing the graphs $G(i, j)$ and $\ell(i, j)$ for a fixed value of i in order of $j = i \dots b$.

Once we have precomputed these values the algorithm to compute the optimal values is simple and takes only $O(b^2)$ time. We will use $M[0 \dots b]$ to store the optimal values.

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For i=0,...,b
  M[i] = min((i + 1)\ell(0, i); min_{1 \leq j < i} M[j] + (i - j)\ell(j + 1, i))
EndFor

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¹ex377.520.504