Canadians should use memory

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Abstract

We establish that the competitive ratio of any randomized memoryless strategy is not less than 2k + 1. Only randomized strategies using memory potentially overpass this ratio now.

1 Definitions

We start by introducing the notation. For any graph $G = (V, E, \omega)$, let $G \setminus E'$ denotes its subgraph $(V, E \setminus E', \omega)$. If P is a path, we note its cost as $\omega(P) = \sum_{e \in P} \omega(e)$.

1.1 Memoryless Strategies for the k-CTP

We remind the definition of CTP. Let $G = (V, E, \omega)$ be an undirected graph with positive weights. The objective is to make a traveller traverse the graph from a source node s to a target one t, with $s, t \in V$. There is a set $E_* \subsetneq E$ of blocked edges. The traveller does not know a priori which edges are blocked. He discovers a blocked edge only when arriving to one of its endpoints. For example, if (v, w) is a blockage he will discover it when arriving to v (or v). The goal is to design the strategy v0 with the minimal competitive ratio.

We focus on memoryless strategies (MS). Concretely, we suppose that the traveller remembers the blocked edges he has discovered but forgets the nodes which he has already visited. In other words, a decision of an MS is independent of the nodes already visited. In the literature, the term memoryless was used in the context of online algorithms (e.g. PAGING PROBLEM [?], LIST UPDATE PROBLEM [?]) which take decisions according to the current state, ignoring past events. An MS can be either deterministic or randomized.

Definition 1 (Memoryless Strategies for the k-**CTP)** A deterministic strategy A is an MS if and only if (iff) the next node w the traveller visits depends on graph G deprived of blocked edges already discovered E'_* and the current traveller position $v: w = A(G \setminus E'_*, v)$. Similarly, a randomized strategy A is an MS iff node w is the realization of a discrete random variable $X = A(G \setminus E'_*, v)$.

MSes are easy to be implemented because they do not use past moves to take a decision. For the same reason, they do not need any extra memory either.

For example, the GREEDY strategy [?] is an MS. It consists in choosing at each step the first edge of the shortest path between the current node v and the target t. This strategy, illustrated

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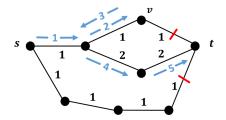


Figure 1: Illustration of the GREEDY strategy on a graph with 2 blocked edges

in Figure 1, does not refer to anterior moves. In contrast, the REPOSITION strategy [?] is not an MS as any decision refers to the past moves of the traveller.

We propose the following process to identify whether a strategy A is a deterministic MS. Let us suppose that a traveller T_1 executes strategy A: he has already visited certain nodes of the graph, he is currently at node v but he has not reached target t yet. Let us imagine a second traveller T_2 who is airdropped on node v and starts applying strategy A. If the traveller T_2 always follows the same path as T_1 until reaching t, A is a deterministic MS. If T_1 and T_2 may follow different paths, then A is not an MS. Formally, prooving that a strategy is an MS consists in finding the function which transforms the pair $(G \setminus E_*, v)$ into node $w = A(G \setminus E'_*, v)$.

1.2 Competitive ratio

Let (G, E_*) be a road map, i.e. a pair with graph $G = (V, E, \omega)$ and blocked edges $E_* \subseteq E$, such that there is an (s,t)-path in graph $G \setminus E_*$ (nodes s and t remain in the same connected component when all blocked edges are discovered). We note $\omega_A(G, E_*)$ the distance traversed by the traveller reaching t with strategy A on graph G with blocked edges E_* and $\omega_{\min}(G, E_*)$ the cost of the shortest (s,t)-path in graph $G \setminus E_*$.

The ratio $\omega_A(G, E_*)/\omega_{\min}(G, E_*)$ is abbreviated as $c_A(G, E_*)$. A strategy A is c_A -competitive [?, ?] iff for any (G, E_*) , $\omega_A(G, E_*) \leq c_A\omega_{\min}(G, E_*)$. Otherwise stated, for any (G, E_*) , $c_A(G, E_*) \leq c_A$. If strategy A is randomized, $\omega_A(G, E_*)$ is replaced by $\mathbb{E}(\omega_A(G, E_*))$ which is the expected distance traversed by the traveller to reach t with strategy A. The competitive ratio can also be evaluated on a family \mathcal{R} of road maps, put formally,

$$c_{A,\mathcal{R}} = \max_{(G,E_*)\in\mathcal{R}} c_A(G,E_*) \tag{1}$$

This "local" competitive ratio fulfils $c_{A,\mathcal{R}} \leq c_A$. The competitive ratio can also be extended to families of strategies. We note c_{MS} the competitive ratio of MSes, which is the minimum over competitive ratios of any MSes: $c_{\text{MS}} = \min_{A \text{ MS}} c_A$. In the remainder, we identify a set of road maps \mathcal{R} such that the competitive ratio of MSes on it is 2k + 1, *i.e.* $c_{\text{MS},\mathcal{R}} \geq 2k + 1$.

1.3 Graphs G_k

We define recursively a sequence of graphs G_i for $i \geq 1$ with weights from $\{1, \varepsilon\}$, $0 < \varepsilon \ll 1$. Graphs G_1 and G_{i+1} are represented in Figures 2a and 2b, respectively. Edges with weight 1 are thicker than edges with weight ε . For every graph G_k , we will only consider the road maps where the blocked edges are at the right of the graph, the left edges affecting negligibly the total cost of a traveler. We note E_{right} the set of all the right edges.

When the traveller discover blocked edges in G_k , parts of the graph become dead ends. If a traveller visits a dead end, the only chance for him to reach t is to leave it anyway. Dead ends

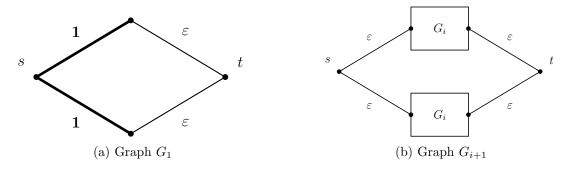


Figure 2: Recursive construction of graphs G_i

only provide to the traveller additional costs which make the competitive ratio increase. As a consequence, the most competitive strategies will not visit dead ends. From now on, as soon as a dead end appear when the traveller discovers a blockage, we make it disappear from the graph: graph $G \setminus E'$ becomes graph G deprived of both edges E' and dead ends appeared when edges E' are deleted. Let \mathcal{G}_k denote the set of all sub-graphs of G_k where at most k edges have been removed. Formally, $\mathcal{G}_k = \{G_k \setminus E' : |E'| \leq k\}$.

1.4 Equivalent binary trees

We define for each graph of \mathcal{G}_k an equivalent binary tree, which will represent all changes of direction from t. We note T_{\varnothing} the empty tree, and (v, T_a, T_b) a non empty tree with $v \in V$, T_a the binary tree above and T_b the binary tree below.

We obtain the equivalent binary tree by applying the BIN-TREE algorithm. This algorithm take as an entry a directed tree obtained by first removing the left part of the graph and then directed all edges from t. Then, the BIN-TREE algorithm will construct a binary tree according to the outdegree of each node.

Algorithm 1: BIN-TREE algorithm

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Data: a directed tree T and a node v

Result: binary tree

if deg^-(v) = 0 then return T_{\varnothing};

if deg^-(v) = 1 then return BIN-TREE(T, v_{\text{next}});

if deg^-(v) = 2 then return (v, \text{BIN-TREE}(T, v_{\text{up}}), \text{BIN-TREE}(T, v_{\text{down}});
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Given an edge in a binary tree, we will note A(e) the child edge from above and B(e) the child edge from below. We also note C(e) the set of all the descendants of e. Formally, $C(e) = C(A(e)) \cup C(B(e)) \cup \{A(e), B(e)\}$. Finally, we note P(e) the parent edge of $e: P(e) = e' \Leftrightarrow e = A(e')$ or e = B(e').

We define recursively a set of edges E_k , E_0 being the edges connected to the root and $E_k = \{e : P(e) \in E_{k-1}\}$. E_k represents the edges which are at distance k from the root.

When an edge is cut from the original graph, the equivalent tree can be easily obtain by replacing the tree (v, T_a, T_b) by T_a (resp. T_b) if the cut edge is before T_b (resp. T_a)

Lemma 1 Given $G_k \setminus E' \in \mathcal{G}_k$, $(k - |E'|) \leq \min_{v \in V} h(v)$