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Online Graph Exploration with Advice

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advice

optimality

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algorithm

Graph exploration

Gedeutstatt sur sechshundert jährigen Dubelbier der Käniglichen Boupt und Nesiden; Stadt Sinigsberg in Preufen.



advice

optimality

algorithm

Graph exploration

Gedentiftatt sur sechshundert jährigen Dubelfeier der Käniglichen Kompt und Nesiden; Stadt Sinigsberg in Preufen.



general statement

Given a ... graph, find a ... closed walk that visits all

advice

optimality

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algorithm

Graph exploration

Gedenbälatt zur sechshundert jährigen Bibelbier der Königlichen Kompt und Nesiten; Stadt Sonigsberg in Preufen.



our case

Given a weighted undirected graph, find a shortest closed walk that visits all vertices.

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problem stater	nent			
Given a weight	ted undirected gra	aph,		
find a shortest	closed walk that	visits all vert	ices.	J



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	problem statement				
	Given a weighted ı	undirected graph,	,		
	find a shortest close	ed walk that visi	ts all vertices		



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	problem statemen	t			
	Given a weighted find a shortest clo	undirected grapl sed walk that vi	n, sits all vert	ices.	



• equivalent to TSP in the metric closure (vertices may repeat)

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problem statem Given a weighte find a shortest o	ent ed undirected gra closed walk that	aph, visits all vert	ices.	



- equivalent to TSP in the metric closure (vertices may repeat)
- 2.MST is 2-approximation

online graph exploration	lower bound	advice	optimality	algorithm



online graph exploration	lower bound	advice	optimality	algorithm



online graph exploration	lower bound	advice	optimality	algorithm



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lower bound

advice

optimality

algorithm





agent in vertex *v* sees

unique ID

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online graph exploration	lower bound	advice	optimality	algorithm







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online graph exploration low	ver bound	advice	optimality	algorithm





agent in vertex v sees

- unique ID
- weights
- neighbors' IDs

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online graph exploration	lower bound	advice	optimality	algorithm



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online graph exploration	lower bound	advice	optimality	algorithm



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online graph exploration	lower bound	advice	optimality	algorithm



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main question				
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What is the (worst case) **length of** the agent's **traversal compared to** (offline) **optimum?**

online grant

Is there a constant competitive algorithm?

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online graph exploration	lower bound	advice	optimality	algorithm

main question

What is the (worst case) **length of** the agent's **traversal compared to** (offline) **optimum?**

Is there a constant competitive algorithm?

what has been known about competitive ratio

- [Rosenkranz et al., 1977] NN: $\Theta(\log n)$ even on unweighted planar
- [Myazaki et al., 2009] cycles: $\frac{1+\sqrt{3}}{2} \approx 1.366$, unweighted graphs: 2
- [Kalyanasundaram et al., 1994] planar: 16-competitive (general?)
- [Megow et al., 2011] K+ algorithm is not constant competitive

genus g: 16(1+2g)-competitive

k distinct weights: 2k-competitive

wer bound from	n [Myazaki et a	I., 2009]	
	wer bound from	wer bound from [Myazaki et a	wer bound from [Myazaki et al., 2009]

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first task			
Improve the 2	2-arepsilon lower bound	rom [Myazaki	i et al., 2009]
Improve the 2	2-arepsilon lower bound t	from [Myazaki	i et al., 2009]
Improve the 2 we got	2-arepsilon lower bound t	from [Myazaki	i et al., 2009]

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ratio with x blocks: $\frac{(x-2)\frac{5}{2}(x-1)+2(x-1)+x(x-2)}{x(x-1)+x(x-2)}$





$$\begin{array}{rcl} r_1 & = & \frac{1}{2}(x-1) \\ t_1 & = & x-2 \\ o_1 & = & x-1 \\ e_1 & \geq & \frac{5}{2}(x-1) \\ \tilde{e}_1 & \geq & x-1 \end{array}$$

ratio x blocks, k levels: $\frac{(x-5)e_k+5\tilde{e}_k+xt_k}{xo_k+xt_k}$

 $k + \chi \iota_k$



ratio x blocks, k levels: $\frac{(x-5)e_k+5\tilde{e}_k+xt_k}{xo_k+xt_k} \approx \frac{5}{2} - 2^{-O(\sqrt{\log n})}$

online graph exploration	lower bound	advice	optimality	algorithm
advice complexity				



advice complexity



advice

- given to the agent at start
- function of input graph
- s-bit binary string
- "relevant" topology information

advice complexity





 v_0

 v_1





• $w(v_i, v_j) = 4 - \min\{i, j\}$

• unique optimal solution



lower bound on advice for optimality

Any optimal algorithm requires $\Omega(n \log n)$ bits in the worst case.



- $w(v_i, v_j) = 4 \min\{i, j\}$
- unique optimal solution
- agent needs log *n* advice

online graph exploration	lower bound	advice	optimality	algorithm

lower bound on advice for optimality

Any optimal algorithm requires $\Omega(n \log n)$ bits in the worst case.



- $w(v_i, v_j) = 4 \min\{i, j\}$
- unique optimal solution
- agent needs log n advice
- cannot distinguish next

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• problem: reversal



lower bound on advice for optimality

Any optimal algorithm requires $\Omega(n \log n)$ bits in the worst case.



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online graph exploration	lower bound	advice	optimality	algorithm
algorithm	tant-competitive	algorithm wi	th linear advice	
	tant-competitive			

rough idea

• traverse some tree



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lower bound

advice tells the value $\lceil \log(\frac{M}{n}) \rceil$ (*M* is unbounded, but can be done using $O(\log n)$ bits and traversing O(M) total cost.)

algorithm

Advice tells the value $\lceil \log(\frac{M}{n}) \rceil$ by encoding: (n, n', p, l')

Search for the first encountered edge e with $w(e) \in [\frac{M}{n^2}, M]$

- keep traversing the *cheapest* outgoing edge until *n*'-th vertex is encountered,
- consider the p-th incident edge e and let w(e) be its weight,
- $\lceil \log(M) \rceil = \lceil \log(w(e)) \rceil + l'$
- (n', p, l') are chosen in such way that e has the right property.

$O(\log n)$ bits are sufficient to encode n, n', p, l'; Cost: O(M)

- Such an edge e must exist (otherwise the MST weight < M).
- $l' \leq \lceil \log(M) \rceil \lceil \log(\frac{M}{n^2}) \rceil \leq 2 \log n$
- at most *n* cheapest $\left(\frac{M}{n^2}\right)$ outgoing edges; the cost to reach each one is at most $O\left(\frac{M}{n}\right)$.



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- cheap edges always explored by DFS
- cheap clusters connected with expensive edges



- cheap edges always explored by DFS
- cheap clusters connected with expensive edges
- some expensive edges are *tree edges* (i.e. from MST)
- advice must tell which ones in an efficient way

online graph exploration	lower bound	advice	optimality	algorithm

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cluster edges

- level 0: cheap
- level *i*: $w(e) \leq 2^i \frac{M}{n}$
- $\leq \frac{n}{2^{i}}$ level-*i* edges in MST

online graph exploration	lower bound	advice	optimality	algorithm

- level 0: cheap
- level *i*: $w(e) \leq 2^i \frac{M}{n}$
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identify tree edges

- levels in parallel
- separate advice for levels
- O(log i) bits per i-edge

online graph exploration	lower bound	advice	optimality	algorithm

- level 0: cheap
- level *i*: $w(e) \leq 2^i \frac{M}{n}$
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• all *i*-edges out: OUT

identify tree edges

- levels in parallel
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- O(log i) bits per i-edge



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identify tree edges

- levels in parallel
- separate advice for levels
- O(log i) bits per i-edge

- all *i*-edges out: OUT
- in edges: WAIT



online graph exploration	lower bound	advice	optimality	algorithm

- level 0: cheap
- level *i*: $w(e) \leq 2^{i} \frac{M}{n}$
- $\leq \frac{n}{2^i}$ level-*i* edges in MST

identify tree edges

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- all *i*-edges out: OUT
- in edges: WAIT
- Iater TRIGGER



online graph exploration	lower bound	advice	optimality	algorithm

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identify tree edges

- levels in parallel
- separate advice for levels
- $O(\log i)$ bits per *i*-edge

- all *i*-edges out: OUT
- in edges: WAIT
- Iater TRIGGER

managing multiple triggers is the core of the algorithm



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online graph exploration	lower bound	advice	optimality	algorithm
open problems				

- so is there a constant competitive algorithm or not?
- improve the lower bound $\frac{5}{2}-\varepsilon$
- any general algorithm better than $O(\log n)$?
- what can be done with polylogarithmic advice? or o(n)?

• lower bounds on advice / trade-off

online graph exploration	lower bound	advice	optimality	algorithm
open problems				

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