Introduction	Preliminaries	Building a QPTAS	$O(\sqrt{\log n})$ -approximation algorithm	Conclusion and Perspectiv	
A	pproximation		or Generalized Binary S Ited Trees	Search in	

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ANR DESCARTES, Poitier Oct. 4th, 2017

Introduction	Preliminaries	Building a QPTAS	$O(\sqrt{\log n})$ -approximation algorithm	Conclusion and Perspective
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1 Introduction

- 2 Preliminaries
- 3 Building a QPTAS
- 4 $O(\sqrt{\log n})$ -approximation algorithm
- 5 Conclusion and Perspective

I	ntroduction	Preliminaries	Building a QPTAS	$O(\sqrt{\log n})$ -approximation algorithm	Conclusion and Perspective
	Introductio	n			

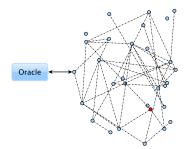
General Configuration of Searching Problem

A set of data organized in some structure

Introduction	Preliminaries	Building a QPTAS	$O(\sqrt{\log n})$ -approximation algorithm	Conclusion and Perspective
Introduc	tion			

General Configuration of Searching Problem

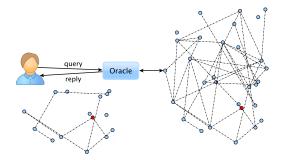
- A set of data organized in some structure
- An oracle replies to queries on the data



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Introduc	tion			

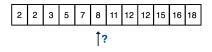
General Configuration of Searching Problem

- A set of data organized in some structure
- An oracle replies to queries on the data
- The oracle returns a subset of the data set which contains the target element



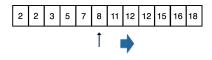
Binary Search

For an ordered array (or totally ordered set)



Binary Search

For an ordered array (or totally ordered set)



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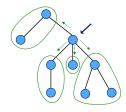
target x: 11

Binary Search

For an ordered array (or totally ordered set)

Our problem : Searching in Trees

- Data organized into a tree
- Target node x is known to the oracle, but not to the search algorithm
- The oracle returns the subtree in which the target lies



- Query : a node v





- Query : a node v



- Reply :
 - **true**, if *v* is the target

target is v



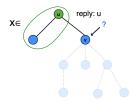


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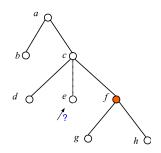


otherwise, return a neighbor u of v which is closer to the target x



Introduction	Preliminaries	Building a QPTAS	$O(\sqrt{\log n})$ -approximation algorithm	Conclusion and Perspective
Example	1			

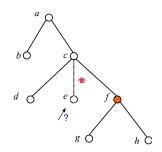
Query e



Target : f

Introduction	Preliminaries	Building a QPTAS	$O(\sqrt{\log n})$ -approximation algorithm	Conclusion and Perspective
Example	1			

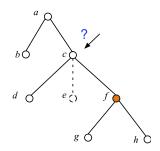
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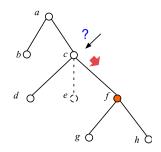
Query c



Target : f

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Example	1			

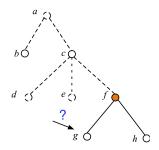
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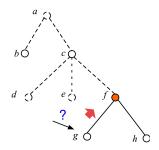
Query g



Target : f

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Example	1			

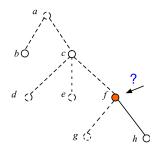
Query g



Target : f

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Example	1			

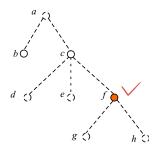
Query f



Target : f

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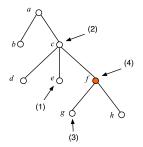
Found



Target : f

Example 1 : cost of locating the target

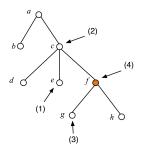
Unweighted variant



Total of 4 queries to locate target: e, c, g, f

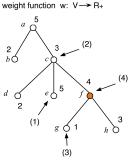
Example 1 : cost of locating the target

Unweighted variant



Total of 4 queries to locate target: e, c, g, f

Weighted variant

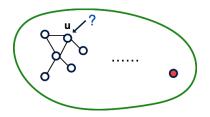


Cost of locating target = w(e)+w(c)+w(g)+w(f)= 5+3+1+4=13

General Graph Variation

General Graph :

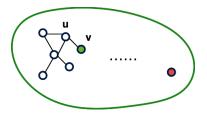
Query u



General Graph Variation

General Graph :

Query uReply a $v \in N(u)$, s.t. v is on the shortest path to the target



Search Strategy Problem in Trees

Setting

- Tree T = (V, E, w) with root r(T)
- **Cost of query** to vertex $v : w : V \to \mathbb{R}_+$, $\max_v w(v) = 1$
- **Cost of search strategy** A on tree T : worst-case cost of finding a target
- **Optimal strategy** : search strategy with minimal cost on *T*, costs *OPT*(*T*)

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Our Problem :

- Input : tree T = (V, E, w)
- Compute : Optimal strategy for generalized binary search query model

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Application Aspects

- Locating buggy nodes in network models
- Finding specific data in organized databases

Introduction P	Building a QPTAS	$O(\sqrt{\log n})$ -approximation algorithm	Conclusion and Perspective

State-of-the-Art

Time complexity to compute optimal search strategy in different graphs

Graph class	unweighted	weighted
Path	O(n) time	O(n ²) time [1]
Tree	O(n) time [2]	NP-hard [3]
Undirected Graph	$m^{\Theta(\log n)}$ under ETH [4]	PSPACE-complete [4]
Directed Graph	PSPACE-complete [4]	PSPACE-complete [4]

[1] Cicalese, Jacobs, Laber, Valentin, 2012

[2] Onak, Parys, 2006

[3] Dereniowski, Nadolski, 2006

[4] Emamjomeh-Zadeh, Kempe, Singhal, 2016

Our scenario : Weighted trees

- NP-hard, O(log n)-approximation algorithm. [Dereniowski, 2006]
- O(log log n log log n)-approximation. [Cicalese, Jacobs, Laber, Valentin, 2012]
- O(log n / log log n)-approximation. [Cicalese, Keszegh, Lidický, Pálvölgyi, Valla, 2015]

Introduction	Preliminaries	Building a QPTAS	$O(\sqrt{\log n})$ -approximation algorithm	Conclusion and Perspective
Our Res	ults			

Results for the Search Strategy Problem in Weighted Trees :

Introduction	Preliminaries	Building a QPTAS	$O(\sqrt{\log n})$ -approximation algorithm	Conclusion and Perspective
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Results for the Search Strategy Problem in Weighted Trees :

Theorem 1. The problem admits a QPTAS.

- **QPTAS** : Quasi-Polynomial-Time Approximation Scheme, $(1 + \varepsilon)$ -approximation algorithm running in $n^{\text{polylog}(n)}$ time, for any given $\epsilon > 0$.
- which implies that the problem is not APX-hard unless $NP \subseteq DTIME(n^{O(\log n)})$

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- which implies that the problem is not APX-hard unless $NP \subseteq DTIME(n^{O(\log n)})$

Theorem 2. The problem admits a poly-time $O(\sqrt{\log n})$ -approximation algorithm. improves previous approximation ratio

[ICALP 2017, Dereniowski, Kosowski, Uznański, Zou]

Preliminaries : Characterization of a Valid Search Strategy

Characterization of a Search Strategy

- A query ⇔ an interval of time
 - length of interval : l(v) = w(v)
 - beginning time of v : when node v is queried during the search(if it is queried)
 - this time interval do not depend on the replies to other queries

Preliminaries : Characterization of a Valid Search Strategy

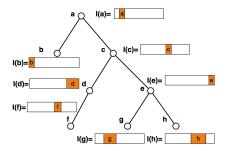
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 - length of interval : l(v) = w(v)
 - beginning time of v : when node v is queried during the search(if it is queried)
 - this time interval do not depend on the replies to other queries
- A query sequence \Leftrightarrow intervals of time I(v) for all $v \in V$

Preliminaries : Characterization of a Valid Search Strategy

Valid Search Strategy

■ If the intervals of nodes *u*, *v* overlap, some node on the u–v path in the tree must be queried before both *u* and *v*.

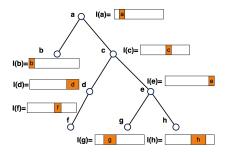


Extension of idea of : [Dereniowski, 2006]

Preliminaries : Characterization of a Valid Search Strategy

Valid Search Strategy

If the intervals of nodes u, v overlap, some node on the u-v path in the tree must be queried before both u and v.



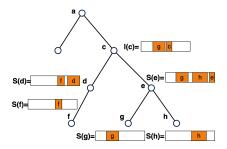
Lemma. There is a equivalence between optimizing search strategy and the following problem :

Assign intervals l(v) = [a, b] to $v \in V$, where |[a, b]| = w(v), s.t. $\forall u, v \in V$, $l(u) \cap l(v) \neq \emptyset \Rightarrow \exists z$ on the path from u to v and $max(l(t)) \leq min(l(u), l(v))$

Introduction	Preliminaries	Building a QPTAS	$O(\sqrt{\log n})$ -approximation algorithm	Conclusion and Perspective
Schedul	e assignme	nts		

Schedule assignments :

- Schedule S(v) : set of time intervals of "uncovered" nodes in the subtree of v
- The interval I(u) is "covered" by an ancestor x if x is assigned an earlier interval i.e. $(\max I(x) \le \min I(u))$.
- Schedule assignment : schedule S(v) for all $v \in V$



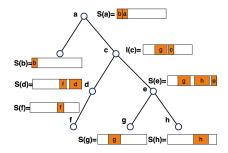
Constraints

No two interval in the schedule of a node could overlap



Schedule assignments :

- Schedule S(v) : set of time intervals of "uncovered" nodes in the subtree of v
- The interval *l(u)* is "covered" by an ancestor *x* if *x* a strictly earlier interval, e.g. (max *l(x)* <= min *l(u)*).
- Schedule assignment : schedule S(v) for all $v \in V$



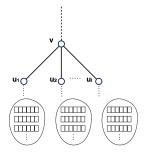
Introduction	Preliminaries	Building a QPTAS	$O(\sqrt{\log n})$ -approximation algorithm	Conclusion and Perspective
Roundin	g and Align	ment		
A		d slot veen integer multipl veen integer multipl	- 5	
		bo	x slot	

Rounding and Alignment: Time intervals of vertices are rounded and aligned to boxes or slots depending on their weight ("heavy": $w(v) > \frac{1}{\log n}$, "light": $w(v) \le \frac{1}{\log n}$)

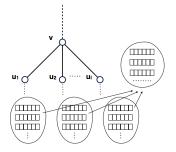


Lemma. Given tree T, there is an aligned schedule assignment S' with $cost_{S'}(T) \leq (1 + 11\varepsilon)OPT(T)$

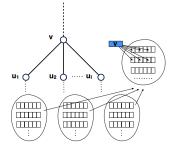
- Assume node v has children u₁,...u_l
- Store all valid aligned schedules S(v) for every node
- Compute in bottom-up manner :



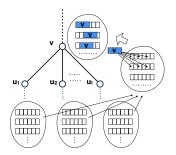
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- Compute in bottom-up manner :
 - Construct all valid aligned schedules for children u₁,...u_l
 - 2 Enumerate all possible start times of *v*



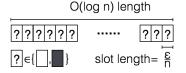
- Assume node v has children u₁,...u_l
- Store all valid aligned schedules S(v) for every node
- Compute in bottom-up manner :
 - Construct all valid aligned schedules for children u₁,...u_l
 - Enumerate all possible start times of v
 - 3 Obtain all possible schedules for v



Introduction	Preliminaries	Building a QPTAS	$O(\sqrt{\log n})$ -approximation algorithm	Conclusion and Perspective
Running	Time			

Fact. We have $1 \leq OPT(T) \leq \lceil log_2n \rceil$.

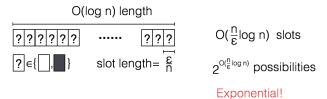
 \Rightarrow In Dynamic Programming, only consider aligned schedules of duration <= $O(\log n)$.

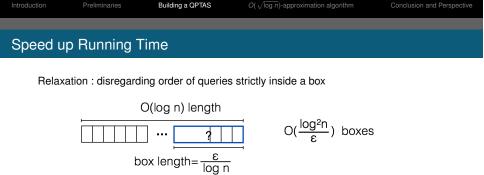


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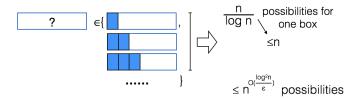
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 \Rightarrow In Dynamic Programming, only consider aligned schedules of duration <= $O(\log n)$.





For each box, store only number of full slots in this box.



Speed up Running Time

Relaxation computed by dynamic programming :

- can be computed exactly in n^{O(log² n)} time
 * running time can be reduced to n^{O(log n}/e²) by adaptively choosing box sizes
- not worse cost than optimal schedule
- not a valid schedule assignment
 * but : can be fixed at small extra cost [non-trivial, based on solution of optimal strategy in unweighted trees]

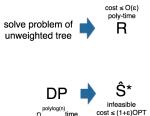
QPTAS	Introduction	Preliminaries	Building a QPTAS	$O(\sqrt{\log n})$ -approximation algorithm	Conclusion and Perspective
QPTAS					
	QPTAS				

 \hat{S}^* – solution by DP routine disregarding orders of light queries strictly inside a box. R – an subsequence of nodes based on optimal solution for unweighted trees S^+ – a valid (1 + ε) approximation



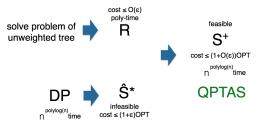
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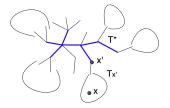
From QPTAS to $O(\sqrt{\log n})$ -approximation algorithm

Corollary of QPTAS

set $\varepsilon = 1 \Rightarrow n^{O(\log n)}$ running time constant-factor approximation

Recursive decomposition with central subtree T^*

Cost on T = Cost of locating x' in T* + Cost of executing the strategy in $T_{x'}$



From QPTAS to $O(\sqrt{\log n})$ -approximation algorithm

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Cost on T = Cost of locating x' in T* + Cost of executing the strategy in $T_{x'}$

Result

- We can choose *T**, such that :
 - Running time of every recursion level is poly(n), constant factor approximation
 - Recursion depth is bounded by $O(\sqrt{\log n})$
- Approximation factors add up along recursions
- Results in a polynomial-time O(\sqrt{log n})-approximation algorithm

Introduction	Preliminaries	Building a QPTAS	$O(\sqrt{\log n})$ -approximation algorithm	Conclusion and Perspective

Conclusion

Main Results

- A QPTAS (quasi-polynomial-time approximation scheme) for strategies of generalized binary search in weighted trees
 - implies the problem is not APX-hard unless $NP \subseteq DTIME(n^{O(\log n)})$
- An O(\sqrt{log n})-approximation polynomial-time algorithm for strategies of generalized binary search in weighted trees
 - improves previous approximation ratio

Open Questions

- Find constant-factor approximation?
- Results for other classes of graphs?
- Oracle with error-reply rate?

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Thanks!

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