



GDAŃSK UNIVERSITY  
OF TECHNOLOGY

Experimental analysis of binary search models in graphs

**Supervised by:** prof. dr. hab. inż. Dariusz Dereniowski

Michał Szyfelbein

October 6, 2025



**The aim:** Experimental analysis of selected generalized binary search problems. The aim of the analysis is to verify the hypotheses regarding the effectiveness of search algorithms.

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- 2024.06 – 2025.05: Development and formal analysis of the proposed algorithms.
- 2025.06 – 2025.10: Selection of models and algorithms for comparison, implementation, thesis writing.
- 2025.11 – 2025.12: execution of experiments, analysis and interpretation of the results, finalization of the thesis.



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- Implementation, about 60% complete:
  - Language: **python**,
  - Libraries: **networkx**,
  - Environment: **PyCharm**,
  - Versioning: **git** + **github**, <https://github.com/MSzyfel/Binary-Search>.
- Thesis, about 80% complete:
  - Language: **LaTeX**,
  - Environment: **Visual Studio Code**,
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- What is left:
  - Experiments,
  - Advanced data generation,
  - Optimization,
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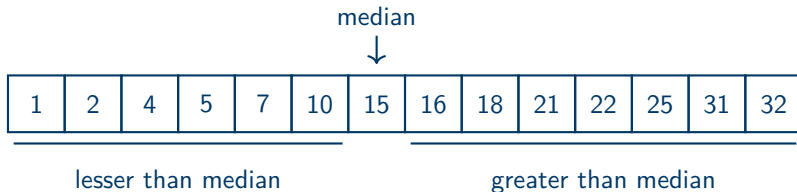


Figure: Example of a sorted array containing 14 elements.



## Definition

A **searcher** is required to find a hidden **target** vertex  $x$  in a graph  $G$ . To do so, they iteratively perform **queries** to an **oracle**, each about a chosen vertex  $v$ . After each such call, the oracle responds whether the target was found and if not, the searcher receives as a reply the connected component of  $G - v$  containing the target.

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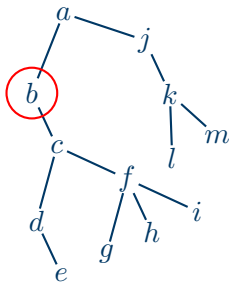


Figure: Query to  $b$

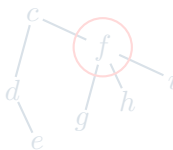


Figure: Query to  $f$



Figure: Query to  $d$



Figure: Query to  $c$

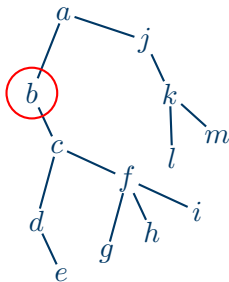


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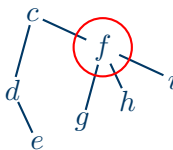


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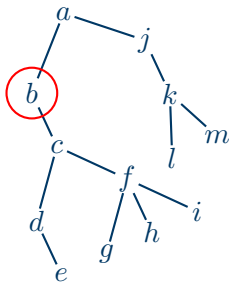


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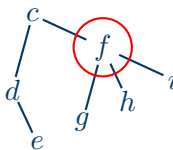


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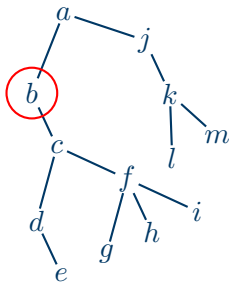


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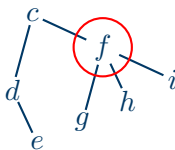


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**There are three main classes of graphs to be considered:**

- **Paths** - equivalent to searching in a sorted array.
- **Trees** - The most extensively studied model. **Our choice.**
- **General graphs** - Computationally hardest.



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A decision tree:

- $D = (V(D), E(D))$ ,  $V(D) = V(T)$  are vertices and  $E(D)$  are edges of  $D$ .
- $Q_D(T, x)$  - sequence of queries performed in order to find  $x$ .
- Cost of  $D$  in  $(T, c)$ :

$$\text{COST}_D(T, c) = \max_{x \in V(T)} \left\{ \sum_{q \in Q_D(T, x)} c(q) \right\}.$$

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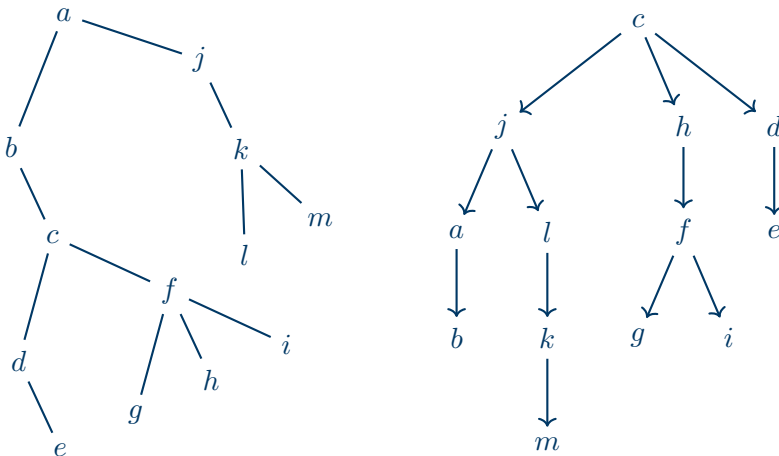


Figure: Sample input tree and a decision tree for it.



## Definition

Given a tree  $T$  and weight function  $c$ , the **Tree Search Problem** consists of finding a decision tree  $D$ , such that  $\text{COST}_D(T, c) = \text{OPT}(T, c)$ .

Unluckily, the Tree Search Problem is **strongly NP-Hard** even when restricted to binary trees and spiders of diameter at most 6. However, one can find **approximate** solutions.



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## Theorem

*Let  $c(v) = 1$  for every  $v \in V(T)$ . There exists an exact algorithm called **RankingBasedDT** for the Tree Search Problem running in linear time, such that the resulting decision tree uses at most  $\lfloor \log n \rfloor + 1$  queries.*

## Theorem

*Fix  $0 < \epsilon \leq 35$ . There exists a  $(1 + \epsilon)$ -approximation algorithm for the Tree Search Problem running in  $n^{O(\log n / \epsilon^2)}$  time.*

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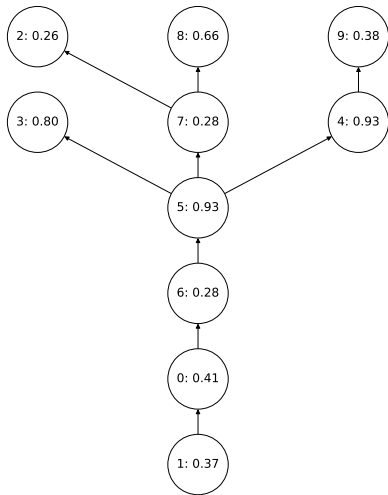


Figure: Input tree of size 10.

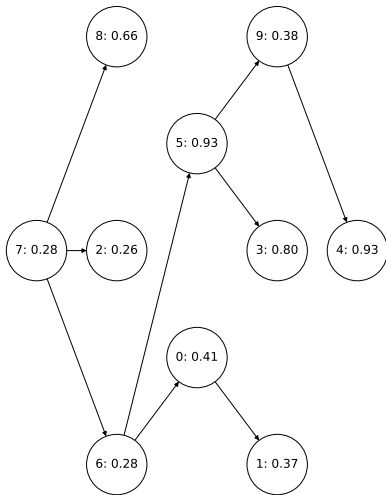


Figure: Decision tree of cost 2.8.

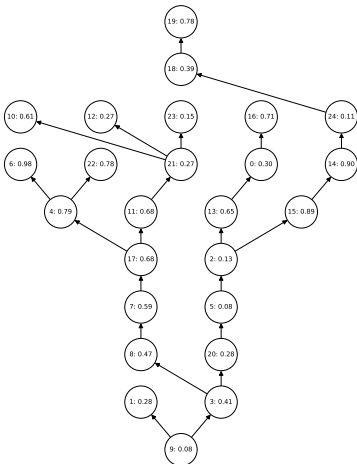


Figure: Input tree of size 25.

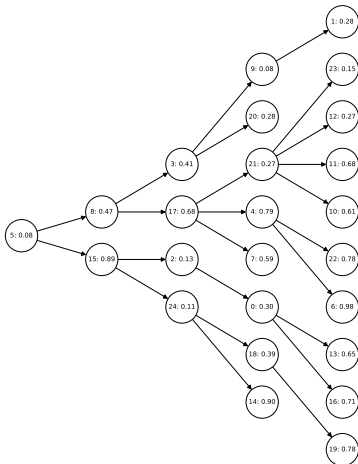


Figure: Decision tree of cost 3.

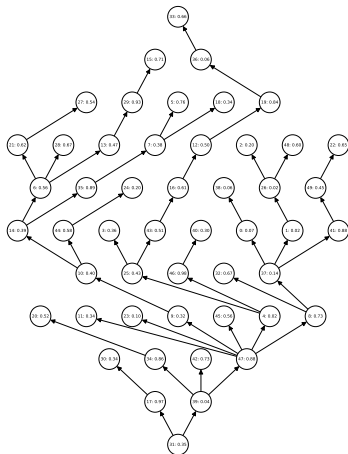


Figure: Input tree of size 50.

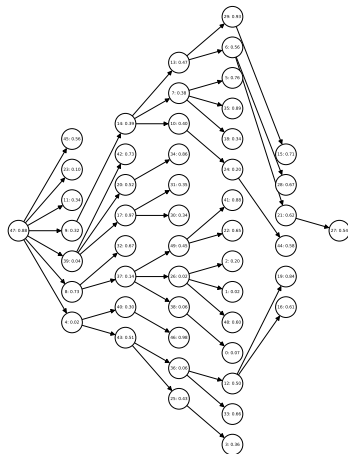


Figure: Decision tree of cost 3.78.

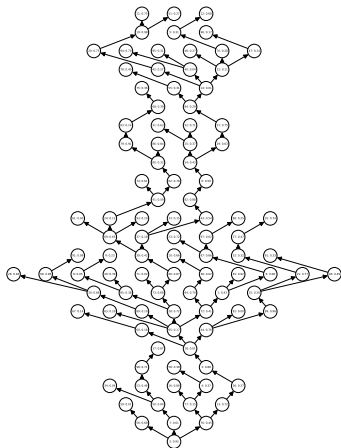


Figure: Input tree of size 100.

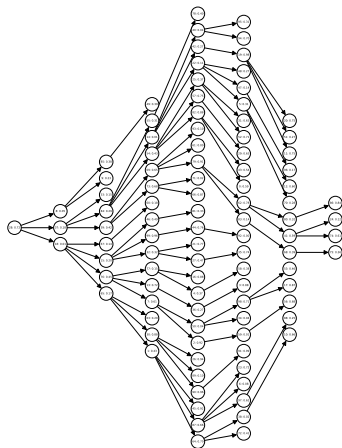


Figure: Decision tree of cost 4.85.

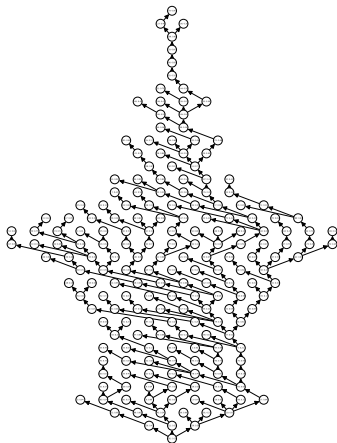


Figure: Input tree of size 200.

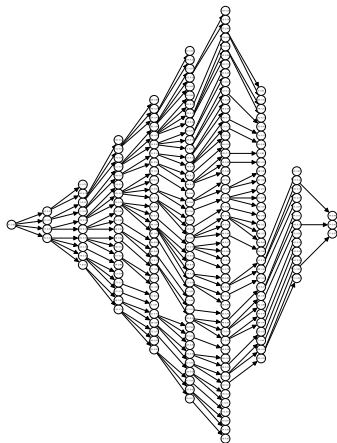


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**proc** ApproxDT ( $T$ ):

1. Set a global parameter  $k = 2^{\lfloor \sqrt{\log n} \rfloor + 2}$ .
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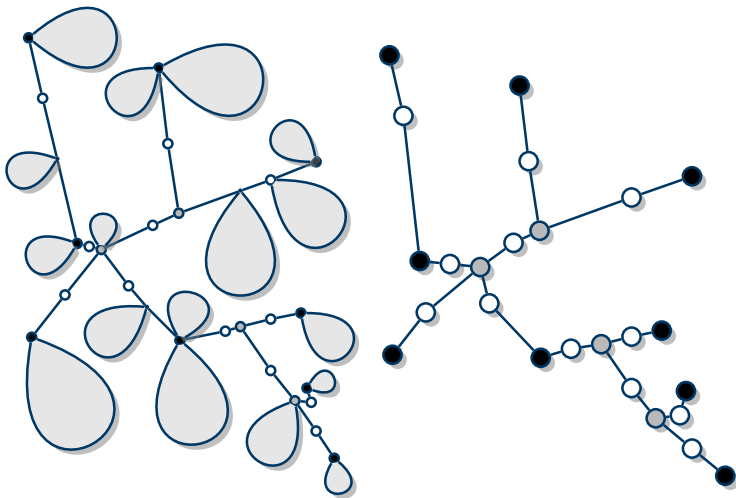
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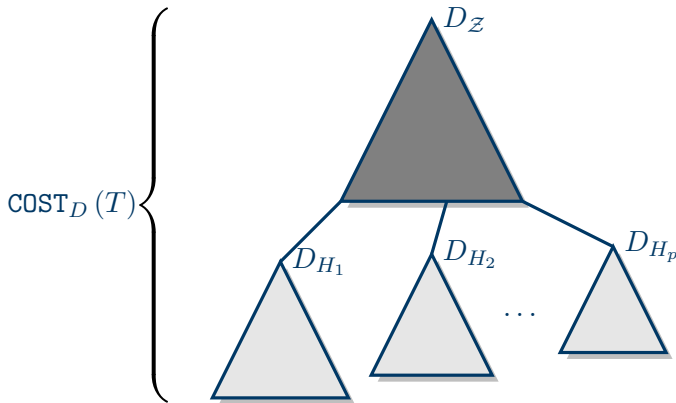


Figure: Structure of the decision tree  $D$

Since  $\text{COST}_{D_Z} \leq 2 \cdot \text{OPT}(T)$ , we have:

$$\begin{aligned}\text{COST}_D(T) &\leq \text{COST}_{D_Z} + \max_{H \in T-Z} \{\text{COST}_{D_H}(H)\} \\ &\leq 2 \cdot \log_{2\sqrt{\log n}}(n) \cdot \text{OPT}(T) \\ &= \frac{2 \log n}{\sqrt{\log n}} \cdot \text{OPT}(T) \\ &= 2\sqrt{\log n} \cdot \text{OPT}(T).\end{aligned}$$

- Let  $p \in \mathbb{N}$ ,
- Let  $k = a/pn$ , for some  $a \in \mathbb{N}$ .
- We define new cost function  $c'$ , called **aligned cost function**:

$$c'(v) = \begin{cases} \lceil c(v) \rceil_k, & \text{if } c(v) > pk, \text{ heavy vertex,} \\ \lceil c(v) \rceil_{\frac{1}{pn}}, & \text{otherwise, light vertex.} \end{cases}$$



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$$OPT(T, c') \leq \left(1 + \frac{2}{p}\right) \cdot OPT(T, c).$$

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*There exists a decision tree  $D$  for  $(T, c')$ , such that:*

1.  $COST_D(T, c') \leq \left(1 + \frac{3}{p}\right) \cdot OPT(T, c')$
2. *Starting point of each heavy query is aligned to a multiple of  $c$ ,*
3. *Starting point of each light query is aligned to a multiple of  $\frac{1}{pn}$ .*

We call a decision tree fulfilling the requirements 2. and 3. of the above lemma an **aligned decision tree**.



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For any vertex  $v \in V(T)$  and query  $q \in Q_D(T, v)$  the **contribution**  $\kappa_{T,c,k}(q, v)$  of  $u$  is defined as:

$$\kappa_{T,c}(q, v) = \begin{cases} 0, & \text{if its a light down response,} \\ c(q), & \text{otherwise.} \end{cases}$$

Then, the **aligned cost** of  $D$  is defined as:

$$\text{COST}'_D(T, c', k) = \max_{v \in V(T)} \left\{ \sum_{q \in Q_D(T, v)} \kappa_{T,c',k}(q, v) \right\}.$$

Let  $\text{OPT}'(T, c', k)$  denote the optimal aligned cost.

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## Proposition

Let  $T$  be a tree,  $c'$  an aligned cost function,  $p \in \mathbb{N}$ ,  $k$  the box size,  $n$  the size of the original input tree, and  $d$  be the depth. There exists a `DPTimelinesCosts` procedure which calculates an optimal aligned decision tree  $D$ , running in  $pn^{O(d)}$  time.

## Proposition

Let  $T$  be a tree,  $c'$  be an aligned cost function,  $D_A$  be a decision tree for  $T$  and  $F_C$  be forest of decision trees for  $T$  with all heavy groups contracted,  $p \in \mathbb{N}$  be a constant,  $k \in \mathbb{R}_{>0}$  be the box size. There exists a polynomial time `MergeDTs` procedure which returns a decision tree of cost at most:

$$\text{COST}_D(T, c', k) = \text{OPT}'(T, c', k) + 2pk \cdot \text{COST}_{F_C}(T, 1).$$



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Let  $T$  be a tree,  $c'$  an aligned cost function,  $p \in \mathbb{N}$ ,  $k$  the box size,  $n$  the size of the original input tree, and  $d$  be the depth. There exists a `DPTimelinesCosts` procedure which calculates an optimal aligned decision tree  $D$ , running in  $pn^{O(d)}$  time.

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**proc** QPTAS ( $T, \epsilon$ ):

1.  $p \leftarrow \lfloor 35/\epsilon \rfloor$ ,
2.  $d \leftarrow p^2 \cdot (\lfloor \log n \rfloor + 1)$ .
3.  $k \leftarrow 0$ .
4. Repeat the following steps until a decision tree  $D$  is found:
  1.  $k \leftarrow k + \frac{1}{pn}$ .
  2. For every  $v \in V(T)$ , if  $c(v) > pk$ , then  $c'(v) \leftarrow \lceil c(v) \rceil_k$ ,  
 $c'(v) \leftarrow \lceil c(v) \rceil_{\frac{1}{pn}}$  otherwise.
  3.  $D_A \leftarrow \text{DPTimelinesCosts}(T, c', p, k, n, d)$
  4. If  $D_A \neq \emptyset$ :
    1.  $T_C \leftarrow T$  with all heavy modules contracted.
    2.  $D_C \leftarrow \text{RankingBasedDT}(T_C)$ .
    3.  $D \leftarrow \text{MergeDTs}(T, D_A, D_C)$ .
5. Return  $D$ .

Let  $k'$  be the value of  $k$  for which  $D$  was found. Since:

$$k' \leq \frac{\text{OPT}'(T, c', k')}{d} = \frac{\text{OPT}'(T, c')}{p^2 \cdot (\lfloor \log n \rfloor + 1)}, \text{ we have that:}$$

$$\begin{aligned} \text{COST}_D(T, c') &\leq \text{OPT}'(T, c') + 2pk' \cdot (\lfloor \log n \rfloor + 1) \\ &\leq \text{OPT}'(T, c') + 2p \cdot (\lfloor \log n \rfloor + 1) \cdot \frac{\text{OPT}'(T, c')}{p^2 \cdot (\lfloor \log n \rfloor + 1)} \\ &\leq \left(1 + \frac{2}{p}\right) \cdot \text{OPT}'(T, c') \\ &\leq \left(1 + \frac{2}{p}\right) \cdot \left(1 + \frac{2}{p}\right) \cdot \left(1 + \frac{3}{p}\right) \cdot \text{OPT}(T, c) \\ &\leq \left(1 + \frac{35}{p}\right) \cdot \text{OPT}(T, c) = \left(1 + \frac{35}{\lceil \frac{35}{\epsilon} \rceil}\right) \cdot \text{OPT}(T, c) \\ &\leq (1 + \epsilon) \cdot \text{OPT}(T, c) \end{aligned}$$

**proc** MergeDTs ( $T, D_A$ ):

1. If  $F_C$  is connected,  $r = r(F_C)$ .
2. Else,  $r = r(D_A)$ .
3.  $D \leftarrow$  decision tree with root  $r$ .
4. For each  $T' \in T - r$ :
  1. Let forest  $F'_C$  be  $F_C$  restricted to  $T'_C$
  2. Let  $D'_A$  be  $D$  restricted to  $T'$ .
  3.  $D' \leftarrow \text{MergeDTs}(T', D'_A, F'_C)$
  4. Hang  $D'$  below  $r$  in  $D$ .
5. Return  $D$ .

## Definition

### Boxed decision tree:

1.  $D = (V(D), E(D), u, l)$ ,  $V(D)$  - nodes of  $D$ , called *boxes*,  $E(D)$  - edges of  $D$ ,  $u : V(T) \times V(D) \rightarrow \{0, 1/pn, 2/pn, \dots, k\}$  - *usage function* and  $l : V(D) \rightarrow \{0, 1/pn, 2/pn, \dots, k\}$  - *load function*.
2. For every  $b \in V(D)$ ,  $Q(b) = \{v \in V(T) \mid u(v, b) > 0\}$  - *query assignment*, vertices of  $T$ , such that queries to them overlap with  $b$ .
3. All boxes containing  $v$  form a left path. Additionally, for each interior box  $b$ ,  $l(b) = 0$  and  $u(v, b) = k$ .
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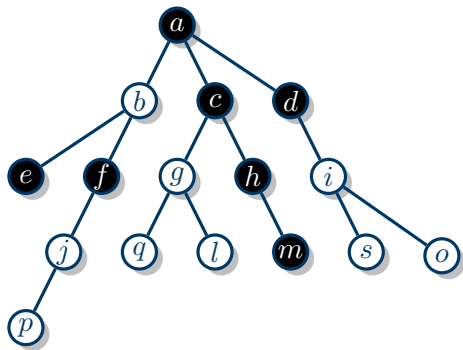


Figure: Input tree.

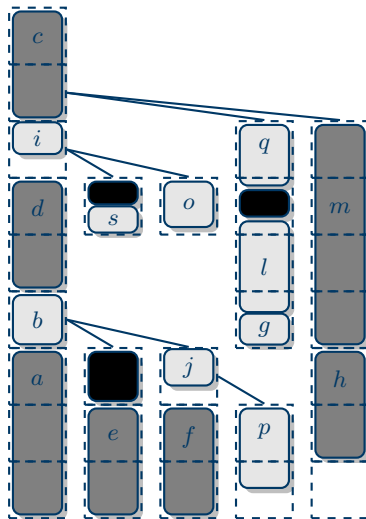


Figure: Boxed decision tree.

## Definition

**Boxline:**  $B = \langle (b_1, \tau_1), (b_2, \tau_2), \dots, (b_d, \tau_d) \rangle$ ,  $b_j$  - box, such that  $Q(b_j) = \emptyset$ ,  $\tau_j$  - boolean flag.

## Definition

**Left box-path** of  $D$ :

1.  $B_D = \langle q_1, f_1, (q_2, f_2), \dots, (q_h, f_h) \rangle$ ,  $q_j$  - box  $f_j$  - boolean flag, obtained by traversing boxes of  $D$  towards left. For each such box  $b_j$ ,  $q_j = l(b) + \sum_{v \in Q(b)} q(v, b)$ , whereas  $f_j$  denotes whether there exists a **transcending** query in  $Q(b_j)$ , i. e.:  $v \in Q(b_j)$  such that  $v \in Q(b_{j+1})$ .
2. Decision tree  $D$  with a left box-path  $B_D$  is **box-compatible** with boxline  $B$  ( $h \leq d$ ), if  $l(q_j) \geq l(b_j)$  and  $\tau_j \implies f_j$ .

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- Putting a query to vertex  $v$  at  $s$ -th slot of a box  $b$ :

1.  $\sigma(v) \leftarrow c(v)$ :
2. While  $\sigma(v) > 0$ :
  1.  $u(v, b) \leftarrow \min \{k - s/pn, \sigma(v)\}$ .
  2.  $\sigma(v) \leftarrow \sigma(v) - u(v, b)$ .
  3.  $b \leftarrow \text{left child of } b$ .

If such operation violates the definition of  $D$  or query to  $v$  transcends any box  $b_j$ , such that  $\tau_j$ . we mark  $D$  as **conflicted**.

- Bipartitioning of  $B$ . A bipartition of  $B$  consists of  $(B_1, B_2)$  such that:
  - $|B| = |B_1| = |B_2|$ .
  - $l(b_{1,j}) + l(b_{2,j}) - k = l(b_j)$ .
  - $(\tau_{1,j} \wedge \tau_{2,j} \iff \tau_j)$ .
  - $(\tau_{1,j} \vee \tau_{2,j})$ .

- Rotating a decision tree  $D$  around vertex  $v$ :
  1.  $q_h \leftarrow$  the box containing the end of the query to  $v$ .
  2. Sort vertices whose queries start in  $Q(q_h)$  according to  $c$ .
  3. Create box  $q$ .
  4. Move queries from  $Q(q_h)$  to  $Q(q)$ , so that all queries after  $v$  are in  $Q(q)$ ,
  5. Hang  $q'_h$  as a right child of  $q_h$ .
  6. Rehang left child of  $q_h$  as a left child of  $q$ .



Figure: No  
children

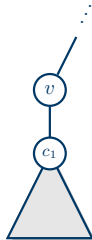


Figure: One child

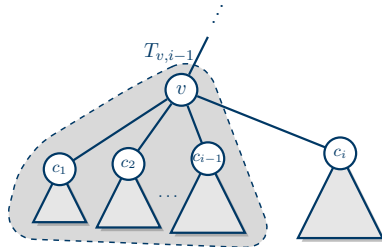


Figure: Many children





1. For  $1 \leq b \leq d$  and  $0 \leq s \leq (k/pn$  **if**  $c(v) > pk$  **else**  $0)$ :
  1.  $D \leftarrow P$
  2. Try putting query to  $v$  at the  $s$ -th slot of  $q_b$ .
  3. If there are no conflicts in  $D$ .
    1. If  $\text{COST}'_D(T_{v,i}, c, k) \leq dk$ , then return  $D$ , else return  $\emptyset$ .
2. Return  $\emptyset$ .

1.  $\mathcal{D} \leftarrow \emptyset$ .
2. For  $1 \leq b \leq d$  and  $0 \leq s \leq (k/pn$  **if**  $c(v) > pk$  **else**  $0$ ):
  1.  $D \leftarrow P$
  2. Try putting query to  $v$  at the  $s$ -th slot of  $q_b$ .
  3. If there are no conflicts in  $D$ .
    1. If  $\text{COST}'_D(T_{v,i}, c, k) \leq dk$ :
      1.  $P' \leftarrow$  left box-path of  $D$ .
      2.  $h \leftarrow$  index of the last box  $q_h$  occupied by the query to  $v$ .
      3. For  $h < j \leq d$ ,  $b'_j \leftarrow 0$ ,  $t'_j \leftarrow \text{False}$ .
      4.  $D' \leftarrow \text{DPTimelinesCosts}(T_{c_1}, c, P')$ .
      5. Put query to  $v$  at the  $s$ -th slot of  $q_b$ .
      6. Rotate left path of  $D'$  around  $v$ .
      7.  $\mathcal{D} \leftarrow \mathcal{D} \cup \{D \text{ and } D' \text{ with their left paths aligned}\}$ .
3. Return  $\arg \min_{D \in \mathcal{D}} \{\text{COST}'_D(T_{v,i}, c, k)\}$ .

1.  $\mathcal{D} \leftarrow \emptyset$ .
2. For each bipartition  $(B_1, B_2)$  of  $B$ :
  1.  $D_1 \leftarrow \text{DPTimelinesCosts}(T_{v,i-1}, c, P_1)$ .
  2.  $h \leftarrow$  index the last box  $q_{1,h}$  occupied by the query to  $v$ .
  3. For  $h \leq j \leq d$ ,  $b_{2,j} \leftarrow 0$ ,  $t'_{2,j} \leftarrow \text{False}$ .
  4.  $D_2 \leftarrow \text{DPTimelinesCosts}(T_{c_i}, c, P_2)$ .
  5. Rotate left path of  $D'$  around  $v$ .
  6.  $\mathcal{D} \leftarrow \mathcal{D} \cup \{D_1 \text{ and } D_2 \text{ with their left paths aligned}\}$ .
3. Return  $\arg \min_{D \in \mathcal{D}} \{\text{COST}'_D(T_{v,i}, c, k)\}$ .



Figure: Boxline  $B$



Figure: Bipartition of  $B$



Figure: Decision tree  $D_1$



Figure: Boxline  $B_2$

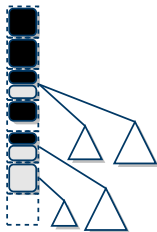


Figure: Decision tree  $D_2$

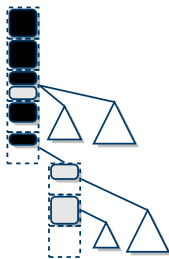


Figure: Rehang step

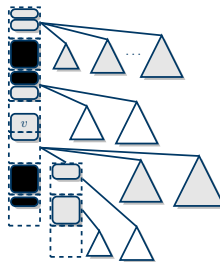


Figure: Resulting decision tree  $D$