

The Eigen-Distribution of Weighted AND-OR Trees

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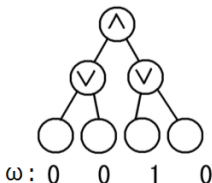
Outline

In 2007, Liu–Tanaka characterized the distribution which achieves the distributional complexity for binary AND-OR tree, and showed the uniqueness of such distribution.

The characterization is extended to multi-branching trees, but the uniqueness was not proved.

We introduce the *weighted tree*, and give a proof of the uniqueness.

An **AND-OR tree** is a tree whose root is labeled by AND (\wedge) and the internal nodes are level-by-level labeled by OR (\vee) or AND alternatively except for leaves.
Such a tree is also called *Game tree*.
For AND-OR tree \mathcal{T} , function ω from the set of all leaves of \mathcal{T} to $\{0, 1\}$ is called an **assignment**.



An assignment is denoted by a 0-1 sequence.
(e.g $\omega = 0010$)

Algorithms

An algorithm A tells how to proceed to evaluate a tree.
Algorithms have the following properties:

- deterministic :
The choice of leaves in each step is unique.
- depth-first :
If an algorithm evaluates the value of some subtree, it never evaluate another subtree until it finishes to evaluate the current one.

An algorithm is *directional* if there is some linear ordering on the leaves such that the computation follows this ordering.

Example of algorithm

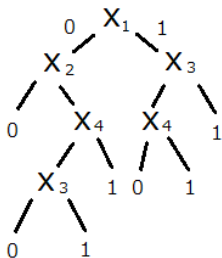


Figure: algorithm

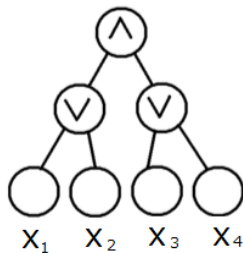
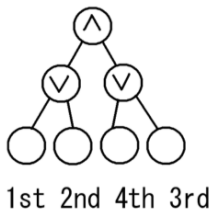


Figure: AND-OR tree

Notation

Let A be a directional algorithm whose priority of searching leaves is as follows:

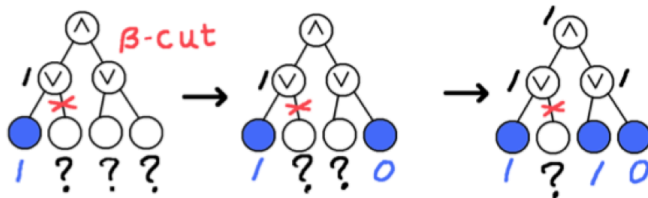


We say $A = \mathbf{1243}$.

Given an assignment ω and an algorithm A , $C(A, \omega)$ denotes the number of leaves checked by A under ω .

How to evaluate the cost

For the directional algorithm $A = \mathbf{1243}$ and $\omega = 1010$



In this case, $C(A, \omega) = 3$

Let d be a (probability) distribution on the set of assignments, the expected cost of A under the distribution d is defined by

$$C(A, d) := \sum_{\omega} d(\omega) C(A, \omega)$$

Given a class of algorithms \mathcal{A} ,
distributional complexity w.r.t \mathcal{A} is defined by:

$$\max_d \min_{A \in \mathcal{A}} C(A, d)$$

A distribution d satisfying

$$\min_{A \in \mathcal{A}} C(A, d) = \max_d \min_{A \in \mathcal{A}} C(A, d)$$

is called *eigen-distribution* (w.r.t \mathcal{A}).

Liu–Tanaka gave a characterization of the eigen-distribution for uniform binary tree.

The “ i -set” is the set of assignments which are difficult to evaluate.

Definition (i -set for n -branching trees)

Given an n -branching tree \mathcal{T} , $i \in \{0, 1\}$, the **i -set** consists of assignments such that

- the root has value i ,
- if an AND-node has value 0 (or OR-node has value 1), just one of its children has value 0 (1), and other $n-1$ children have 1 (0).

Example of i -set

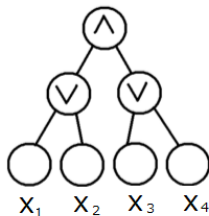
For AND-OR tree \mathcal{T}_2^2 ,

1-set=

{1010, 1001, 0110, 0101}

0-set=

{1000, 0100, 0010, 0001}



For example, assignment 1110 \notin 1-set.

Definition

E^1 -distribution (w.r.t \mathcal{A}) is a distribution on 1-set whose expected cost is independent of the choice of algorithms in \mathcal{A} .

Theorem (Liu–Tanaka (2007))

Let \mathcal{T} be a uniform binary AND-OR tree.

Then, the E^1 -distribution is the unique eigen-distribution, especially, which is the uniform distribution on 1-set.

Theorem (Suzuki–Nakamura (2012))

\mathcal{T} : uniform binary AND-OR tree,

\mathcal{A} : (closed) set of algorithms

Then, eigen-distribution is equivalent to E^1 -distribution w.r.t \mathcal{A} .

Furthermore, if \mathcal{A} is the set of all directional algorithms, the uniqueness fails.

Theorem (Peng et al. (2016))

\mathcal{T} : n -branching AND-OR tree

\mathcal{A} : (closed) set of algorithms

Then, eigen-distribution is equivalent to E^1 -distribution w.r.t \mathcal{A} .

Theorem (Peng et al. (2016))

Let \mathcal{T} be a n -branching AND-OR tree of height 2. Then E^1 -distribution is the uniform distribution on 1-set.

It remains that

Theorem

*For any n -branching AND-OR tree,
 E^1 -distribution is the uniform distribution on 1-set.*

To show this theorem,
we generalize the definition of “cost”.

For any algorithm A , and assignment ω , we define
 $\#_0(A, \omega) :=$ the number of leaves checked by A and
assigned 0 under ω .

$\#_1(A, \omega) :=$ the number of leaves checked by A and
assigned 1 under ω .

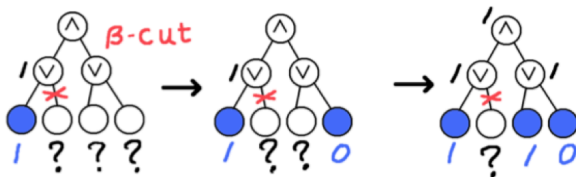
Definition

Let $a, b \in \mathbb{R}_{>0}$. The **generalized cost** $C(A, \omega; a, b)$ of A under ω is defined as follows:

$$C(A, \omega; a, b) := a \cdot \#_0(A, \omega) + b \cdot \#_1(A, \omega)$$

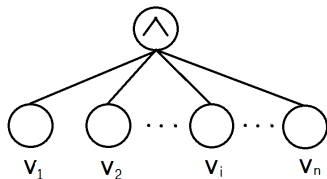
$C(A, d; a, b)$, $E^1(a, b)$ -distribution, ... are defined by the
same way.

For the directional algorithm $A = \mathbf{1243}$ and $\omega = 1010$



In this case, $C(A, \omega; a, b) = a + 2b$

Example (AND-OR tree of height 1)



1-set = $\{\omega_0\}$, 0-set = $\{\omega_i \mid i = 1, 2, \dots, n\}$

where $\omega_0(v_j) = 1$, $\omega_i(v_j) = \begin{cases} 1 & (i \neq j) \\ 0 & (i = j) \end{cases}$ ($j = 1, 2, \dots, n$)

Then,

$$C(A, d_{uni}(1\text{-set}); a, b) = nb,$$

$$C(A, d_{uni}(0\text{-set}); a, b) = a + \frac{n-1}{2}b$$

Theorem

Let $a, b \in \mathbb{R}_{>0}$. For any n -branching AND-OR tree, $E^1(a, b)$ -distribution is the uniform distribution on 1-set.

The same statement hold for $E^0(a, b)$ -distribution.

We prove this by induction on the height h .

Induction step

We consider the $E^1(a, b)$ -distribution d and assume h is even. The proof consists of two parts.

Lemma (1)

The probability of an assignment depends only on the value of nodes in height h .

The proof is essentially the same as the case height 1.
We use the condition “*nondirectional*” here.

We define the distribution d' for AND-OR tree of height h .

$$d'(\omega') := \sum_{\omega \in \Omega_{\omega'}} d(\omega)$$

where $\Omega_{\omega'} := \{\omega \mid \omega \text{ assigns } \omega' \text{ to the nodes of height } h.\}$

By the previous lemma, d' can be represented by

$$d'(\omega') = C \cdot d(\omega)$$

We should note that the cardinality of $\Omega_{\omega'}$ is independent of ω' ,

Lemma (2)

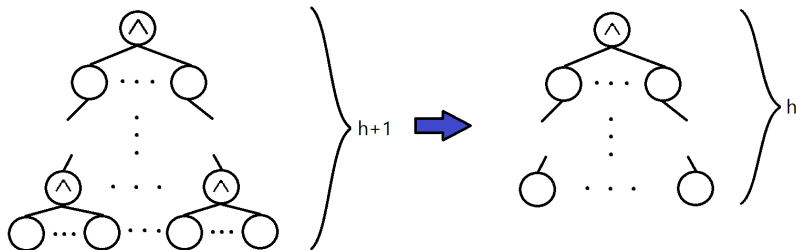
d' is an $E^1(a + \frac{n-1}{2}b, nb)$ -distribution for AND-OR tree of height h .

(sketch)

Given any algorithm A for height h , we can define an A' for height $h + 1$ satisfying

$$C(A', d; a, b) = C(A, d'; a + \frac{n-1}{2}b, nb)$$

Since d is $E^1(a, b)$ -distribution, the claim holds.



cost: $0 \Rightarrow a, 1 \Rightarrow b$

$d : E^1(a, b)$ -distribution

cost: $0 \Rightarrow a + \frac{n-1}{2}b, 1 \Rightarrow nb$

$d' : E^1(a + \frac{n-1}{2}, nb)$ -distribution

Recall (height 1)

$$C(A, d_{uni}(0\text{-set}); a, b) = a + \frac{n-1}{2}b,$$

$$C(A, d_{uni}(1\text{-set}); a, b) = nb$$

Theorem

Let $a, b \in \mathbb{R}_{>0}$. For any n -branching AND-OR tree, $E^1(a, b)$ -distribution is the uniform distribution on 1-set.

(proof)

By induction hypothesis, d' is the uniform distribution on 1-set for height h .

Since $d = \frac{1}{c}d'$, so d is also the uniform distribution on 1-set for height $h + 1$.



Since eigen-distribution is equivalent to E^1 -distribution, we get the uniqueness of the eigen-distribution.

Corollary

Let \mathcal{T} be an n -branching AND-OR tree.

Then E^1 -distribution is the uniform distribution on 1-set.

Generally, the following does NOT hold for AND-OR tree:

$$E^1(a, b)\text{-distribution} \Leftrightarrow \text{eigen-distribution}$$

For example, if height is 1

$$C(A, d_{uni}(1\text{-set}); a, b) \leq C(A, d_{uni}(0\text{-set}); a, b)$$

$$\Leftrightarrow nb \leq a + \frac{n-1}{2}b \quad \Leftrightarrow \quad \frac{n+1}{2}b \leq a$$

Moreover, if the equality holds, then there are uncountably many eigen-distributions.

So, the uniqueness of the eigen-distribution for weighted tree fails.

References

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2. T. Suzuki and R. Nakamura, “The eigen distribution of an AND-OR tree under directional algorithms,” *IAENG International Journal of Applied Mathematics*, vol. 42, no. 2, pp. 122–128, 2012.
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4. S. Okisaka, W. Peng, W. Li and K. Tanaka. “The eigen-distribution of weighted game trees,” *Lecture Notes in Computer Science*. 10627(2017), pp. 286–297.