



Associative and commutative tree representation for Boolean functions

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 - Plane binary model
 - Commutative or associative models
- 2 Probability distributions on Boolean functions
 - Relation between complexity and probability
 - Key ideas of the proof
- 3 Conclusion and perspectives

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Tree representation for Boolean functions

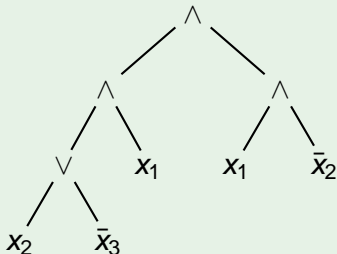
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Boolean formulas

- A set of literals: $\{x_1, \bar{x}_1, \dots, x_n, \bar{x}_n\}$
- A set of connectors: $\{\wedge, \vee\}$

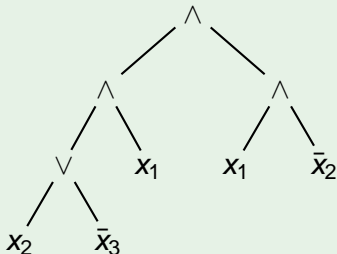
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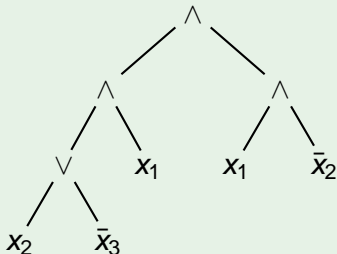
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- The set of all plane binary and labelled trees: \mathcal{T}_n
- The **size** of a tree corresponds to its number of leaves

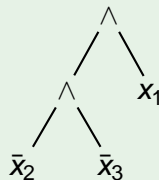
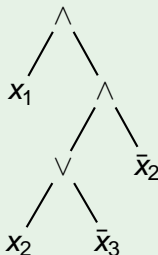
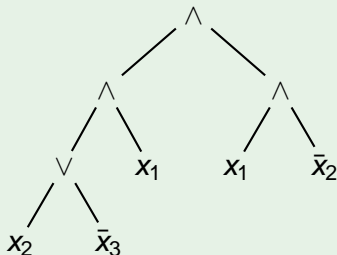
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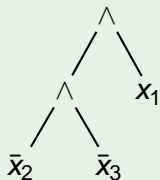
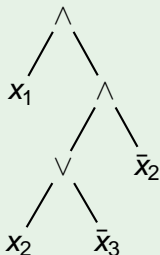
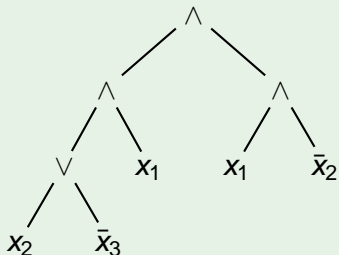
$$((x_2 \vee \bar{x}_3) \wedge x_1) \wedge (x_1 \wedge \bar{x}_2) \sim x_1 \wedge ((x_2 \vee \bar{x}_3) \wedge \bar{x}_2) \sim (\bar{x}_2 \wedge \bar{x}_3) \wedge x_1$$



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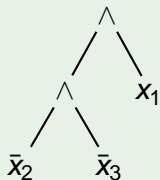
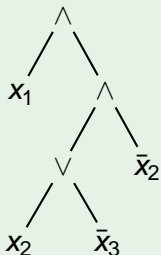
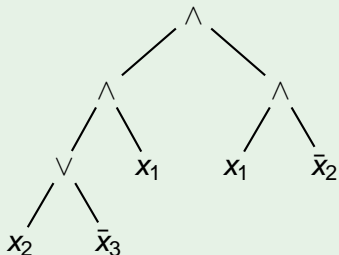


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- The **complexity** of a function is the size of its smallest trees

Probability distribution

- \mathcal{A} a subset of \mathcal{T}_n
- $\mathcal{A}(m)$ the set of all formulas of size m in \mathcal{A}

[Lefman, Savický. *Some typical properties of large And/Or boolean formulas*, 1997]

[Chauvin, Flajolet, Gardy, Gittenberger. *And/Or trees revisited*, 2004]

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$$\mathbb{P}_n(f) = \mu_n(\mathcal{A}_f).$$

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- Using Drmota-Lalley-Woods Theorem:

$\mathbb{P}_n(\cdot)$ is a probability distribution on \mathcal{F}_n .

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Previous results

Let $f \in \mathcal{F}_n$ be a function of complexity $L(f)$.

And/Or plane binary trees

$$\mathbb{P}_n(f) = \Theta \left(\frac{1}{n^{L(f)+1}} \right), \text{ when } n \text{ tends to infinity.}$$

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Implication plane binary trees

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[Fournier, Gardy, G., Gittenberger, 2008. *Complexity and Limiting Ratio of Boolean Functions over Implication.*]

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Almost all functions are of polynomial complexity in n , when $n \rightarrow \infty$.

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Remark: Formulas with symmetries have a smaller limiting ratio than others.

Natural properties of the connectors

Let us take into account some properties of the operators.

What about:

associativity or

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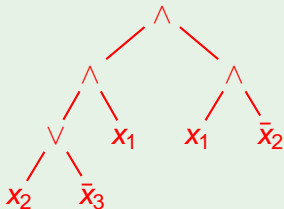
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- How does the tree structure evolve with such properties?
- Which behaviour for the new distributions on Boolean functions?

Tree models

connectors without
associativity and commutativity

plane binary trees



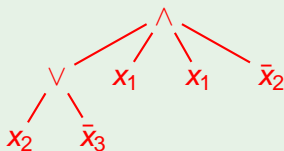
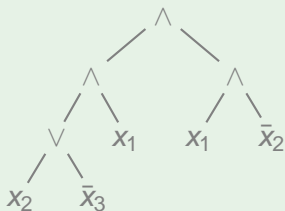
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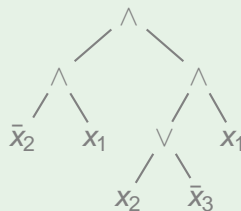
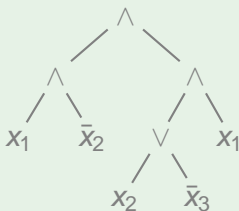
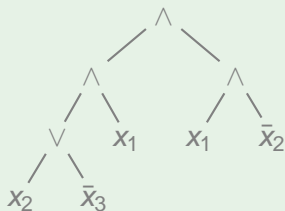
associative connectors

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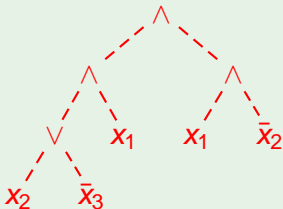
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connectors without associativity and commutativity	plane binary trees
associative connectors	plane non-binary trees
commutative connectors	non-plane binary trees



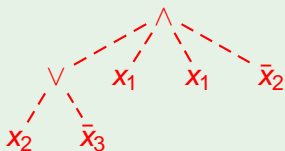
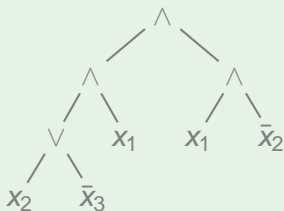
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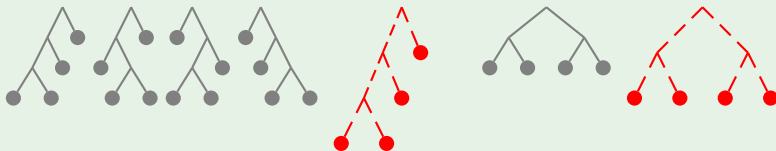
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Remark: The commutative models increase the limiting ratio of symmetric formulas.



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Probability distributions on Boolean functions

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Theorem

Let us fix one of the four tree models, then

$$\mathbb{P}_n(f) = \frac{\lambda_f}{n^{L(f)+1}} + \mathcal{O}\left(\frac{1}{n^{L(f)+2}}\right), \text{ when } n \text{ tends to infinity.}$$

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The constant λ_f is not the same from one model to another.

Corollary: Most trees computing f have a simple structure.

Key ideas [Kozik; Fournier *et al*]

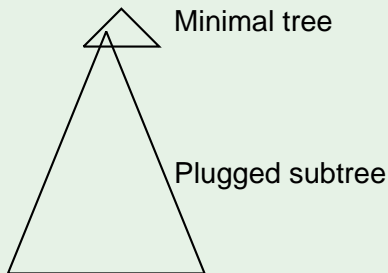
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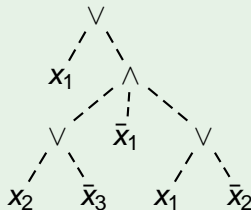
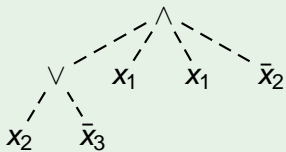
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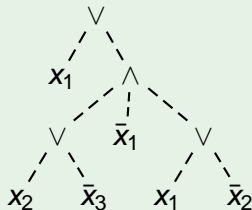
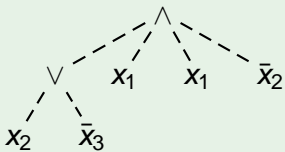
1. Definition of **simple family** of trees for f
2. All other trees for f :
 - i. some “non-necessary” restrictions in their structures
 - ii. their quantity is negligible
3. **Theorem:** Almost all trees computing the function f are obtained by plugging a special subtree in a minimal tree of f .



The associative and commutative model

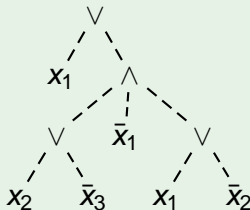
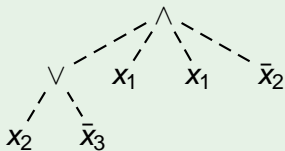


The associative and commutative model



$$\mathcal{T}^{\vee} = \mathcal{X} \mid \vee \text{MSET}^{\geq 2}(\mathcal{T}^{\wedge})$$

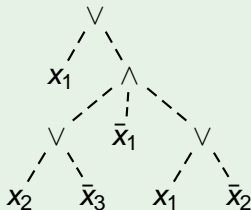
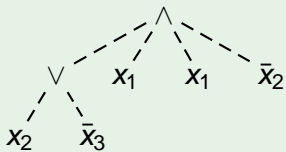
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$$T^\vee(z) = 2nz + \exp\left(\sum_{i \geq 1} \frac{T^\wedge(z^i)}{i}\right) - 1 - T^\wedge(z),$$

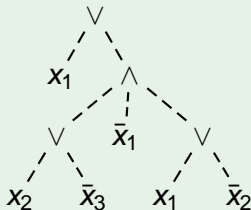
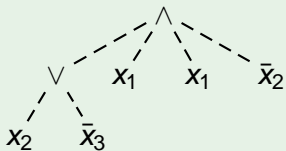
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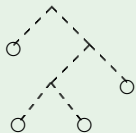
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$$T(z) = 2T^\vee(z) - 2nz.$$

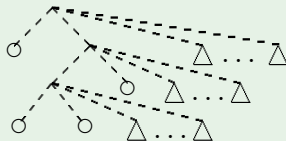
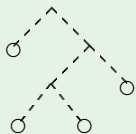
Main lemma

Let \mathcal{V} be a fixed subset of literals and
A a tree whose leaves are labelled by \mathcal{V} :



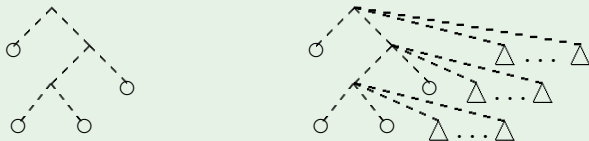
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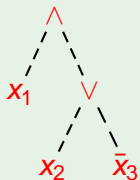
Lemma

$$\mu_n(\mathcal{F}_A) = \Theta\left(\frac{1}{n^{|A|}}\right), \text{ when } n \text{ tends to infinity.}$$

The proof is adapted from Kozik's proof on plane binary trees.

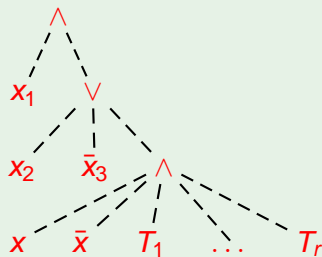
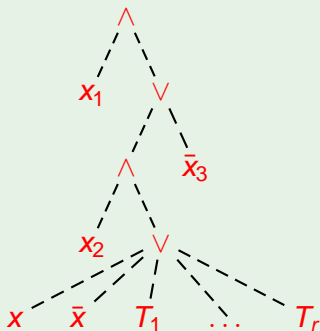
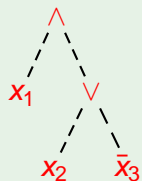
Valid expansions

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Four types of valid expansions applied on minimal trees of f give:

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$$\mathbb{P}_n(f) \underset{n \rightarrow \infty}{\gtrsim} \frac{\lambda_f}{n^4}.$$

All other trees computing f belong to families containing at least 5 restrictions.

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Conclusion and perspectives

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Conclusion

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- Although the commutative models give more importance to symmetric formulas, it is not sufficient for modifying the asymptotic order of the probability of fixed functions.
- According to these results on probabilities on fixed functions, we conjecture to have the same behaviour:
Almost all functions are of low complexity.

Perspectives

To obtain other kind of probability distributions:

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Perspectives

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- Changing the size notion for associative model,
- Using direct acyclic graphs instead of trees.

