

# Asymptotic probability of Boolean functions over implication

Danièle Gardy, Univ. Versailles

*with*

Hervé Fournier and Antoine Genitrini, Univ. Versailles

Bernhard Gittenberger, T.U. Wien

April 2008

# Outline

- Boolean expressions and trees
- A restricted propositional calculus
- Tautologies
- Probability and complexity of a Boolean function
- Main result: sketch of proof
- Extensions and open questions

# Boolean expressions

$$((x \vee \bar{x}) \wedge x) \wedge (\bar{x} \vee (x \vee \bar{x}))$$

$$(x \vee (y \wedge \bar{x})) \vee (((z \wedge \bar{y}) \vee (x \vee \bar{u})) \wedge (x \vee y))$$

# Boolean expressions

$$((x \vee \bar{x}) \wedge x) \wedge (\bar{x} \vee (x \vee \bar{x}))$$

$$(x \vee (y \wedge \bar{x})) \vee (((z \wedge \bar{y}) \vee (x \vee \bar{u})) \wedge (x \vee y))$$

Probability that a “random” expression on  $n$  boolean variables is a tautology (always true)?

# Boolean expressions

$$((x \vee \bar{x}) \wedge x) \wedge (\bar{x} \vee (x \vee \bar{x}))$$

$$(x \vee (y \wedge \bar{x})) \vee (((z \wedge \bar{y}) \vee (x \vee \bar{u})) \wedge (x \vee y))$$

Probability that a “random” expression on  $n$  boolean variables is a tautology (always true)?

- $n = 1$ : 4 boolean functions;  $Proba(True) = 0.2886$
- $n = 2$ : 16 boolean functions;  $Proba(True) = 0.209$
- $n = 3$ : 256 boolean functions;  $Proba(True) = 0.165$

# Boolean expressions

$$((x \vee \bar{x}) \wedge x) \wedge (\bar{x} \vee (x \vee \bar{x}))$$

$$(x \vee (y \wedge \bar{x})) \vee (((z \wedge \bar{y}) \vee (x \vee \bar{u})) \wedge (x \vee y))$$

Probability that a “random” expression on  $n$  boolean variables is a tautology (always true)?

- $n = 1$ : 4 boolean functions;  $Proba(True) = 0.2886$
- $n = 2$ : 16 boolean functions;  $Proba(True) = 0.209$
- $n = 3$ : 256 boolean functions;  $Proba(True) = 0.165$
- $n \rightarrow +\infty$ :  $2^{2^n}$  boolean functions

$$Proba(True) \sim ?$$

# Boolean expressions

$$((x \vee \bar{x}) \wedge x) \wedge (\bar{x} \vee (x \vee \bar{x}))$$

$$(x \vee (y \wedge \bar{x})) \vee (((z \wedge \bar{y}) \vee (x \vee \bar{u})) \wedge (x \vee y))$$

Probability that a “random” expression on  $n$  boolean variables is a tautology (always true)?

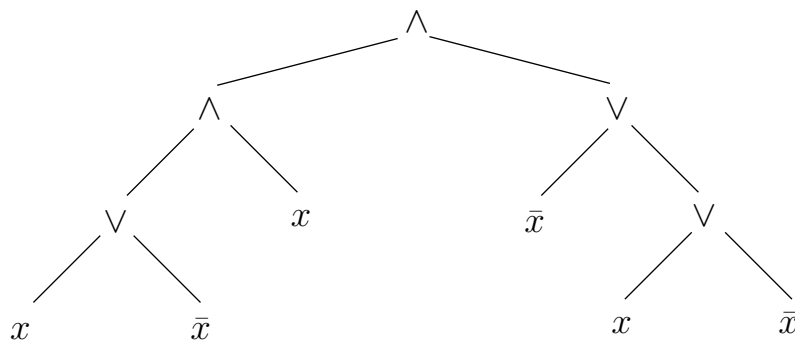
- $n = 1$ : 4 boolean functions;  $Proba(True) = 0.2886$
- $n = 2$ : 16 boolean functions;  $Proba(True) = 0.209$
- $n = 3$ : 256 boolean functions;  $Proba(True) = 0.165$
- $n \rightarrow +\infty$ :  $2^{2^n}$  boolean functions

$$Proba(True) \sim ?$$

$Proba(f)$  for any boolean function  $f$ ?

# Boolean expressions and trees

$$((x \vee \bar{x}) \wedge x) \wedge (\bar{x} \vee (x \vee \bar{x}))$$



Consider a well-formed boolean expression

- Choose set of logical connectors, with arities  
↔ Choose labels and arities for internal nodes
- Choose set of boolean literals for the leaves  
↔ Choose labels for leaves



# Boolean expressions and trees

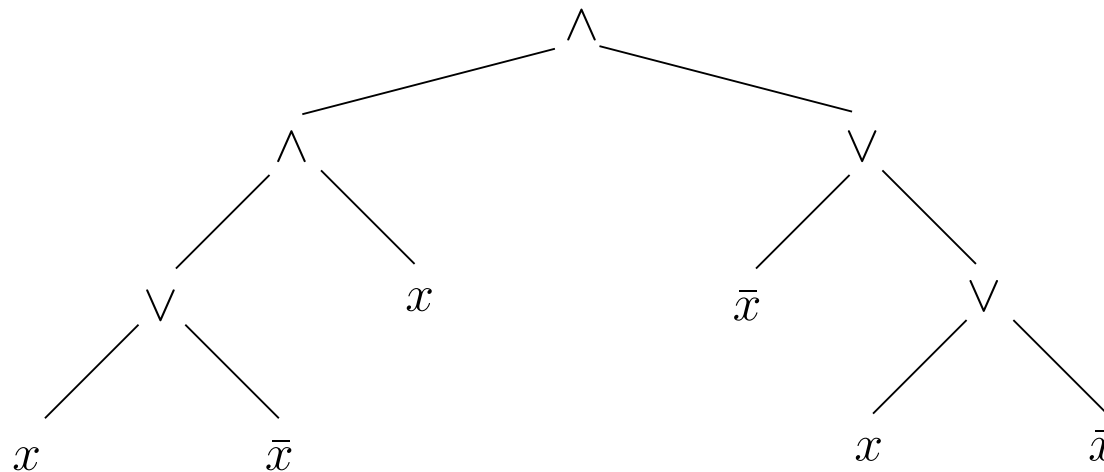
- Expression  $\sim$  labelled tree
- **Random** expression  $\sim$  **random** labelled tree
- What notion of randomness on trees?
  - Choose size  $m$  of the tree; assume all trees of same size are equiprob.  
Then let  $m \rightarrow +\infty$
  - Choose tree at random (e.g., by a branching process): size is also random. Then label tree at random.

# Boolean expressions and trees

- Expression  $\sim$  labelled tree
- Random expression  $\sim$  random labelled tree
- Two notions of randomness on trees/boolean expressions
- Each boolean expression computes a boolean function
- A boolean function is represented by an *infinite number of expressions*
- Can we use random boolean expressions to define a *probability distribution* on boolean functions?

## Former work : And/Or trees

- One of the most studied models for random boolean expressions
- Binary trees; no simple node
- Internal nodes are labelled by  $\vee$  or  $\wedge$
- Leaves are labelled by the literals:  $x_1, \dots, x_n, \bar{x}_1, \dots, \bar{x}_n$



# And/Or trees

- Paris et al. 94: first definition of a tree distribution on boolean functions
- Lefman and Savicky 97:
  - Proof of existence of a tree distribution (by pruning)
  - **Tree complexity** of  $f$ :  $L(f)$  = size of smallest tree that computes  $f$
  - $\frac{1}{4} \left(\frac{1}{8n}\right)^{L(f)} \leq P(f) \leq e^{-cL(f)/n^3} (1 + O(1/n))$
- Chauvin et al. 04: alternative definition of probability by generating functions; improvement on upper bound:  $P(f) \leq e^{-cL(f)/n^2} (1 + O(1/n))$
- For tautologies:
  - Woods 05: Asymptotic probability  $P(True) \sim 1/4n$  and probable shape of tautologies:  $l \vee \dots \vee \bar{l} \vee \dots$
  - Kozik 08: Alternative derivation of asymptotic probability and shape

# And/Or trees: probability and complexity

To sum up:

- definition of a tree-induced probability distribution on boolean functions
- probability of constant functions *True* and *False*: known
- probability of a non-constant function:
  - lower bound  $(1/4) (8n)^{-L(f)}$  (not that bad; order looks right)
  - upper bound  $e^{-cL(f)/n^2} (1 + O(1/n))$  (probably not tight)

# And/Or trees: probability and complexity

To sum up:

- definition of a tree-induced probability distribution on boolean functions
- probability of constant functions *True* and *False*: known
- probability of a non-constant function:
  - lower bound  $(1/4) (8n)^{-L(f)}$  (not that bad; order looks right)
  - upper bound  $e^{-cL(f)/n^2} (1 + O(1/n))$  (probably not tight)
- Partial results. Can we go further?

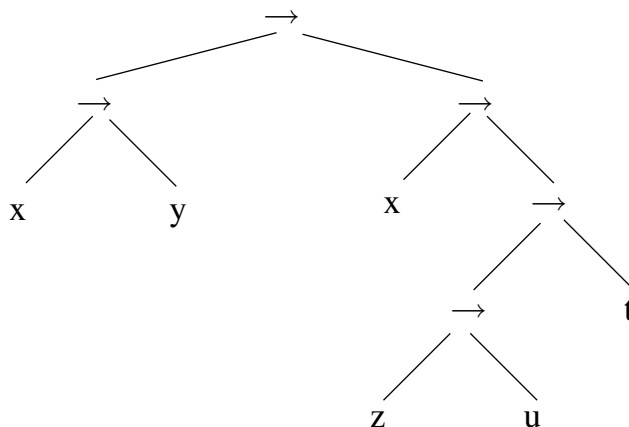
# And/Or trees: probability and complexity

To sum up:

- definition of a tree-induced probability distribution on boolean functions
- probability of constant functions *True* and *False*: known
- probability of a non-constant function:
  - lower bound  $(1/4) (8n)^{-L(f)}$  (not that bad; order looks right)
  - upper bound  $e^{-cL(f)/n^2} (1 + O(1/n))$  (probably not tight)
- Partial results. Can we go further?
- Consider a simpler system

# A restricted propositional calculus

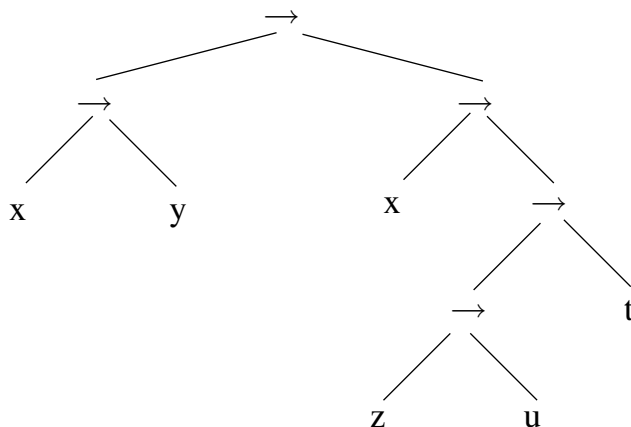
- Finite number of boolean variables :  $x_1, x_2, \dots, x_n$ ; **no** negative literals.
- A single connector  $\rightarrow$  ( $x_1 \rightarrow x_2$  is also  $\overline{x_1} \vee x_2$ ).
- Expressions are *binary trees*:  $(x \rightarrow y) \rightarrow (x \rightarrow (z \rightarrow u) \rightarrow t)$





# A restricted propositional calculus

- Finite number of boolean variables :  $x_1, x_2, \dots, x_n$ ; **no** negative literals.
- A single connector  $\rightarrow$  ( $x_1 \rightarrow x_2$  is also  $\overline{x_1} \vee x_2$ ).
- Expressions are *binary trees*:  $(x \rightarrow y) \rightarrow (x \rightarrow (z \rightarrow u) \rightarrow t)$



- An expression is a (possibly empty) sequence of expressions: **premises**, followed by a variable: **goal**.

# A restricted propositional calculus

- Finite number of boolean variables :  $x_1, x_2, \dots, x_n$ ; no negative literals.
- A single connector  $\rightarrow$
- “Simple” system: may hope for a detailed study of random expressions and boolean functions.
- Relevance to intuitionistic logic:

Tautology  $\sim$  proof of a goal from premises

# Boolean functions and expressions

An expression (a tree) computes a boolean function on  $k$  variables.

- What is the set of boolean functions that can be computed?

$$\Rightarrow \text{Post set } S_0 = \{x \vee g(x_1, \dots, x_k)\}$$

# Boolean functions and expressions

An expression (a tree) computes a boolean function on  $k$  variables.

- What is the set of boolean functions that can be computed?

$$\Rightarrow \text{Post set } S_0 = \{x \vee g(x_1, \dots, x_k)\}$$

- Many different expressions compute the same boolean function.

Probability that a “random” expression computes a specific function?

# Probability of a boolean function

- Informally, it is the ratio of trees that compute  $f$  to the total number of trees (assuming this ratio can be defined).
- Define the size of a formula (tree) as the number of variable occurrences (leaves).
- Define  $A_m = \{\text{trees of size } m\}$ ;  $A_m(f) = \{\text{trees in } A_m \text{ that compute } f\}$ .

Assume a **uniform distribution** on  $A_m$ .

- Probability that a tree of size  $m$  computes  $f$ :

$$P_m(f) = \frac{|A_m(f)|}{|A_m|}$$

- For any boolean function  $f$ ,  $\lim_{m \rightarrow +\infty} P_m(f)$  exists?

# Probability of a boolean function

Existence of a limit  $P(f) = \lim_{m \rightarrow +\infty} P_m(f)$ ?

- Enumerate trees by size: g.f.  $\Phi(z) = \sum_m |A_m| z^m = (1 - \sqrt{1 - 4nz})/2$
- Enumerate the set  $A(f)$  of trees computing a specific function  $f$ :

Generating function  $\phi_f(z)$ ?

Consider *all* boolean functions

$$A(f) = \cup_{g,h} (A(g), \rightarrow, A(h)) \Rightarrow \phi_f = \sum_{g,h} \phi_g \phi_h$$

$\Rightarrow$  write a system of algebraic equations for the enumerating functions

$\Rightarrow$  Drmota-Lalley-Woods theorem gives asymptotics of  $[z^m] \phi_f(z)$

- Putting all this together proves the existence of the prob. distribution  $P$

For any boolean function  $f$ , we compute

$$P(f) = \lim_{m \rightarrow +\infty} \frac{[z^m] \phi_f(z)}{[z^m] \Phi(z)}$$

# Probability of a boolean function

- We have proved the existence of  $P(f)$  for any  $f$   
( $f \notin S_0: P(f) = 0$ )
- *Can we compute explicitly the probability of a boolean function?*

# Probability of a boolean function

- We have proved the existence of  $P(f)$  for any  $f$   
( $f \notin S_0: P(f) = 0$ )
- *Can we compute explicitly the probability of a boolean function?*
- The complexity of a function  $f$  is the smallest size of a tree that computes  $f$ .
- *What is the relation between the complexity and the probability of a boolean function?*
- *What is the typical shape of a tree that computes a specific function?*
- *What is the average complexity of a random boolean function?*



# Tautologies

We begin with the simplest function: the constant *True*

- **Simple** tautology: a premise is equal to the goal.
- We know the probability of simple tautologies:

$$\frac{4n + 1}{(2n + 1)^2} \sim \frac{1}{n}$$

- Almost all tautologies are simple (Fournier et al. 07)
- Hence  $P(\textit{True}) \sim 1/n$
- Consequence: almost all tautologies in the system of implication and positive literals are intuitionistic tautologies.

# Probability of boolean functions

We know a.s. the shape of a random tautology.

We can compute the probability of *True*.

*Can we extend this to a non-constant boolean function  $f$ ?*

# Probability of boolean functions

- *True*:  $1/n + O(1/n^2)$
- Literal  $x$ :  $1/2n^2 + O(1/n^3)$
- Function  $x \rightarrow y$ :  $9/16n^3 + O(1/n^4)$
- For all  $f \in S_0 \setminus \{1\}$ :

$$P(f) = \frac{\lambda(f)}{4^{L(f)} n^{L(f)+1}} (1 + O(1/n))$$

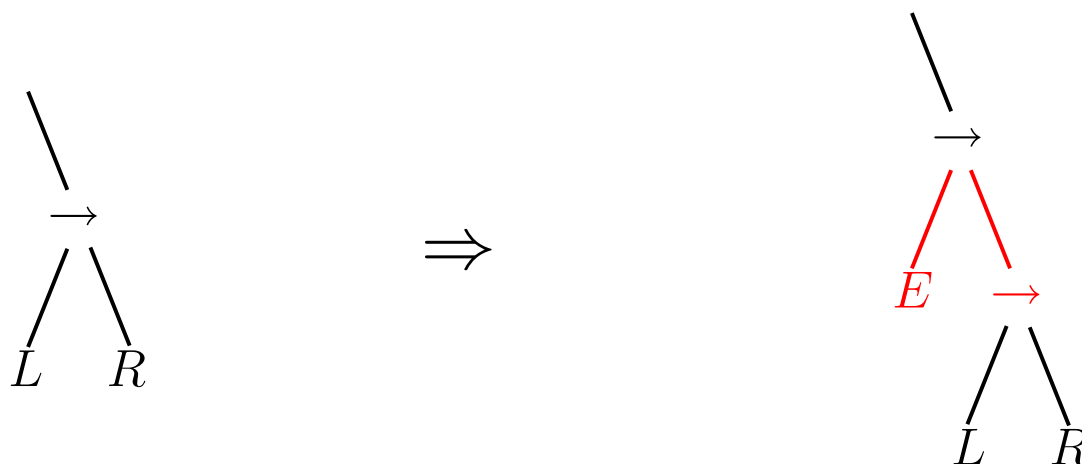
- $\lambda(f)$  is related to the minimal trees for  $f$
- The trees of  $A(f)$  are **simple**: a.s. obtained from a minimal tree by a **single** expansion

## Sketch of proof

- Start from the set of minimal trees that compute  $f$ .
- Define extension rules: we obtain a larger (infinite) set of trees, still computing  $f$ ; we can compute the probability of this set.
- Probability of this new set is related to the sizes of the initial trees, i.e. to the **tree complexity** of  $f$ .
- Do we obtain a.s. all the trees that compute  $f$ ?
- If so, we know the probability of  $f$ , and we can express it in terms of its complexity.

# Extensions of minimal trees

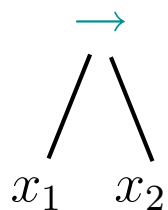
Consider a tree  $A$  that computes  $f$ , and a node of  $A$



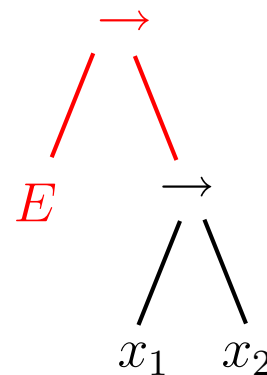
When can we expand a node of  $A$ , and still get a tree that computes  $f$ ??

# Extensions of minimal trees: example

$f = x_1 \rightarrow x_2$  has a unique minimal tree  $A_{min}$ :



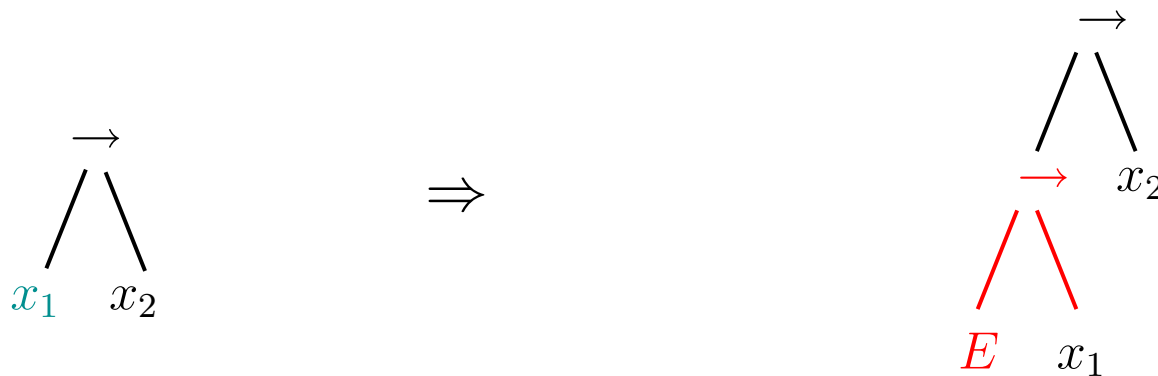
$\Rightarrow$



- $E$  is a tautology
- $E$  has goal  $x_1$
- $E$  has a premise  $x_2$

# Extensions of minimal trees: example

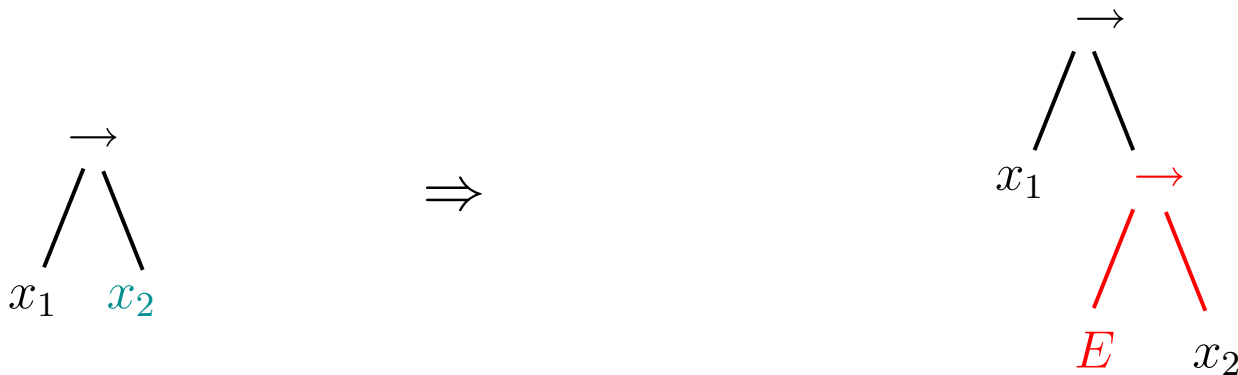
$f = x_1 \rightarrow x_2$  has a unique minimal tree  $A_{min}$ :



- $E$  is a tautology
- $E$  has goal  $x_2$
- $E$  has a premise  $x_1$

# Extensions of minimal trees: example

$f = x_1 \rightarrow x_2$  has a unique minimal tree  $A_{min}$ :



- $E$  is a tautology
- $E$  has goal  $x_1$
- $E$  has a premise  $x_2$



## Extensions of minimal trees: example

$f = x_1 \rightarrow x_2$  has a unique minimal tree  $A_{min}$

- Nine possible types of expansion  $\Rightarrow$  set  $\mathcal{E}(A_{min})$  of trees computing  $f$
- We can compute the probability of  $\mathcal{E}(A_{min})$ :

$$\frac{9}{16n^3} + O\left(\frac{1}{n^4}\right)$$

- This is the probability of  $f$

# Extensions of minimal trees

- Define extensions for minimal trees
- Compute probability of the set  $\mathcal{E}(f)$  obtained by one extension
- Compute probability of the set  $\mathcal{E}^+(f)$  obtained by a finite number of extensions
- Compute probability of  $A(f) \setminus \mathcal{E}^+(f)$ :
  - Define pruning rules: inverses of expansion rules
  - Any tree of  $A(f)$  can be pruned into an irreducible tree
  - $\{\text{Minimal trees}\} \subset \{\text{Irreducible trees}\}$
  - Almost all trees of  $f$  can be pruned into irreducible trees.

# Probability of a boolean function $f$

- Expression of the probability

$$P(f) = \frac{\lambda(f)}{4^{L(f)} n^{L(f)+1}} (1 + O(1/n))$$

- We obtain almost all the trees by a single expansion of a minimal tree

$$P(f) = \text{Proba}(\mathcal{E}(f)) (1 + o(1))$$

- The number of possible expansions is related to properties of minimal trees:

- $m$  = number of minimal trees for  $f$
- $e$  = number of essential variables of  $f$

Then

$$2(2m - 1)L(f) \leq \lambda(f) \leq (1 + 2e)(2L(f) - 1)m$$

## Possible extensions

- Computation of the constant factor  $\lambda(f)$ ?  
Done for read-once functions; for other functions?
- Result can be adapted when trees are obtained by a growing process
- What if we allow negative literals?
- What if we choose a different set of connectors?

# Average complexity of a boolean function

- For a uniform distribution on boolean functions, maximal and average tree complexity is  $2^k / \log k$  (Shannon, Lupanov...)
- What if the distribution is not uniform? for example, a tree distribution?
- We have computed the probability of a boolean function of *known* (hence, “fixed, small” and independent of  $k$ ) complexity.
- What about the probability of a function of “large” (dependent on  $k$ ) complexity?