# Counting $\lambda$-terms CLA 2008 - Kraków 

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## Outline of the talk

## Introduction, Problem Statement and First Results

Counting specific terms

Counting variables

In the Turing world.

Research directions

## Outline of the talk

## Introduction, Problem Statement and First Results

## Polynomials

BoundingCounting specific terms In the Turing world...


## The main polynomial

Number of $\lambda$-terms with at most $X$ variables
$P_{1}(X)=X$
$P_{n+1}(X)=P_{n}(X+1)+\sum_{l+r=n} P_{l}(X) P_{r}(X)$
This gives:

$$
\begin{array}{ll}
P_{2}(X)=1+X & P_{6}(X)=42+49 X+26 X^{2}+10 X^{3} \\
P_{3}(X)=2+X+X^{2} & P_{7}(X)=139+\ldots+5 X^{4} \\
P_{4}(X)=4+5 X+3 X^{2} & P_{8}(X)=506+\ldots+35 X^{4} \\
P_{5}(X)=13+17 X+6 X^{2}+2 X^{3}
\end{array}
$$

What are the coefficients of $P$ ?
$P_{n}(X)=\sum_{i=0}^{\left[\frac{n}{2}\right\rceil} c_{n, i} X^{i}$

## The coefficients of the main polynomial

## What are the coefficients of $P$ ? <br> $P_{n}(X)=\sum_{i=0}^{\left[\frac{n}{2}\right\rceil} c_{n, i} X^{i}$

Number of $\lambda$-terms with 1 variable occurring $i$-times
$c_{1,1}=1, c_{1, k}=0(k \neq 1)$
$c_{n+1, k}=\sum_{i=0}^{+\infty}\binom{k+i}{i} c_{n, k+i}+\sum_{l+r=n} \quad c_{l, i} c_{r, j}$

$$
i+j=k
$$

## The derivatives of the main polynomial

Derivation
$\frac{1}{k!} P_{n}^{(k)}(X)=\frac{1}{k!} \sum_{i=k}^{\left[\frac{n}{2}\right]-k} c_{n, i} \frac{i!}{(i-k)!} X^{i-k}=\sum_{i=k}^{\left[\frac{n}{2}\right]-k} c_{n, i}\binom{i}{k} X^{i-k}$
Therefore
$\frac{1}{k!} P_{n}^{(k)}(X)$ :
Number of $\lambda$-terms with at most $X$ variables +1 variable occurring $k$-times
$P_{n}(X+1)=\sum_{i=0}^{+\infty} \frac{1}{k!} P_{n}^{(k)}(X)$

## Term with exactly $X$ variables

Formula
$Q_{n}(0)=P_{n}(0), Q_{n}(X+1)=P_{n}(X+1)-P_{n}(X)$
$Q_{1}(1)=1, Q_{1}(X)=0(X \neq 1)$
$Q_{n+1}(X)=Q_{n}(X)+Q_{n}(X+1)$
$\quad+\sum_{\substack{1+r=n \\ X_{0}+X_{1}+X_{2}=X^{\prime}}}^{X_{0}!X_{!}!X_{2}!} Q_{l}\left(X_{0}+X_{1}\right) Q_{r}\left(X_{0}+X_{2}\right)$
Useful to count some classes of terms
Example: affine terms

$$
\begin{aligned}
& A_{1}(1)=1, A_{1}(X)=0(X \neq 1) \\
& A_{n+1}(X)=A_{n}(X)+A_{n}(X+1) \\
& \quad+\sum_{1+r=n} \quad \frac{X!}{X_{1}+X_{2}=x} A_{1}\left(X_{1}\right) A_{r}\left(X_{2}\right)
\end{aligned}
$$

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PolynomialsBounding
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In the Turing world...
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Introduction, Problem Statement and

## Bounding

First result
$\forall \alpha<1$ we have $f, g$ exponential, such that:

$$
f\left(n-n^{\alpha}\right)\left(n^{\alpha}+X-1\right)^{\frac{n-n^{\alpha}}{2}} \leq P_{n}(X) \leq g(n)(n+X)^{\left\lceil\frac{n}{2}\right\rceil}
$$

The asymptotic development of $P_{n}(0)$

$$
P_{n}(0) \sim n^{\left\lceil\frac{n}{2}\right\rceil} \ldots
$$

## Better upper bound

Size zero for variables
$P_{0}(X)=X$ instead of $P_{1}(X)=X$

$$
P_{n}(X) \leq C(n)(n+X)^{n+1}
$$

The dominant coefficient is exact.

## Conjecture for size one variables

$$
\begin{gathered}
P_{2 n}(X) \leq(2 n-1) C(n-1)(2 n-1+X)^{n} \\
P_{2 n+1}(X) \leq C(n)(2 n+X)^{n}
\end{gathered}
$$

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## Terms avoiding a given term as subterm



## Terms avoiding a given pattern as subterm



## Normalisable terms



## Other results by Jue Wang (Boston) 2004

- Typable terms converges toward probability 0
- Ratio of application nodes / lambda nodes



## Outline of the talk



Counting variables
In the Turing worid．．． Research directions

## General setting

Recurrence formula
$P_{0}(X)=0$
$P_{n+1}(X)=V_{n+1}(X)+P_{n}(X+1)+\sum_{i+j=n} P_{i}(X) P_{j}(X)$

- $n$ : size
- $X$ : number of available variables
- $V_{n}(X)$ : the number of terms which are variables


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Is it realistic?

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## Until now

unit size: $V_{1}(X)=X$ and $V_{n}(X)=0$ if $n>1$

## Is it realistic? Well...

> 1 bit to represent many possible choices

- a super-exponential number of terms of a given "size"


## Non-unit cost of variables

A concrete notion of size

- use De Bruijn indexes
- take into account the cost of such indexes


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A concrete notion of size

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- take into account the cost of such indexes

Possible choices

- unary De Bruijn index:

$$
V_{n}(X)= \begin{cases}1 & \text { if } n \leq X \\ 0 & \text { else }\end{cases}
$$

- binary De Bruijn index:

$$
V_{n}(X)=\#\{i: i \leq X \text { and }|\operatorname{bin}(i)|=n\}
$$

## Small closed terms in the 3 models

| size | constant |  | unary | binary |  |
| :---: | :---: | :---: | :---: | :---: | :---: |
| 2 | $\lambda x$. $x$ |  | $\lambda .1$ | $\lambda .0$ |  |
| 3 | $\lambda x_{1}, x_{2}, x_{i}$ | ( $\times 2$ ) | $\lambda^{2} .1$ | $\begin{aligned} & \lambda^{2} .0 \\ & \lambda^{2} .1 \end{aligned}$ |  |
| 4 | $\begin{gathered} \lambda x_{1}, x_{2}, x_{3}, x_{i} \\ \lambda x . x x \end{gathered}$ | $(\times 3)$ | $\begin{gathered} \lambda^{3} \cdot 1 \\ \lambda^{2} .11 \\ \lambda .11 \end{gathered}$ | $\begin{gathered} \lambda^{3} \cdot 0 \\ \lambda^{3} \cdot 1 \\ \lambda .00 \end{gathered}$ |  |
| 5 | $\begin{gathered} \lambda x_{1}, x_{2}, x_{3}, x_{4} \cdot x_{i} \\ \lambda x_{1}, x_{2} \cdot x_{i} x_{j} \\ \lambda x_{1}, x_{1}\left(\lambda x_{2} \cdot x_{i}\right) \\ (\lambda x \cdot x)(\lambda x, x) \end{gathered}$ | $\begin{aligned} & (\times 4) \\ & (\times 4) \\ & (\times 4) \end{aligned}$ | $\lambda .11$ $\lambda^{4} .1$ $\lambda^{3} .11$ $\lambda^{2} .11$ $\lambda .(1 \lambda .1)$ $\lambda .(\lambda .11)$ $(\lambda .1)(\lambda .1)$ | $\begin{gathered} \lambda^{4} \cdot(0 / 1) \\ \lambda^{3} \cdot 11 \\ \lambda^{2} \cdot(0 / 1)(0 / 1) \\ \lambda .0(\lambda .(0 / 1)) \\ (\lambda .0)(\lambda .0) \end{gathered}$ | $\begin{aligned} & (\times 2) \\ & (\times 4) \\ & (\times 4) \end{aligned}$ |

## Differences between the 3 models

## For non-unit costs

- the growth rate is only exponential
- $P_{n}(X)$ is not a polynomial in $X$



## Proportion of terms in normal form



## Proportion of terms starting with a lambda



Proportion of terms starting with $[\sqrt{\sqrt{n}}]$ lambdas


## Proportion of terms not containing identity

Remark: id has the same size for the 3 models


## Proportion of terms not containing $\Omega$

Remark: $\Omega$ has the same size for the 3 models

$L_{\text {In }}$ the Turing world...

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## Analogous questions in the Turing world

- syntactical objects: fully specified Turing machines

$$
\delta:\{0,1\} \times Q_{n} \rightarrow\{0,1\} \times\left(Q_{n} \cup\{\text { halt }\}\right) \times\{L, R\}
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- studied property: the halting problem (H)


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## Simple facts

1. there are $(4 n+4)^{2 n}$ machines with $n$ states
2. $d(H)<1-1 / e^{2}$ (if it exists)

- determining $d(H)$ in the general setting is still open, but...


## Result of another kind...

## Theorem (J.D. Hamkins, A. Miasnikov, 2005)

There exists a set B of machines such that:

1. $B$ is decidable (in polynomial time);
2. B has asymptotic density 1 ;
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it uses semi-infinite tape Turing-machines!
if falling-off the tape is not halting then $d(H)=0$

## Proof idea

Uses a trick to convert the first steps of a random Turing machine into a random walk of the head.

## Yet another kind of result...

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## Proof idea

The set of machines computing a given (computable) function is not strongly negligible (one can add garbage code at some places without affecting the computation).

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## Research directions

- "combinatorial" proof that for any fixed tree $T_{0}$, a large random tree always contains $T_{0}$ (Kolmogorov incompressibility method?)
- typical tree shape of random $\lambda$-terms (e.g. relation between height and size)
- other experiments: random generation of big $\lambda$-terms to have an idea of their shape


## Research directions

- possible properties to look at:
- about shape: normality, beginning by a $\lambda$,
- depending on the reduction rule: SN, WN,
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- finding sets of terms with density 1 for which SN or WN becomes decidable
- consider other distributions than the uniform one (e.g., similar to the BST model)
- any non-trivial $\quad$ result about random $\lambda$-terms with any representation of variables is welcome!

