On the number of $\beta$-redices in random closed linear $\lambda$-terms


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What is the $\lambda$-calculus?

2 A

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feeding an argument $t$ to a function $s$
- We're interested in terms up to $\alpha$-equivalence:

$$
(\lambda x . x x)(\lambda x . x x) \stackrel{\alpha}{=}(\lambda y . y y)(\lambda x . x x) \stackrel{\alpha}{\neq}(\lambda y . y a)(\lambda x . x x)
$$

## Computing with the $\lambda$-calculus

- Substitution rule:

$$
\mathrm{T}_{1}\left[v:=\mathrm{T}_{2}\right]
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"replace free occurences of $v$ in $T_{1}$ with $T_{2}$ "
( $\alpha$-converting $T$ if necessary, to avoid capturing variables of $T_{2}$ )

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- Examples of substitutions
- $(\lambda x .(x y))[y:=x] \neq(\lambda x .(x x))$
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- Dynamics of the $\lambda$-calculus: $\beta$-reductions
( $\lambda$-terms together with $\beta$-reduction are enough to encode any computation!)

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\left(\left(\lambda x \cdot t_{1}\right) t_{2}\right) \xrightarrow{\beta} t_{1}\left[x:=t_{2}\right]
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More on $\beta$-reductions

- An occurence of the $\left(\left(\lambda x . t_{1}\right) t_{2}\right)$ "pattern" is called a $\beta$-redex:

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((\lambda x .((\lambda y \cdot(y x)) x))(a b))
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normal form!


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- Terms may never reach a normal form, their size might even increase! $((\lambda x .(x x))(\lambda x .(x \times x))) \xrightarrow{\beta}(\lambda x .(x x x))(\lambda x .(x x x))(\lambda x .(x x x))$


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- Order in which redices are reduced matters!

$$
\begin{gathered}
(\lambda x . z)((\lambda x .(x x))(\lambda x .(x x))) \longrightarrow(\lambda x . z)((x x)[x:=(\lambda x .(x x))])=\ldots \\
\longrightarrow z[x:=(\lambda x . x x)(\lambda x . x x)]=z \quad 4 \mathrm{G}
\end{gathered}
$$

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Previous work on the reduction of $\lambda$-terms

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For terms expressed in the previously-presented syntax and size defined recursively as:

$$
|x|=0,|(a b)|=1+|a|+|b|,|\lambda x . t|=1+|t|
$$

- Asymptotically almost no $\lambda$-term is strongly normalizing. [DGKRTZ13,BGLZ16]
For terms expressed using de Bruijn indices or combinators (together with appropriate size functions)

Parameter sensitive to the definition of the syntax and the size of terms!

- Almost every simply-typed $\lambda$-term has a long $\beta$-reduction sequence [SAKT17]

Subfamilies of $\lambda$-terms
General terms: no restrictions on variable use

$$
\lambda x . \lambda y . x\left(\begin{array}{ll}
y & a
\end{array} \quad \lambda x . \lambda y . x \quad(\lambda x . x x)(\lambda y . y y)\right.
$$

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General terms: no restrictions on variable use

| $\lambda x . \lambda y . x(y)$ | $\lambda x . \lambda y . x$ |
| :---: | :---: |
| free variable |  |

## Subfamilies of $\lambda$-terms

General terms: no restrictions on variable use


Affine Terms: bound variables occur at most once $(\lambda x . \lambda y . a)(\lambda x . x)$ $\lambda x . \lambda y . y$

Linear Terms: bound variables occur exactly once

$$
\lambda x \cdot \lambda y \cdot(y x) a \quad \lambda x \cdot \lambda y \cdot\left(\begin{array}{ll}
y & a)(b x)
\end{array}\right.
$$

$\lambda x . a(\lambda z \cdot(\lambda y . y(x z)))$
$\beta$-reducing closed linear terms

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A lower bound is given by the number of $\beta$-redices!
This motivates the central question of this work:
What is the number of $\beta$-redices in a random linear $\lambda$-term?
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random variable!
uniform distribution

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 seq. of random variables!
uniform distribution on the set of terms of size $n$

What are maps?


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We're interested in unrestricted genus, restricted vertex degrees

Why should you, a logician, be interested in maps?
String diagrams! [BGJ13, Z16]


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Closed linear terms $\leftrightarrow$ trivalent maps Closed affine terms $\leftrightarrow(2,3)$-valent maps Established in [BGJ13, BGGJ13]

Why should you, a combinatorialist, be interested in $\lambda$-terms?
Decomposing (closed) rooted trivalent maps [BGJ13]

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Why should you, a combinatorialist, be interested in $\lambda$-terms?
Decomposing (closed) rooted trivalent maps [BGJ13] and closed linear terms!



Some of our previous results: limit distributions


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$\#$ loops $=\#$ id-subterms


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$\beta$-reduction as map rewriting

$\lambda x .(\lambda y . y(\lambda z . \lambda w . z w))((\lambda u . u) x)$
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$$
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$\beta$-reduction as map rewriting


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\left[z^{n}\right](A \cdot B)=\sum_{k=n_{0}}^{n} a_{k} b_{n-k}
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## crucial ingredient: coefficients are growing rapidly

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- Tracking redices during the decomposition

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no redex
b

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Mean number of $\beta$-redices in closed terms

- Tracking redices during the decomposition Abstractions, subcase 1.1


Mean number of $\beta$-redices in closed terms

- Tracking redices during the decomposition Abstractions, subcase 1.2

\#ways to do this

number of abstractions in $t$

Mean number of $\beta$-redices in closed terms

- Tracking redices during the decomposition Abstractions, subcase 1.3

\#ways to do this

number of subterms in $t=$ size of $t$


Mean number of $\beta$-redices in closed terms

- Building the specification of the OGF
- $|t|_{\lambda}=\frac{|t|+1}{3},|t|-|t|_{\lambda}=\frac{2|t|-1}{3}$
- $r \partial_{r} T_{0}=\sum_{t \in T_{0}}|t|_{\beta} z^{|t|} r^{|t|_{\beta}}$
- $\frac{z \partial_{z} \mathrm{~T}_{0}+\mathrm{T}_{0}}{3}=\sum_{\mathrm{t} \in \mathrm{T}_{0}} \frac{|\mathrm{t}|+1}{3} z^{|t|} v^{|t|_{\beta}}$
$\bullet \frac{2 z \partial_{z} \mathrm{~T}_{0}-\mathrm{T}_{0}}{3}=\sum_{t \in \mathrm{~T}_{0}} \frac{2|\mathrm{t}|-1}{3} z^{|t|} v^{|t|_{\beta}}$

Mean number of $\beta$-redices in closed terms
-Translating to a differential equation and pumping

$$
\begin{aligned}
\mathrm{T}_{0} & =-z\left(z^{2}(r+1)(1+(r-1) z T)(r-1) \partial_{\mathrm{r}} \mathrm{~T}_{0}\right. \\
& \left.-\frac{(1+z(r-1) \mathrm{T}) z^{3}(r+5) \partial_{z} \mathrm{~T}_{0}}{3}-\frac{z^{3}(r-1)^{2} \mathrm{~T}_{0}^{2}}{3}-\frac{4 z^{2}(r-1) \mathrm{T}_{0}}{3}-z-\mathrm{T}_{0}^{2}\right)
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\end{aligned}
$$

A plot of the dist. of redices for terms/maps of size $n=119$


On the number of $\beta$-redices in random closed linear $\lambda$-terms - Bodini, Singh, Zeilberger

## Whats next?

## Whats next?

- Tracking the three patterns whose reduction alters the number of redices (WIP with Bodini, Zeilberger, Wallner, Gittenberger)

$$
\begin{gathered}
(\lambda x \cdot C[(x u)])\left(\lambda y \cdot t_{2}\right) \quad\left(\left(\lambda x . \lambda y \cdot t_{1}\right) t_{2}\right) t_{3} \\
(\lambda x \cdot x)\left(\lambda y \cdot t_{1}\right) t_{2}
\end{gathered}
$$

## Whats next?

- Tracking the three patterns whose reduction alters the number of redices (WIP with Bodini, Zeilberger, Wallner, Gittenberger)

$$
\begin{gathered}
(\lambda x \cdot C[(x u)])\left(\lambda y \cdot t_{2}\right) \quad\left(\left(\lambda x . \lambda y \cdot t_{1}\right) t_{2}\right) t_{3} \\
(\lambda x \cdot x)\left(\lambda y \cdot t_{1}\right) t_{2}
\end{gathered}
$$

- More parameters:


Mean path length


Profile

## Whats next?

- Tracking the three patterns whose reduction alters the number of redices (WIP with Bodini, Zeilberger, Wallner, Gittenberger)

$$
\begin{gathered}
(\lambda x \cdot C[(x u)])\left(\lambda y \cdot t_{2}\right) \quad\left(\left(\lambda x . \lambda y \cdot t_{1}\right) t_{2}\right) t_{3} \\
(\lambda x \cdot x)\left(\lambda y \cdot t_{1}\right) t_{2}
\end{gathered}
$$

- More parameters:


Thank you!

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