

# Bounding normalization time through intersection types

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## Intersection types and complexity

- **I.T. with idempotence**: used for semantic purposes, i.e. for building filter models for  $\lambda$ -calculus  $\rightsquigarrow$  modulo equivalence  $\sigma \wedge \sigma = \sigma$
- **I.T. without idempotence**: recently used to reason about **complexity** of  $\beta$ -reductions through type assignment systems for  $\lambda$ -calculus
  - Terui (2006): any normalizing  $\beta$ -reduction sequence for  $M$  is bounded by the size of the type derivation for  $M$
  - Lengrand (2011): the measure of a principal type derivation for  $M$  corresponds to the maximal length of a normalizing  $\beta$ -reduction sequence for  $M$

## Our aim

- Using I.T. **without idempotence nor associativity** to express the functional dependence of the length of a normalizing  $\beta$ -reduction sequence for  $M$  on the size of  $M$  itself
- We take inspiration from the system STA (Gabori-RDR 2007) characterizing PTIME
  - STA gives a bound on the number of normalization steps from  $M$  to its normal form, in the form  $|M|^{d+1}$
  - $d$  is a measure depending on the type derivation for it (the *depth*)
  - Since for every normalizing term there is a type derivation with minimal depth, this bound depends on  $M$

## System STI (1)

- **Soft Type** assigning system with **Intersection**
- The system assigns types to strongly normalizing  $\lambda$ -terms
- Intersection with commutativity but **without idempotence nor associativity**

### Our aim

Proving a normalization bound on any reduction sequence for a given (strongly normalizing) term

## System STI (2)

- $\lambda$ -terms:  $M ::= x \mid MM \mid \lambda x.M$
- $\beta$ -reduction:  $(\lambda x.M)N \rightarrow M[N/x]$
- STI types:

$A ::= a \mid \sigma \rightarrow A$  (linear types)

$\sigma ::= A \mid \underbrace{\sigma \wedge \dots \wedge \sigma}_n \quad (n > 1)$  (intersection types)

- Context: finite set of assumptions  $x : \sigma$ 
  - $\Gamma \wedge \Delta = \{x : \sigma \wedge \tau \mid x : \sigma \in \Gamma, x : \tau \in \Delta\}$
  - $\Gamma, \Delta$  if  $\text{dom}(\Gamma) \cap \text{dom}(\Delta) = \emptyset$

- $\Pi \triangleright \Gamma \vdash M : \sigma$  denotes a derivation  $\Pi$  with conclusion  $\Gamma \vdash M : \sigma$ .

## STI rules

$$\frac{}{x : A \vdash x : A} \text{ (Ax)} \quad \frac{\Gamma \vdash M : \sigma \quad x \notin \text{dom}(\Gamma)}{\Gamma, x : A \vdash M : \sigma} \text{ (w)}$$

$$\frac{\Gamma, x : \sigma \vdash M : A}{\Gamma \vdash \lambda x.M : \sigma \rightarrow A} \text{ (}\rightarrow I\text{)} \quad \frac{\Gamma \vdash M : \sigma \rightarrow A \quad \Delta \vdash N : \sigma \quad \Gamma \# \Delta}{\Gamma, \Delta \vdash MN : A} \text{ (}\rightarrow E\text{)}$$

$$\frac{\Gamma_1 \vdash M : \sigma_1 \quad \dots \quad \Gamma_n \vdash M : \sigma_n \quad n > 1}{\Gamma_1 \wedge \dots \wedge \Gamma_n \vdash M : \sigma_1 \wedge \dots \wedge \sigma_n} \text{ (}\wedge_n\text{)}$$

$$\frac{\Gamma, \mathbf{x}_1 : \sigma_1, \dots, \mathbf{x}_n : \sigma_n \vdash M : \tau}{\Gamma, \mathbf{x} : \sigma_1 \wedge \dots \wedge \sigma_n \vdash M[\mathbf{x}/\mathbf{x}_1, \dots, \mathbf{x}/\mathbf{x}_n] : \tau} \text{ (m)}$$

## Observations

- Terms are built in a linear form
- The  $(m)$  rule controls the number of contractions, so the growth of the reduction time (SLL- Lafont)
- Contraction on the right side of a derivation is handled by rule  $(\wedge_n)$ , which is parametric in  $n$

Also, we can distinguish between

- **constructive rules:**  $(Ax)$ ,  $(\rightarrow I)$ ,  $(\rightarrow E)$
- **non-constructive rules:**  $(w)$ ,  $(\wedge_n)$ ,  $(m)$

## Strong normalization

STI is sound and complete w.r.t. **strong normalization**:

- if a term is typable in STI then it is strongly normalizing (**soundness**)

*Proof.* Termination of reduction does not depend on the strategy

- all strongly normalizing terms can be typed in STI (**completeness**)

*Proof.* Following the guidelines of Neergaard (2005):

- 1 Longest reduction strategy
- 2 Subject expansion under perpetual strategy
- 3 Completeness follows as a corollary



## Measures (1)

### Size $|\Pi|$

- if the last rule of  $\Pi$  is  $(Ax)$ , then  $|\Pi| = 1$
- if the last rule of  $\Pi$  is a rule with  $n$  premises  $\Pi_i$ , then
$$|\Pi| = \left(\sum_{i=1}^n |\Pi_i|\right) + 1$$

### Size $|M|$

- $|x| = 1$
- $|\lambda x.M| = |M| + 1$
- $|MN| = |M| + |N| + 1$

## Measures (2)

$$\frac{\Gamma, \mathbf{x}_1 : \sigma_1, \dots, \mathbf{x}_n : \sigma_n \vdash \mathbf{M} : \tau}{\Gamma, \mathbf{x} : \sigma_1 \wedge \dots \wedge \sigma_n \vdash \mathbf{M}[\mathbf{x}/\mathbf{x}_1, \dots, \mathbf{x}/\mathbf{x}_n] : \tau} \quad (m)$$

Rank  $\mathbf{rk}(\Pi)$  = max between 1 and rank of any  $(m)$  rule in  $\Pi$

- Rank of  $(m)$  =  $k \leq n$  variables  $\mathbf{x}_i$  s.t.  $\mathbf{x}_i \in \text{FV}(\mathbf{M})$

Degree  $\mathbf{d}(\Pi)$  = max nesting of  $(\wedge_n)$  rules in  $\Pi$

Weight  $\mathbf{W}(\Pi, r)$  of  $\Pi$  w.r.t.  $r$

- $(\wedge_n)$  is the last applied rule  $\Rightarrow \mathbf{W}(\Pi, r) = r \cdot \max_{i=1}^n \mathbf{W}(\Sigma_i, r)$
- $(w)$  or  $(m)$  is the last applied rule  $\Rightarrow \mathbf{W}(\Pi, r) = \mathbf{W}(\Sigma, r)$

## Intersection tree

- $M$  is an **instance** of  $N$  if it is obtained from  $N$  by renaming a subset of its free variables
- Let  $(\delta)$  be a sequence of applications of rules  $(w)$  and  $(m)$
- **intersection tree**: maximal (sub)proof of the shape defined inductively as
  - If the last rule of  $\Sigma$  is a constructive rule, then

$$\frac{\Sigma}{\Gamma \vdash M : \sigma} (\delta)$$

is an **empty intersection tree** with one leaf  $\Sigma$

- If  $\Sigma_i$  is a (possibly empty) intersection tree ( $1 \leq i \leq n$ ), then

$$\frac{\frac{\Sigma_i \triangleright \Gamma_i \vdash M : \sigma_i \quad (1 \leq i \leq n)}{\Gamma_1 \wedge \dots \wedge \Gamma_n \vdash M : \sigma_1 \wedge \dots \wedge \sigma_n} (\wedge_n)}{\Gamma' \vdash M' : \sigma} (\delta)$$

is an **intersection tree** and its leaves are the leaves of all the  $\Sigma_i$

## Weighted subject reduction

### Weighted substitution

Let  $\Pi \triangleright \Gamma, x : \sigma \vdash M : \tau$  and  $\Sigma \triangleright \Delta \vdash N : \sigma$ , with  $\Gamma \# \Delta$  and  $x \notin \text{dom}(\Delta)$ , then  $S(\Sigma, \Pi) \triangleright \Gamma, \Delta \vdash M[N/x] : \tau$  and  $W(S(\Sigma, \Pi), r) \leq W(\Pi, r) + W(\Sigma, r)$ , for every  $r \geq \max\{\text{rk}(\Pi), \text{rk}(\Sigma)\}$

### Weighted subject reduction

$\Pi \triangleright \Gamma \vdash M : \sigma$  and  $M \xrightarrow{\beta} M'$  imply there is  $\Pi' \triangleright \Gamma \vdash M' : \sigma$  s.t.  $W(\Pi', r) < W(\Pi, r)$  for every  $r \geq \text{rk}(\Pi)$

Here we must take into account that a  $\beta$ -redex (in the term) may appear in **finitely many** copies in the type derivation  $\rightsquigarrow$  **parallel reductions**

## Example (1)

We want to reduce  $(\lambda x.xx)((\lambda y.y)z)$  to normal form

$$\begin{array}{c}
 \frac{}{x_1 : A \vdash x_1 : A} (Ax) \quad \frac{}{x_2 : a \vdash x_2 : a} (Ax) \\
 \hline
 \frac{}{x_1 : A, x_2 : a \vdash x_1 x_2 : a} (\rightarrow E) \\
 \frac{x_1 : A, x_2 : a \vdash x_1 x_2 : a}{x : A \wedge a \vdash xx : a} (m) \\
 \frac{}{\vdash \lambda x.xx : (A \wedge a) \rightarrow a} (\rightarrow I) \quad \frac{\Sigma_1 \triangleright z : A \vdash (\lambda y.y)z : A \quad \Sigma_2 \triangleright z : a \vdash (\lambda y.y)z : a}{\Sigma \triangleright z : A \wedge a \vdash (\lambda y.y)z : A \wedge a} (\wedge_2) \\
 \hline
 \frac{}{z : A \wedge a \vdash (\lambda x.xx)((\lambda y.y)z) : a} (\rightarrow E)
 \end{array}$$

$\Sigma$  ends by a non empty intersection tree  $\rightsquigarrow$  there are two “virtual” copies of the same redex

$$\frac{\frac{}{y : A \vdash y : A} (Ax)}{\vdash \lambda y.y : A \rightarrow A} (\rightarrow I) \quad \frac{}{z : A \vdash z : A} (Ax)}{\Sigma_1 \triangleright z : A \vdash (\lambda y.y)z : A} (\rightarrow E) \quad \text{and} \quad \frac{\frac{}{y : a \vdash y : a} (Ax)}{\vdash \lambda y.y : a \rightarrow a} (\rightarrow I) \quad \frac{}{z : a \vdash z : a} (Ax)}{\Sigma_2 \triangleright z : a \vdash (\lambda y.y)z : a} (\rightarrow E)$$

## Example (2)

If we reduce  $(\lambda y.y)z$  we get the following derivation

$$\frac{\frac{\frac{\frac{}{x_1 : A \vdash x_1 : A} (Ax)}{x_1 : A, x_2 : a \vdash x_1 x_2 : a} (m)}{\frac{}{x : A \wedge a \vdash xx : a} (\rightarrow I)}{\vdash \lambda x.xx : (A \wedge a) \rightarrow a} (\rightarrow E)}{\frac{\frac{\frac{}{z : A \vdash z : A} (Ax)}{z : A \wedge a \vdash z : A \wedge a} (\wedge_2)}{\frac{}{z : A \wedge a \vdash z : a} (\rightarrow E)}{\Pi \triangleright z : A \wedge a \vdash (\lambda x.xx)z : a} (\rightarrow E)}$$

where both the redexes of  $\Sigma_1$  and  $\Sigma_2$  have been reduced

## Measure of reduction

If  $\Pi \triangleright \Gamma \vdash M : \sigma$  and  $M \xrightarrow[\beta]{n} M'$ , then

i)  $n < |M|^{d(\Pi)+1}$

ii)  $|M'| < |M|^{d(\Pi)+1}$

Observe that

- The exponent of the function depends on the **term**
- The bound on the normalization procedure can easily be **exponential**  $\rightsquigarrow$  no bound on the complexity!
- The proof given above is independent on a given reduction strategy  $\rightsquigarrow$  **strong normalization soundness**