

# From normal functors to logarithmic space queries

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NGUYỄN Lê Thành Dũng (LIPN, U. Paris 13) — [nltd@nguyentito.eu](mailto:nltd@nguyentito.eu)  
j.w.w. Pierre PRADIC (LIP, ENS Lyon & MIMUW, U. Warsaw)  
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- categorical semantics (*normal functors*)
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- We use *linear logic* to do *implicit computational complexity*, i.e. machine-free characterization of complexity classes (ICC vs descriptive complexity  $\approx$  functional vs declarative)
- Semantics is involved in our upper bounds proofs

Roughly, *linearity*  $\approx$  no-copy / single use restriction

# Linear logic as a logic

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- Linear logic is constructive:  
proof of  $A \oplus B \cong$  proof of  $A$  or proof of  $B$
- $\implies$  Can be seen as a *functional programming language*,  
via the *Curry–Howard correspondence*

# Proofs as programs

## The Curry–Howard correspondence

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theoretical basis for OCaml, Haskell, etc.

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normalization	$\leftrightarrow$	computation

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# Linear logic as a programming language

One motivation for inventing linear logic:  
analyze the computational content (i.e. normalization) of  
intuitionistic proofs.

## Idea

Duplication is responsible for the complexity of computation,  
so make it explicit using  $!(-)$ .

$$\begin{aligned} \text{non-linear function} &\cong \text{linear function} + \text{duplicable input} \\ A \rightarrow B &\cong !A \multimap B \end{aligned}$$

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## Idea 2

Restricting  $!(-)$  can induce complexity effects  
→ Starting point for implicit complexity in linear logic

## Implicit computational complexity in linear logic (1)

In linear logic,  $!(-)$  satisfies not just  $!A \multimap !A \otimes !A$ , but also:

$$!A \multimap A \quad !A \multimap !!A \quad (\text{i.e. } !(-) \text{ is a comonad})$$

In *Elementary Linear Logic* (ELL), these rules are *removed*; ensures that  $!$ -nesting is somehow an invariant of computation. (LL vs ELL  $\approx$  modal logic S4 vs K)

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## Theorem (Girard 1998 / Danos and Joinet 2003)

*The class of languages decided by programs of type  $!\text{Str} \multimap !^k \text{Bool}$  in second-order ELL (for varying  $k \in \mathbb{N}$ ) is ELEMENTARY.*

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This holds for any “reasonable” encoding  $Str$  of strings.

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characterize smaller classes (P, EXPTIME...) in variants of ELL by taking fixed values of  $k \in \mathbb{N}$ .

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What happens for fixed  $k$  in ELL itself? Some issues arise:

- Sensitivity to *input representation*
- Previous proof techniques not sufficient

## ICC in LL meets finite models

Our paper: ELL-based implicit complexity

+ inspiration from Hillebrand's PhD thesis (1994),  
*Finite Model Theory in the Simply Typed  $\lambda$ -Calculus*

We define a type  $\text{Inp}_\Sigma$  of inputs as *finite structures*, and look at queries expressed by programs of type  $\text{Inp}_\Sigma \multimap \text{!}!\text{Bool}$ .

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(unless we add some ad-hoc restriction)

# Our results

## Theorem

*The class of queries over finite structures (with linear orders) computed by 2nd-order ELL proofs/programs of type  $\text{Inp}_\Sigma \multimap \text{!!Bool}$  lies between L and NL.*

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NL: non-deterministic logarithmic space

UL: *unambiguous* NL (unique accepting runs)

$\text{NL}^{\text{NL}} = \text{NL}$  (since  $\text{NL} = \text{coNL}$ ) so  $L^{\text{UL}} \subseteq \text{NL}$

(It is conjectured that  $\text{UL} = \text{NL}$ , and known that  $\text{UL}/\text{poly} = \text{NL}/\text{poly}$ .)

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## Lower bound via descriptive complexity

Recall that over linearly ordered finite structures:

**Theorem (Immerman 1983)**

$L = \text{queries in 1st-order logic} + \text{deterministic transitive closure.}$

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How to compute (deterministic) transitive closure in ELL?

$$\Phi_R : Q \mapsto \{(x, z) \mid x = z \vee (\exists y : (x, y) \in R \wedge (y, z) \in Q)\}$$

$R^* = \text{least fixpoint of } \Phi_R, \text{ obtained by iteration } (\Phi_R \text{ monotone}).$

Problem: because of  $\exists y$ ,  $\Phi_R$  uses  $Q$  once for each  $y$ ,  
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### Determinism (L vs NL) corresponds to linearity

$R$  deterministic  $\iff R$  corresponds to *partial function*  $f_R$ ; then

$\Phi_R : Q \mapsto \{(x, z) \mid x = z \vee (f_R(x), z) \in Q\}$  looks at  $Q$  only once!

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Evaluate the program in some *semantics*  $\llbracket - \rrbracket$ .

- type  $T \rightsquigarrow$  some mathematical structure  $\llbracket T \rrbracket$
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  - apply logspace algorithmic trickery here

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Evaluate the program in the *coherence space* semantics  $\llbracket - \rrbracket$ .

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Semantic evaluation has been used for complexity in  $\lambda$ -calculi before (e.g. in Hillebrand's PhD) but no precedent in linear logic.

# Conclusion

We brought some ideas to *implicit computational complexity* (ICC) based on (*elementary*) *linear logic* from other areas:

- descriptive complexity:  
finite models as inputs, lower bound via FO+DTC
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Part of a larger research program; other results include a LL-based characterization of *regular transductions*.