

TYPES 2023

Differentiation as a monad

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What's your favorite monad ?

What's your favorite monad ?

A monad over a type A :

- ▶ It *encapsulate* a certain kind of values: $u_A : A \rightarrow M(A)$.
- ▶ It allows *computation* on these values: $\mu_A : M(M(A)) \rightarrow M(A)$

Examples:

- ▶ Partiality: $M : A \mapsto A + \perp$, $u_A : a \mapsto a$
- ▶ Non-determinism: $M : A \mapsto \mathcal{P}(A)$, $u_A : a \mapsto \{a\}$
- ▶ Effect: $M : A \mapsto (S \rightarrow (A \times S))$, $u_A : a \mapsto (s \mapsto (a, s))$

The continuation monad

$$\begin{aligned} u_A : A &\Rightarrow ((A \Rightarrow B) \Rightarrow B) \\ a &\mapsto \lambda k.ka \end{aligned}$$

The continuation monad, twisted

Linear arrow \multimap : using exactly once its argument

$$\begin{aligned} u_A : A &\Rightarrow ((A \Rightarrow B) \multimap B) \\ a &\mapsto \lambda k. ka \end{aligned}$$

The continuation monad

Linear arrow \multimap : using exactly once its argument

$$u_A : A \multimap ((A \Rightarrow B) \multimap B)$$
$$a \mapsto \lambda k. k a$$

Making k , a non-linear map, linear

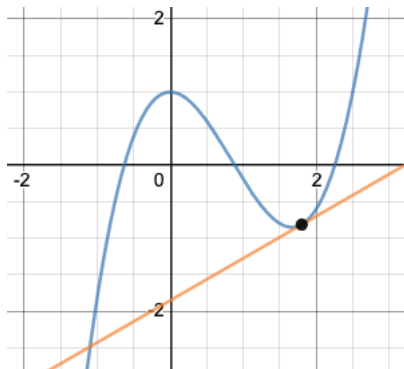
The continuation monad

Linear arrow \multimap : using exactly once its argument

$$u_A : A \multimap ((A \Rightarrow B) \multimap B)$$
$$a \mapsto \lambda k. D_0(k)a$$

Making k , a non-linear map, linear: differentiation

What's differentiation ?



The differential of a function at a point is its *best linear approximation* at that point.

From linearity to quantitative models

Functions

Power series

$$f = \sum_n f_n$$

Programs

Resources consumption or Probabilistic sums

$$p(x) = \sum p_n$$

From linearity to quantitative models

Functions

Power series

$$f = \sum_n f_n$$

f_n is n -linear

f is *Taylor*

$$f = \sum_n \frac{1}{n!} D_0^{(n)} f$$

Programs

Resources consumption or Probabilistic sums

$$p(x) = \sum p_n$$

p_n consumes exactly n -times its resources.

Programs can be approximated

$$(M)S = \sum_n \frac{1}{n!} \langle M \rangle S^{\otimes n}$$

- ▶ Experimentally, quantitative semantics is what gets you higher-order.
- ▶ It leads to new proof techniques on λ -calculus.
- ▶ A strong link with intersection types.



Simona Ronchi della Rocca's talk tomorrow!



Even when trying to avoid it, we stumble back on quantitative constructions [Dabrowski, K. 2018]

From linearity to quantitative models

Functions

Power series

$$f = \sum_n f_n$$

f_n is n -linear

$$f \text{ is Taylor} \\ f = \sum_n \frac{1}{n!} D_0^{(n)} f$$

Programs

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Core intuition: Differentials are enough to compute

The quantitative monad

Theorem [K. Lemay 2023]

The following:

- ▶ $M : E \rightarrow \mathcal{C}^\infty(E, \mathbb{K})'$
- ▶ $u : v \mapsto (f \mapsto D_0(f)(v))$
- ▶ $\mu : \delta_\phi \mapsto \sum \frac{1}{n!} \phi^{*n}$

is a monad in quantitative models of λ -calculus:

$$!u; \mu = id \quad \Leftrightarrow \quad f = \sum_n \frac{1}{n!} D_0^{(n)} f$$

The monad laws:

$$u_M; \mu = id$$

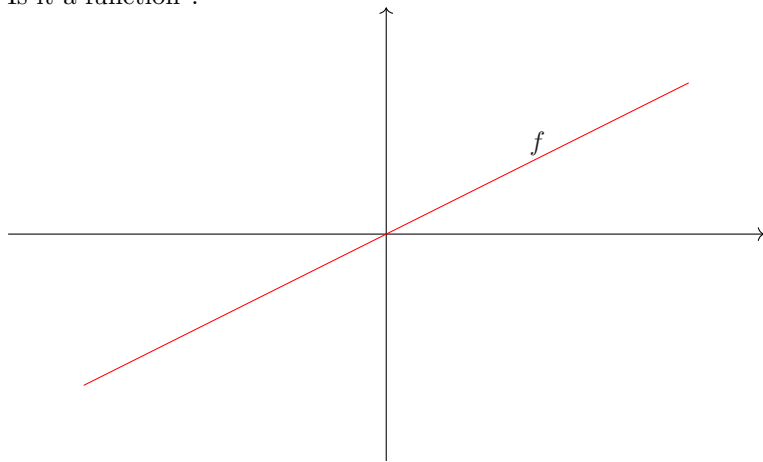
$$M(u); \mu = id$$

$$\mu_M; \mu = M(\mu); \mu$$

From functional analysis to functional programming, and back

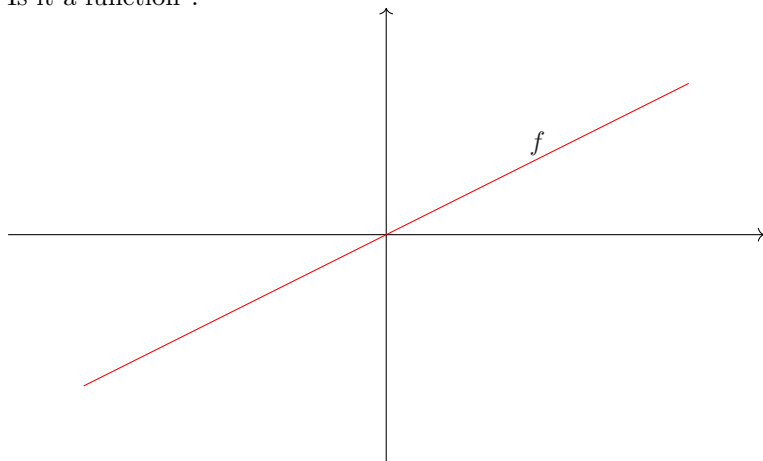
Surprise test

Is it a function ?



Surprise test

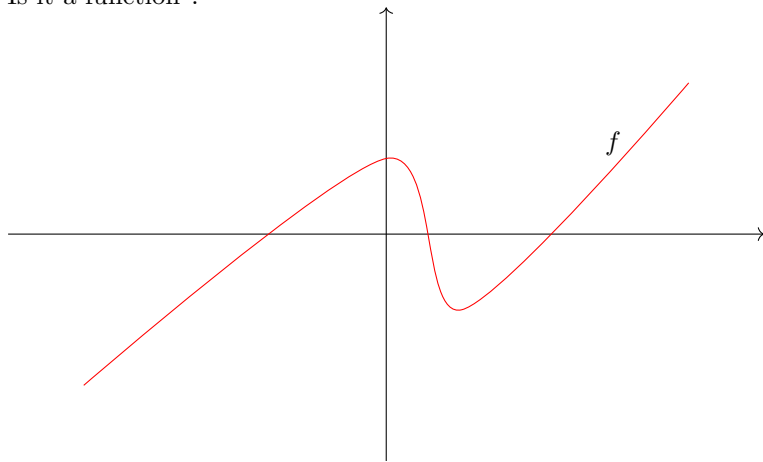
Is it a function ?



Yes, that's a linear function $f \in \mathcal{L}(\mathbb{R}, \mathbb{R})$

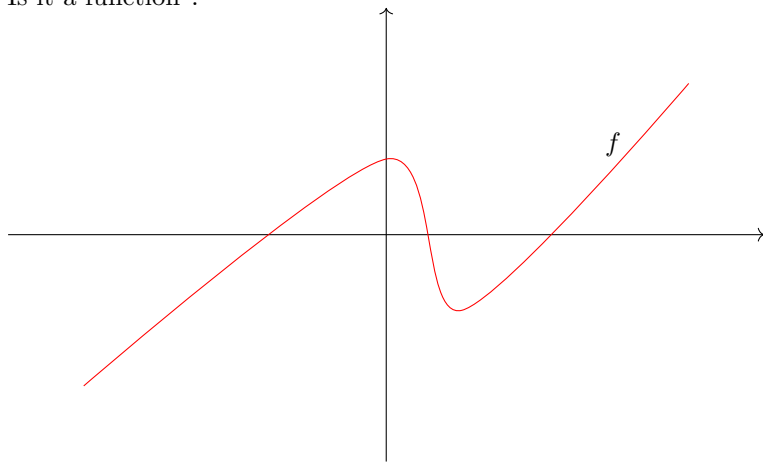
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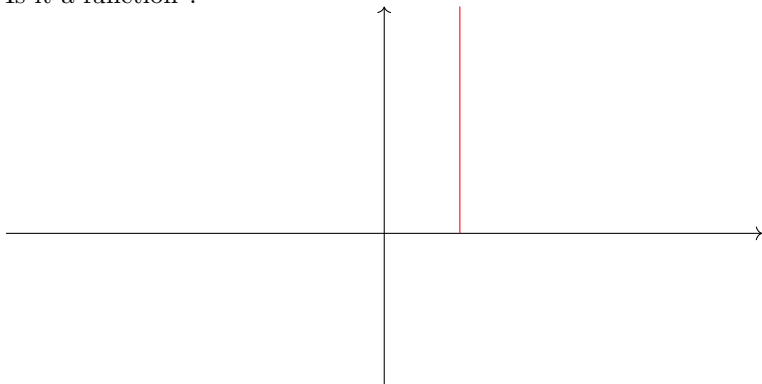
Is it a function ?



Yes, that's a smooth function $f \in \mathcal{C}^\infty(\mathbb{R}, \mathbb{R})$

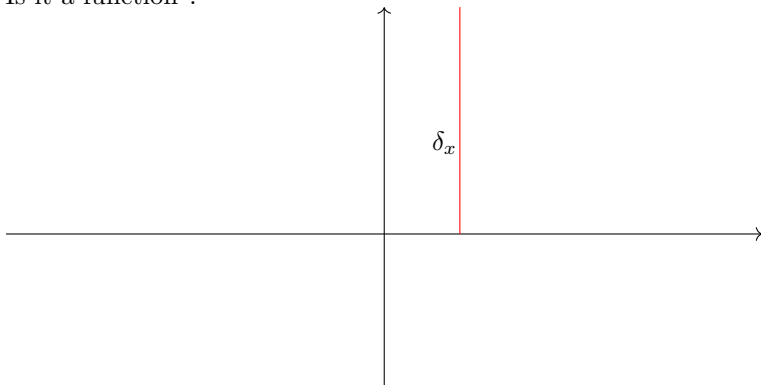
Surprise test

Is it a function ?



Surprise test

Is it a function ?



No, that's:

- ☐ A distribution
- ☐ A generalized function
- ☒ That's the argument to a program.

- 1 Introduction
 - Quantitative Semantics
- 2 Different type of functions
 - Smooth functions
 - Linear functions
 - Distribution theory
- 3 Analytic and Differential Linear Logic
- 4 Graded Monads in smooth settings

Programs are interpreted as functions...

Programs	Logic	Semantics
$\text{fun } (x:A) \rightarrow (t:B)$	Proof of $A \vdash B$	$f : A \rightarrow B.$
Types	Formulas	Objects
Execution	Cut-elimination	Equality

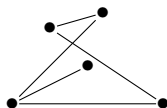
.. but *special* ones.

Programs act on programs $f : \mathcal{C}(A, B) \rightarrow C$

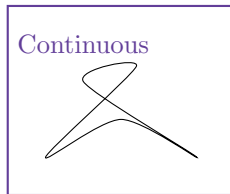
- (AxO) Domains A and spaces of functions $\mathcal{C}(A, B)$ are of the same kind.
- (AxF) Programs and function compute on several arguments:

$$f : A \times B \rightarrow C \equiv f : A \rightarrow \mathcal{C}(B, C)$$

Discrete



Continuous



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- ▶ (AxO) Domains A and spaces of functions $\mathcal{C}(A, B)$ are of the same kind.
- ▶ (AxF) Programs and function compute on several arguments:

$$f : A \times B \rightarrow C \equiv f : A \rightarrow \mathcal{C}(B, C)$$

- ▶ Lattices
- ▶ Games
- ▶ Graphs
- ▶ Vector spaces
- ▶ Sequences
- ▶ Normed spaces
- ▶ Topological vector spaces

Interpreting programs by smooth functions

$$p : A \Rightarrow B \quad f \in \mathcal{C}^\infty(A, B)$$

Probabilistic Programming

$$p \xrightarrow{\alpha} x$$

Differentiable Programming

$$D(p_1; p_2) = D(p_1); D(p_2)$$

- ▶ Correctness Properties $\llbracket D(p) \rrbracket = D(\llbracket p \rrbracket)$
- ▶ Completeness Properties $\forall f, \exists p, \llbracket p \rrbracket = f$
- ▶ **New** programming paradigms $p = d(q)$
- ▶ **New** mathematical structures $\mathcal{C}^\infty(E, F)$

Convenient vector spaces

a first interpretation of Higher-Order Smooth Functions

$$(A \times F): \mathcal{C}^\infty(A \times B, C) \simeq \mathcal{C}^\infty(A, \mathcal{C}^\infty(B, C))$$



Frölicher, Kriegl, Michor (1997)



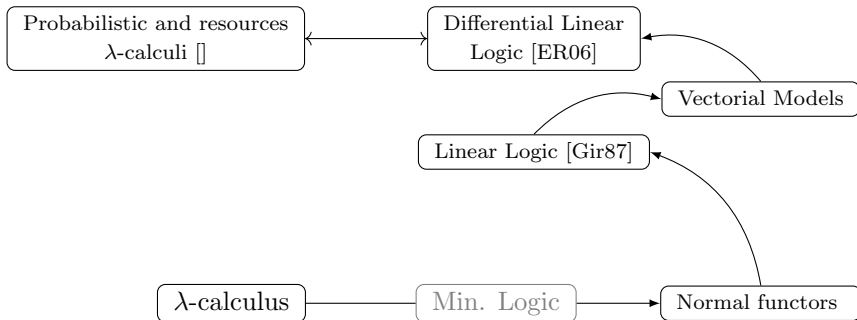
Blute, Ehrhard, Tasson (2012)

Perspective

Programs	Logic	Semantics
<code>fun (x:A)-> (t:B)</code>	Proof of $A \vdash B$	$f : A \rightarrow B.$
Types	Formulas	Objects
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Perspective

Programs	Logic	Semantics
$\text{fun } (x:A) \rightarrow (t:B)$	Proof of $A \vdash B$	$f : A \rightarrow B.$
Types	Formulas	Objects
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Interpreting programs by Linear Functions

$$\llbracket p \rrbracket \in \mathcal{L}(A, B)$$

(Ax0): If B is a complete or metrizable space, then so is $\mathcal{L}(A, B)$.

Trickier for A though

(AxF):

$$\mathcal{L}(A \otimes B, C) \simeq \mathcal{L}(A, \mathcal{L}(B, C))$$

► Always true algebraically.

Interpreting programs by Linear Functions

$$\llbracket p \rrbracket \in \mathcal{L}(A, B)$$

(A \times 0): If B is a complete or metrizable space, then so is $\mathcal{L}(A, B)$.

Trickier for A though

(A \times F):

$$\mathcal{L}_B(A \otimes_B B, C) \simeq \mathcal{L}_B(A, \mathcal{L}_B(B, C))$$

- ▶ Always true algebraically.
- ▶ Topologically, it depends on the set $B \subset \mathcal{P}(A)$ of bounded sets on which uniform convergence must be enforced.
- ▶ MANY topological tensor products: $\otimes_\beta, \otimes_\sigma, \otimes_\mu, \otimes_\varepsilon$.
- ▶ MANY duals: $E'_B := \mathcal{L}_B(E, \mathbb{R})$

WE ARE MISSING AN IMPORTANT CRITERIA

Not Not ... Who's there ?

Not Not ... Who's there ?

$$((A \Rightarrow \perp) \Rightarrow \perp) \simeq A$$

$$\mathcal{C}^\infty(\mathcal{C}^\infty(A, \mathbb{K}), \mathbb{K}) \simeq A$$

No one: not a chance for A smooth enough

Not Not ... Who's there ?

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$$\mathcal{C}^\infty(\mathcal{C}^\infty(A, \mathbb{K}), \mathbb{K}) \simeq A$$

No one: not a chance for A smooth enough

$$((A \multimap \perp) \multimap \perp) \simeq A$$

$$\mathcal{L}(\mathcal{L}(A, \mathbb{K}), \mathbb{K}) \simeq A$$

A lot of people!: Reflexive topological vector spaces.

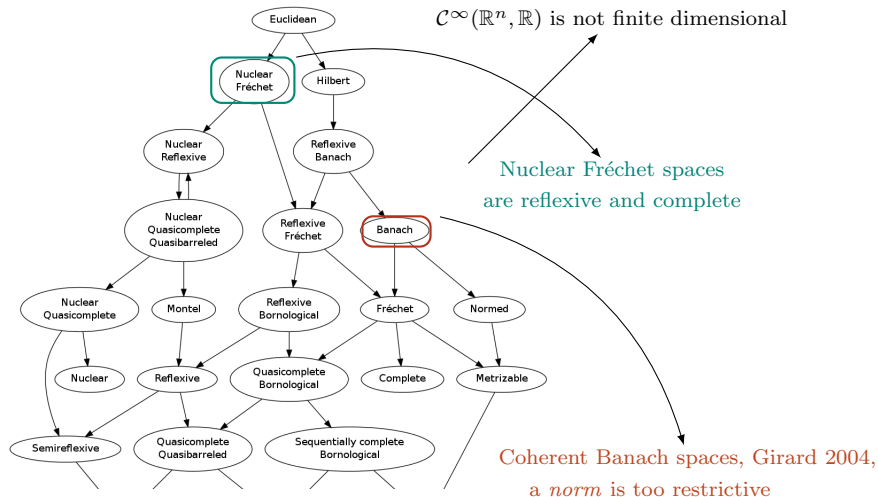
We have plenty of examples!

- ▶ Finite dimensional vector spaces
- ▶ Hilbert spaces
- ▶ Spaces on which an orthogonality relation can be defined ...

In general, reflexive spaces enjoy **poor stability properties**.

✗ higher-order, ✗ tensor product.

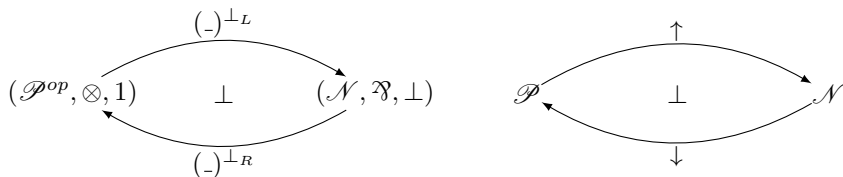
Interpreting types by reflexive topological vector spaces



Let us take the other way around, through Nuclear, Complete+Metrizable (=Fréchet) spaces.

Polarization as a solution to reflexivity

Semantics for polarized MLL : Melliès Chiralities



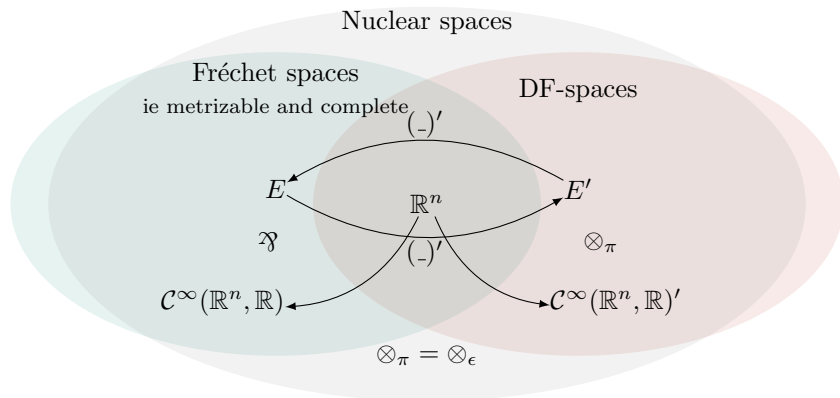
$$N^{\perp_R \perp_L} \simeq N$$

Replacing $(Ax F)$ with:

$$\mathcal{N}(\uparrow p \otimes n^{\perp_L}, m) \simeq \mathcal{N}(\uparrow p, n \otimes m)$$

Interpreting formulas by two categories of topological vector spaces, with a contravariant equivalence interpreting the involutive linear negation

Polarization as a solution to reflexivity



Notation : $E' := \mathcal{L}(E, \mathbb{R})$



Grothendieck, Produits tensoriels topologiques et espaces nucléaires, 1958



Melliès, A micrological study of negation, APAL 2017



K. A Logical Account for Linear Partial Differential Equations, LICS 2018 .

Linear implications and reflexivity



Old and dusty mathematicians

Property: $E \simeq (E'_\beta)'_\beta \Leftrightarrow E$ **barrelled** and E **weakly quasi complete**.

Barrelled spaces (Bourbaki): there for Banach-Steinhaus theorem.

Theorem

- ▶ **Barrelled** and **weak quasi-complete** form a model of polarized calculus (Melliès' Chiralities).
- ▶ Banach-Steinhaus is exactly **(AxF)!**

$$\mathcal{N}(\uparrow p \otimes n^{\perp_L}, m) \simeq \mathcal{N}(\uparrow p, n \wp m)$$

Mixing Linear and Non-Linear Proofs:
here comes the fun!

Not not ... Who's there ?

$$(A \Rightarrow \perp) \multimap \perp$$

$$\mathcal{L}(\mathcal{C}^\infty(A, \mathbb{R}), \mathbb{R}) = \mathcal{C}^\infty(A, \mathbb{R})'$$

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Semantics

Programs

Distributions

$$\phi \in \mathcal{C}^\infty(A, \mathbb{R})'$$

$$\text{e.g.: } \delta_x : f \mapsto f(x)$$

Context

$$C : (p : A \rightarrow \perp) \mapsto (\text{value} : \perp)$$

$$[-](x) : p \rightarrow p[x]$$

Not not ... Who's there ?

$$(A \Rightarrow \perp) \multimap \perp$$

$$\mathcal{L}(\mathcal{C}^\infty(A, \mathbb{R}), \mathbb{R}) = \mathcal{C}^\infty(A, \mathbb{R})'$$

Semantics

Programs

Distributions

$$\phi \in \mathcal{C}^\infty(\textcolor{brown}{A}, \textcolor{teal}{\mathbb{R}})'$$

Context

$$\textcolor{brown}{C} : (p : \textcolor{brown}{A} \rightarrow \textcolor{teal}{\perp}) \mapsto (\text{value} : \textcolor{teal}{\perp})$$

$$\text{e.g.: } \delta_{\textcolor{brown}{x}} : \textcolor{teal}{f} \mapsto \textcolor{teal}{f}(\textcolor{brown}{x})$$

$$[-](\textcolor{brown}{x}) : p \rightarrow p[\textcolor{brown}{x}]$$

Reflexivity:

$$\textcolor{teal}{f}(\textcolor{brown}{x}) = \delta_{\textcolor{brown}{x}}(\textcolor{teal}{f})$$

$$\textcolor{teal}{p}(\textcolor{brown}{x}) = \langle [-](\textcolor{brown}{x}) | \textcolor{teal}{p} \rangle$$

Not not ... Who's there ?

$$(A \Rightarrow \perp) \multimap \perp$$

$$\mathcal{L}(\mathcal{C}^\infty(A, \mathbb{R}), \mathbb{R}) = \mathcal{C}^\infty(A, \mathbb{R})'$$

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$$\phi \in \mathcal{C}^\infty(A, \mathbb{R})'$$

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$$C : (p : A \rightarrow \perp) \mapsto (\text{value} : \perp)$$

$$\text{e.g.: } \delta_x : f \mapsto f(x)$$

$$[-](x) : p \rightarrow p[x]$$

Reflexivity:

$$f(x) = \delta_x(f)$$

$$p(x) = \langle [-](x) | p \rangle$$

Differentiation

$$D_0(f)(x) = \langle D_0(-)(x) | f \rangle$$



Laurent Schwartz, Théorie des distributions, 1950

Distributions: Linear Contexts for Non-Linear Programs

$$\mathcal{C}^\infty(E, F)'$$

- ▶ (AxO) for distributions:
 - ▶ $\mathcal{C}^\infty(\mathbb{R}^n, \mathbb{R})$ is always Nuclear Fréchet and $\mathcal{C}^\infty(\mathbb{R}^n, \mathbb{R})'$ is Nuclear DF.
 - ▶ If F is Fréchet, then $\mathcal{C}^\infty(\mathbb{R}^n, F)$ is Fréchet
 - ▶ Higher order: a bit of work.
- ▶ (AxF) for distributions:
 - ▶ For linear maps: $\mathcal{L}_\beta(\hat{E}, \mathcal{L}_\beta(F, G)) \simeq \mathcal{L}_\beta(\widehat{E \otimes_\beta F}, G)$ ✓
 - ▶ For smooth maps: $\mathcal{C}^\infty(E, \mathcal{C}^\infty(F, G)) \simeq \mathcal{C}^\infty(E \times F, G)$?
 - ▶ From one to another:

Schwartz' Kernel Theorem ✓

$$\mathcal{C}^\infty(E, \mathbb{K})' \hat{\otimes} \mathcal{C}^\infty(F, \mathbb{K})' \simeq \mathcal{C}^\infty(E \times F, \mathbb{K})'$$

A monoidal operation on distributions

$$(\phi \in \mathcal{C}^\infty(E, \mathbb{R})' \otimes \psi \in \mathcal{C}^\infty(E, \mathbb{R})') \mapsto ?$$

Convolution, the monoidal operation on distributions:

$$\phi * \psi := f \mapsto \phi(x \mapsto \psi(y \mapsto f(x + y)))$$

Different from $\phi + \psi : f \mapsto \phi(f) + \psi(f)$

Examples:

$$\delta_x * \delta_y = \delta_{x+y}$$

$$\delta_x * D_0(-)(v) = D_x(-)(v)$$

$$D_0(-)(v) * D_0(-)(v) = D_0^{(2)}(-)(v)$$

There is no "multiplication" extending from functions to distributions, this is our multiplication !

Quantitative semantics, another look

$$\forall x, \forall v, f(x) = \sum_n \frac{1}{n!} D_0^{(n)} f(x)$$

Quantitative semantics, another look

$$\forall f, \forall x, \forall v, \langle f | \delta_x \rangle = \sum_n \frac{1}{n!} \langle f | D_0^{(n)}(-)(x) \rangle$$

Quantitative semantics, another look

$$\forall x, \forall v, \delta_x = \sum_n \frac{1}{n!} D_0^{(n)}(-)(x)$$

Quantitative semantics, another look

$$\forall x, \forall v, \delta_x = \underbrace{\sum_n}_{e} \frac{1}{n!} D_0^{(n)}(-)(x)$$

Quantitative semantics, another look

$$\forall \mathbf{x}, \forall v, \delta_{\mathbf{x}} = \underbrace{\sum_n}_{e} \frac{1}{n!} \overbrace{D_0(-)(\mathbf{x}) * \cdots * D_0(-)(\mathbf{x})}^{D_0(-)(\mathbf{x})^{*n}}$$

Quantitative semantics, another look

$$\forall x, \forall v, \delta_x = \sum_n \frac{1}{n!} D_0(-)(x) * \cdots * D_0(-)(x)$$

$$e^x = \sum_n \frac{1}{n!} x^n \quad id = e^* \circ (D_0(-))$$

Quantitative semantics, another look

$$\forall \mathbf{x}, \forall v, \delta_{\mathbf{x}} = \sum_n \frac{1}{n!} D_0(-)(\mathbf{x}) * \cdots * D_0(-)(\mathbf{x})$$

$$e^{\mathbf{x}} = \sum_n \frac{1}{n!} \mathbf{x}^n \quad id = e^* \circ (D_0(-))$$

A Quantitative Monad

- ▶ A functor $E \mapsto \mathcal{C}^\infty(E, \mathbb{R})'$ acting on a subcategory \mathcal{L} of topological vector spaces and linear maps.
- ▶ Differentiation as a unit: $u : \mathbf{x} \mapsto D_0(-)(\mathbf{x})$
- ▶ The convolutional exponential as a multiplication: $\mu : \delta_\phi \mapsto \sum_n \frac{1}{n!} \phi^{*n}$

Monad $\rightsquigarrow \forall f \in \mathcal{L}_!(A, B) \simeq \mathcal{C}^\infty(A, B)$, f is Taylor.

Examples: Relational model, Weighted Relational Model, Species, Nuclear Fréchet spaces

It was never about the quantitative semantics of
 λ -calculus.

Differential Linear Logic: from resources to distributions,
from discrete to continuous settings

Exponential rules of (Differential) Linear Logic

Programs

`fun (x:A)-> (t:B)`

Logic

Proof of $A \vdash B$

Semantics

$f : A \rightarrow B.$

Exponential rules of (Differential) Linear Logic

Programs	Logic	Semantics
$\text{fun } (x:A) \rightarrow (t:B)$	Proof of $A \vdash B$	$f : A \rightarrow B.$
Resources calculi	LINEAR LOGIC	Topological vector spaces

Exponential connectives:

$$\llbracket !A \rrbracket := \mathcal{C}^\infty(\llbracket A \rrbracket, \mathbb{K})'$$

$$\llbracket ?B \rrbracket := \mathcal{C}^\infty(\llbracket B \rrbracket', \mathbb{K})$$



Linear Logic, Jean-Yves Girard 1987



Differential Interaction Nets, Thomas Erhard and Laurent Regnier, 2006

Linear Logic

A decomposition of the implication

$$A \Rightarrow B \simeq !A \multimap B$$

Linear Logic

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$$A \Rightarrow B \simeq !A \multimap B$$

- Usual **non-linear** implication

A linear proof is in particular non-linear.

$A \vdash B$ is linear.

$!A \vdash B$ is non-linear.

$$\frac{A \vdash \Gamma}{!A \vdash \Gamma} \text{dereliction}$$

Slogan: ! in the hypotheses, speaking of resources.

Linear Logic

A decomposition of the implication

$$A \Rightarrow B \simeq !A \multimap B$$

- ▶ Usual **non-linear** implication
- ▶ **Linear** implication

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Linear Logic

A decomposition of the implication

$$A \Rightarrow B \simeq !A \multimap B$$

- ▶ Usual **non-linear** implication
- ▶ **Linear** implication
- ▶ **Exponential**: Usually, the duplicable copies of A .

A linear proof is in particular non-linear.

$A \vdash B$ is linear.

$!A \vdash B$ is non-linear.

$$\frac{A \vdash \Gamma}{!A \vdash \Gamma} \text{dereliction}$$

Slogan: ! in the hypotheses, speaking of resources.

Differential Linear Logic: co-structural rules



$$\frac{\ell : A \vdash B}{\ell : !A \vdash B} \text{d, dereliction}$$

$\text{linear} \hookrightarrow \text{non-linear}.$

$$\frac{f : !A \vdash B}{D_0(f) : A \vdash B} \bar{\text{d}}, \text{co-dereliction}$$

$\text{non-linear} \hookrightarrow \text{linear}$

Differential Linear Logic: co-structural rules



$$\frac{\ell : A \vdash B}{\ell : !A \vdash B} \text{d}$$

linear \hookrightarrow *non-linear*.

$$\frac{\vdash \Delta, v : A}{\vdash \Delta, (f \mapsto D_0(f)(v)) : !A} \bar{\text{d}}$$

non-linear \hookrightarrow *linear*

Differential Linear Logic: co-structural rules



$$\frac{\ell : A \vdash B}{\ell : !A \vdash B} \text{d}$$

linear \hookrightarrow *non-linear*.

$$\frac{\vdash \Delta, v : A}{\vdash \Delta, (\textcolor{teal}{f} \mapsto D_0(\textcolor{teal}{f})(v)) : !A} \bar{\text{d}}$$

non-linear \hookrightarrow *linear*

Cut-elimination:

$$\frac{\frac{\vdash \Gamma, v : !A}{\vdash \Gamma, !A} \bar{\text{d}} \quad \frac{\ell : A \vdash B}{\ell : !A \vdash B} \text{d, dereliction}}{\vdash \Gamma, \Delta} \text{cut}$$

\rightsquigarrow

$$\frac{\vdash \Gamma, A \quad \vdash \Delta, A^\perp}{\vdash \Gamma, \Delta} \text{cut}$$

Differential Linear Logic: co-structural rules



$$\frac{\ell : A \vdash B}{\ell : !A \vdash B} \text{d}$$

linear \hookrightarrow *non-linear*.

$$\frac{\vdash \Delta, v : A}{\vdash \Delta, (f \mapsto D_0(f)(v)) : !A} \bar{\text{d}}$$

non-linear \hookrightarrow *linear*

Cut-elimination:

$$\frac{\frac{\vdash \Gamma, v : A}{\vdash \Gamma, D_0(-)(v) : !A} \bar{\text{d}} \quad \frac{\ell : A \vdash B}{\ell : !A \vdash B} \text{d, dereliction}}{\vdash \Gamma, \Delta} \text{cut}$$

\rightsquigarrow

$$\frac{\vdash \Gamma, v : A \quad \ell : \vdash \Delta, A^\perp}{\vdash \Gamma, \Delta, D_0(\ell)(v) = \ell(v)} \text{cut}$$

Dereliction and co-dereliction:



$$\frac{\ell : A \vdash B}{\ell : !A \vdash B} \text{d}$$

linear \hookrightarrow *non-linear*.

$$\frac{\vdash \Delta, v : A}{\vdash \Delta, (f \mapsto D_0(f)(v)) : !A} \bar{\text{d}}$$

non-linear \hookrightarrow *linear*

Cut-elimination:

$$\frac{\frac{\vdash \Gamma, v : A}{\vdash \Gamma, D_0(-)(v) : !A} \bar{\text{d}} \quad \frac{\ell : A \vdash B}{\ell : !A \vdash B} \text{d, dereliction}}{\vdash \Gamma, \Delta} \text{cut}$$

\rightsquigarrow

$$\frac{\vdash \Gamma, v : A \quad \ell : \vdash \Delta, A^\perp}{\vdash \Gamma, \Delta, D_0(\ell)(v) = \ell(v)} \text{cut}$$

From resources to functions and distributions

(Co)-weakening

$$\frac{c : \vdash \Gamma}{cst_c : !A \vdash \Gamma} w$$

The constant function is non-linear

$$\frac{\vdash \Gamma}{\vdash \Gamma, \delta_0 : !A} \bar{w}$$

One can evaluate a function at 0

(Co)-contraction

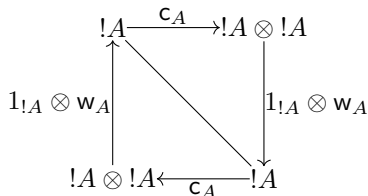
$$\frac{x : !A, y : !A \vdash g(x, y) : \Gamma}{x : !A \vdash g(x, x) : \Gamma} c$$

The multiplication of scalar functions

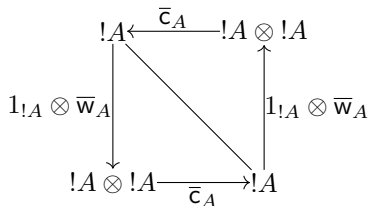
$$\frac{\vdash \Gamma, \phi : !A \quad \vdash \Delta, \psi : !A}{\vdash \Gamma, \Delta, \psi * \phi : !A} \bar{c}$$

Convolution of distributions

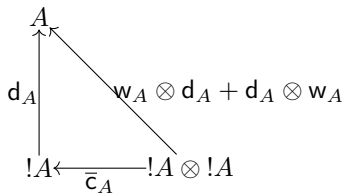
Symmetric cut-eliminations procedures



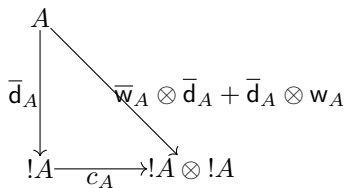
The function cst_1 is neutral for scalar multiplication



The dirac at 0 is neutral for the convolution



$$\phi * \psi(\ell) = \phi(\ell)\psi(cst_1) + \psi(\ell)\phi(cst_1)$$

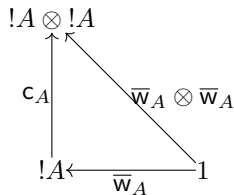
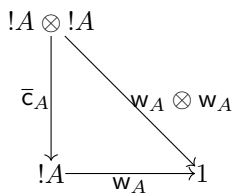


$$D_0(f \cdot g) = D_0(f) \cdot g(0) + D_0(g) \cdot f(0)$$

Symmetric cut-elimination procedures

$$\bar{d}; w = 0 \text{ and } \bar{w}; d = 0$$

$$D_0(cst_1) = 0 \text{ and } \ell(0) = 0$$



$$(\phi * \psi)(cst_1) = \phi(cst_1) \cdot \psi(cst_1)$$

$$(f \cdot g)(0) = f(0) \cdot g(0)$$

$$\otimes = \cdot \text{ in } \mathbb{R}$$

Finitary differential Linear Logic

The first version by Ehrhard and Regnier in 2006:

$$\frac{\vdash \Gamma}{\vdash \Gamma, \text{cst}_1 : ?A} w$$

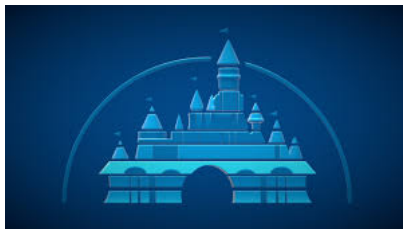
$$\frac{\vdash \Gamma, f : ?A, g : ?A}{\vdash \Gamma, f.g : ?A} c$$

$$\frac{\vdash \Gamma, \ell : A}{\vdash \Gamma, \ell : ?A} d$$

$$\frac{\vdash \Gamma}{\vdash \Gamma, \delta_0 : !A} \bar{w}$$

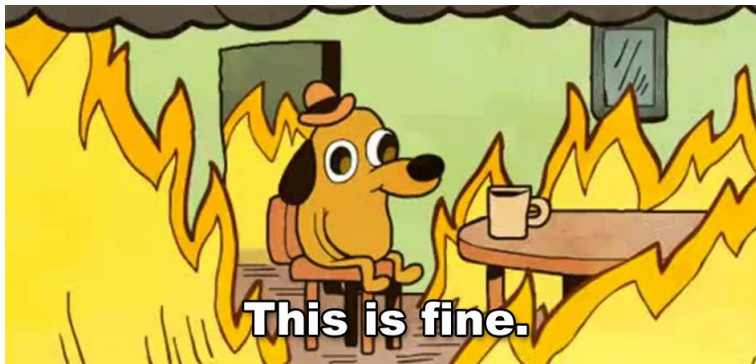
$$\frac{\vdash \Gamma, \phi : !A \quad \vdash \Delta, \psi : !A}{\vdash \Gamma, \Delta, \psi * \phi : !A} \bar{c}$$

$$\frac{\vdash \Gamma, x : A}{\vdash \Gamma, D_0(-)(x) : !A} \bar{d}$$



It's a maths world.

Higher-Order



Higher-Order via promotion

Exponential rules of Linear Logic (Resources)

$$\frac{\vdash \Gamma}{\vdash \Gamma, \text{cst}_1 : ?A} w \quad \frac{\vdash \Gamma, f : ?A, g : ?A}{\vdash \Gamma, f.g : ?A} c \quad \frac{\vdash \Gamma, \ell : A}{\vdash \Gamma, \ell : ?A} d \quad \boxed{\frac{! \Gamma \vdash x : A}{! \Gamma \vdash \delta_x : !A} p}$$

Exponential rules added by Differential Linear Logic (Distributions)

$$\frac{\vdash \Gamma}{\vdash \Gamma, \delta_0 : !A} \bar{w} \quad \frac{\vdash \Gamma, \phi : !A \quad \vdash \Delta, \psi : !A}{\vdash \Gamma, \Delta, \psi * \phi : !A} \bar{c} \quad \frac{\vdash \Gamma, x : A}{\vdash \Gamma, D_0(-)(x) : !A} \bar{d}$$

The promotion rule $p : !A \rightarrow !!A \quad \delta_x \mapsto \delta_{\delta_x}$:

- Makes $(!, p)$ a co-monad : $p; d = \text{id}$.
- Is a co-monoidal operation on $!A$: $p; c = c; p \otimes p$
- The cut-elimination between p and \bar{d} express the chain rule:

$$D_0(g \circ f) = D_{f(0)}g \circ D_0f \quad \bar{d}; p = \bar{w} \otimes \bar{d}; p \otimes \bar{d}; \bar{c}$$

Codigging

Exponential rules of Linear Logic (Resources or functions)

$$\frac{\vdash \Gamma}{\vdash \Gamma, \text{cst}_1 : ?A} w \quad \frac{\vdash \Gamma, f : ?A, g : ?A}{\vdash \Gamma, f.g : ?A} c \quad \frac{\vdash \Gamma, \ell : A}{\vdash \Gamma, \ell : ?A} d \quad \boxed{\frac{! \Gamma \vdash x : A}{! \Gamma \vdash \delta_x : !A} p}$$

Exponential rules added by Differential Linear Logic (Distributions)

$$\frac{\vdash \Gamma}{\vdash \Gamma, \delta_0 : !A} \bar{w} \quad \frac{\vdash \Gamma, \phi : !A \quad \vdash \Delta, \psi : !A}{\vdash \Gamma, \Delta, \psi * \phi : !A} \bar{c} \quad \frac{\vdash \Gamma, x : A}{\vdash \Gamma, D_0(-)(x) : !A} \bar{d} \quad \frac{? \Gamma \vdash x : A}{? \Gamma \vdash - : ?A} \bar{p}$$

Digging $p : !A \rightarrow !!A$:

- ▶ $p; d = \text{id}$.
- ▶ $p; c = c; p \otimes p$
- ▶ $\bar{d}; p = \bar{w} \otimes \bar{d}; p \otimes \bar{d}; \bar{c}$
- ..

Co-digging $? \bar{p} : !!A \rightarrow !A$:

- ▶ $\bar{d}; \bar{p} = \text{id}$
- ▶ $\bar{c}; \bar{p} = \bar{p} \otimes \bar{p}; \bar{c}$
- ▶ $\bar{p}; d = c; \bar{p} \otimes d; w \otimes d$

Codigging

Exponential rules of Linear Logic (Resources or functions)

$$\frac{\vdash \Gamma}{\vdash \Gamma, \text{cst}_1 : ?A} w \quad \frac{\vdash \Gamma, f : ?A, g : ?A}{\vdash \Gamma, f.g : ?A} c \quad \frac{\vdash \Gamma, \ell : A}{\vdash \Gamma, \ell : ?A} d \quad \boxed{\frac{! \Gamma \vdash x : A}{! \Gamma \vdash \delta_x : !A} p}$$

Exponential rules added by Differential Linear Logic (Distributions)

$$\frac{\vdash \Gamma}{\vdash \Gamma, \delta_0 : !A} \bar{w} \quad \frac{\vdash \Gamma, \phi : !A \quad \vdash \Delta, \psi : !A}{\vdash \Gamma, \Delta, \psi * \phi : !A} \bar{c} \quad \frac{\vdash \Gamma, x : A}{\vdash \Gamma, D_0(-)(x) : !A} \bar{d} \quad \frac{? \Gamma \vdash x : A}{? \Gamma \vdash - : ?A} \bar{p}$$

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- ▶ $p; d = \text{id}$.
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- ▶ $\bar{d}; p = \bar{w} \otimes \bar{d}; p \otimes \bar{d}; \bar{c}$

Co-digging $? \bar{p} : !!A \rightarrow !A$: $g : !A \Rightarrow !A$

- ▶ $\bar{d}; \bar{p} = \text{id} \quad D_0(g) = \text{id}$.
- ▶ $\bar{c}; \bar{p} = \bar{p} \otimes \bar{p}; \bar{c} \quad g(x + y) = g(x) * g(y)$
- ▶ $\bar{p}; d = c; \bar{p} \otimes d; w \otimes d$

The missing rule of Differential Linear Logic

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- ▶ $p; d = \text{id}.$
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- ▶ $\bar{d}; p = \bar{w} \otimes \bar{d}; p \otimes \bar{d}; \bar{c}$

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- ▶ $\bar{p}; d = c; \bar{p} \otimes d; w \otimes d$?

Implicit definition of the exponential:

$$g = \exp^* : \phi \mapsto \sum_n \frac{1}{n!} \phi^* \quad \bar{p} : \delta_\phi \mapsto \sum_n \frac{1}{n!} \phi^{*n}$$

The missing rule of Differential Linear Logic

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The co-chain rule:

$$\bar{p}(\delta_\phi)(\ell) = \sum_{n \geq 0} \frac{1}{n!} \phi^{*n}(\ell) = \sum_{n \geq 1} \frac{n}{n!} \phi(l) \cdot \phi^{*(n-1)}(\text{cst}_1) = \phi(\ell) \cdot \bar{p}(\delta_\phi)(\text{cst}_1) = c; \bar{p} \otimes d; w \otimes d$$

The missing rule of Differential Linear Logic

Digging $p : !A \rightarrow !!A$:

- ▶ $p; d = \text{id}.$
- ▶ $p; c = c; p \otimes p$
- ▶ $\bar{d}; p = \bar{w} \otimes \bar{d}; p \otimes \bar{d}; \bar{c}$

Co-digging ? $\bar{p} : !!A \rightarrow !A$: $g : !A \Rightarrow !A$

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The monadic rules:

$$\bar{!}d; \bar{p} = \text{id} \quad \forall v, \bar{p}(\delta_{D_0(-)}(v)) = \delta_v \quad \forall v, \forall f, \sum_n \frac{1}{n!} D_0^{(n)} f(v) = f(v)$$

A completely uniform logical and categorical structure

Exponential connectives:

$$\llbracket !A \rrbracket := \mathcal{C}^\infty(\llbracket A \rrbracket, \mathbb{K})' \quad \llbracket ?A \rrbracket := \mathcal{C}^\infty(\llbracket A \rrbracket', \mathbb{K})$$

$\frac{\vdash \Gamma}{\vdash \Gamma, \text{cst}_1 : ?A} w$	$\frac{\vdash \Gamma, f : ?A, g : ?A}{\vdash \Gamma, f.g : ?A} c$	Linear Logic [Girard1987]
$\frac{\vdash \Gamma, \ell : A}{\vdash \Gamma, \ell : ?A} d$	$\frac{! \Gamma \vdash x : A}{! \Gamma \vdash \delta_x : !A} p$	
$\frac{\vdash \Gamma}{\vdash \Gamma, \delta_0 : !A} \bar{w}$	$\frac{\vdash \Gamma, \phi : !A \quad \vdash \Delta, \psi : !A}{\vdash \Gamma, \Delta, \psi * \phi : !A} \bar{c}$	
$\frac{\vdash \Gamma, x : A}{\vdash \Gamma, D_0(-)(x) : !A} \bar{d}$	$\frac{? \Gamma \vdash x : A}{? \Gamma \vdash e^x - : ?A} \bar{p}$	

Differential Linear Logic
[Ehrhard&Regnier2006]

[Kerjean & Pacaud Lemay 2023]

A reason for this symmetry

Exponential connectives:

$$\llbracket !A \rrbracket := \mathcal{C}^\infty(\llbracket A \rrbracket, \mathbb{K})' \quad \llbracket ?A \rrbracket := \mathcal{C}^\infty(\llbracket A \rrbracket', \mathbb{K})$$

Do you remember the Laplace transformation ?

A continuous version of a power series

$$\mathcal{L} : \begin{cases} !E & \rightarrow ?E \\ \phi & \mapsto (\ell^{E'} \mapsto \phi(y^E \mapsto e^{<\ell|y>})) \end{cases}$$

$$\mathcal{L}(\overline{w}, \overline{c}, \overline{d}, \overline{p}) = w, c, d, p$$

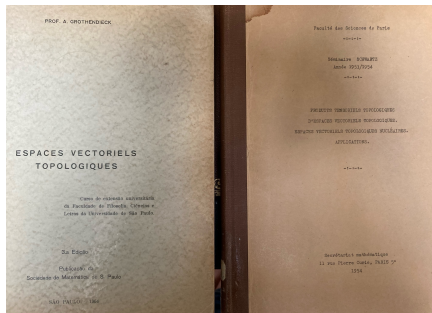
Let's make things concrete

- ▶ $M : E \rightarrow !E := \llbracket (\llbracket E \rrbracket \Rightarrow \perp) \multimap \perp \rrbracket = \mathcal{C}^\infty(\llbracket E \rrbracket, \mathbb{K})'$
- ▶ $u : v \mapsto (f \mapsto D_0(f)(v))$
- ▶ $\mu : \delta_\phi \mapsto \sum \frac{1}{n!} \phi^{*n}$

Let's make things concrete

- ▶ $M : E \rightarrow !E := \llbracket ([E] \Rightarrow \perp) \multimap \perp \rrbracket = \mathcal{C}^\infty([E], \mathbb{K})'$
- ▶ $u : v \mapsto (f \mapsto D_0(f)(v))$
- ▶ $\mu : \delta_\phi \mapsto \sum \frac{1}{n!} \phi^{*n}$

The way I make things concrete



Existential questions

Does $\bar{p} = e^*$ even exists ?

Bad news: sums need to converge.

At least at every point: $(\forall f, \forall \phi, \sum_n \frac{1}{n!} \phi^{*n}(f)) \in \mathbb{R}$

- ▶ In discrete models of computations (e.g: relations over sets), sums are union, not an issue.
- ▶ In continuous models of computations, well...

Convolutional exponential VS exponential maps

(AxC) Proofs/Programs/Functions must compose

$$\bar{\mathbf{p}} : \delta_{\phi} \mapsto \sum \frac{1}{n!} \phi^{*n}$$

Convolutional exponential VS exponential maps

(AxC) Proofs/Programs/Functions must compose

$$\bar{\mathbf{p}} : \delta_{\delta_{\mathbf{x}}} \mapsto \sum \frac{1}{n!} \delta_{n\mathbf{x}} = \left(f \mapsto \sum \frac{1}{n!} f(n\mathbf{x}) \right)$$

Convolutional exponential VS exponential maps

(AxC) Proofs/Programs/Functions must compose

$\bar{p} :$

$$\bar{p}(\delta_{\delta_x})(x \mapsto e^x) = \sum \frac{1}{n!} e^{nx} = e^{e^x} \quad \checkmark$$

$$\bar{p}(\delta_{\delta_x})(x \mapsto e^{e^x}) = \sum \frac{1}{n!} e^{e^{nx}} \quad \times$$

This example is due to T. Ehrhard

- ▶ p and \bar{p} do not mix well.
- ▶ We need to *quantify* over the exponential growth of the functions \bar{p} is applied to.

Grading Exponentials

$$A \multimap B$$

Linear programs/proofs/functions
using exactly once their **resource**

$$!A \multimap B$$

Usual programs/proofs/functions

$$!_n A \multimap B$$

- ▶ n -linear functions.
- ▶ Programs/proofs using exactly n -times their resource.
- ▶ Quantitative semantics: $!A = \sum !_n A$.

Exponentials indexed by semi-rings S

$$\forall s \in S, !_s A \multimap B$$



*Jean-Yves Girard, Andre Scedrov,
Philip J. Scott, 1992*



Martin Hoffman, Ugo Dal Lago, 2009

Applications in implicit complexity and differential privacy.

From resources to differential equations

$$\frac{\vdash \Gamma}{\vdash \Gamma, \text{cst}_1 : ?A} w$$

$$\frac{\vdash \Gamma, f : ?A, g : ?A}{\vdash \Gamma, f.g : ?A} c$$

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$$\frac{! \Gamma \vdash x : A}{! \Gamma \vdash \delta_x : !A} p$$

$$\frac{}{\vdash \delta_0 : !A} \bar{w}$$

$$\frac{\vdash \Gamma, \phi : !A \quad \vdash \Delta, \psi : !A}{\vdash \Gamma, \Delta, \psi * \phi : !A} \bar{c}$$

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$$\frac{? \Gamma \vdash x : A}{? \Gamma \vdash e^{x|-} : ?A} \bar{p}$$

From resources to differential equations

Grading LL: a story of resources, again

$$\begin{array}{c}
 \frac{\vdash \Gamma}{\vdash \Gamma, \text{cst}_1 : ?_0 A} w \qquad \frac{\vdash \Gamma, f : ?_{\mathbf{x}} A, g : ?_{\mathbf{y}} A}{\vdash \Gamma, f.g : ?_{\mathbf{x}+\mathbf{y}} A} c \\
 \\
 \frac{\vdash \Gamma, \ell : A}{\vdash \Gamma, \ell : ?_1 A} d \qquad \frac{! \Gamma_y \vdash x : A}{! \Gamma_{z \times y} \vdash \delta_x : !_z A} p \\
 \\
 \frac{}{\vdash \delta_0 : !A} \bar{w} \qquad \frac{\vdash \Gamma, \phi : !A \quad \vdash \Delta, \psi : !A}{\vdash \Gamma, \Delta, \psi * \phi : !A} \bar{c} \\
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 \end{array}$$

From resources to differential equations

Grading LL: a story of resources, again

$$\begin{array}{c} \frac{\vdash \Gamma}{\vdash \Gamma, \text{cst}_1 : ?A} w \qquad \frac{\vdash \Gamma, f : ?A, g : ?A}{\vdash \Gamma, f.g : ?A} c \\[2ex] \frac{\vdash \Gamma, \ell : A}{\vdash \Gamma, \ell : ?A} d \qquad \frac{! \Gamma \vdash x : A}{! \Gamma \vdash \delta_x : !A} p \end{array}$$

Grading DiLL: well, not so clear

$$\begin{array}{c} \frac{}{\vdash \delta_0 : !_?A} \bar{w} \qquad \frac{\vdash \Gamma, \phi : !_?A \quad \vdash \Delta, \psi : !_?A}{\vdash \Gamma, \Delta, \psi * \phi : !_?A} \bar{c} \\[2ex] \frac{\vdash \Gamma, x : A}{\vdash \Gamma, D_0(-)(x) : !_?A} \bar{d} \qquad \frac{?_? \Gamma \vdash x : A}{?_? \Gamma \vdash e^x | - : ?_?A} \bar{p} \end{array}$$

Joining Quantitative, Graded and Differential Linear Logic

(AxC) Proofs/Programs/Functions must compose

$$\mu : \delta_{\delta_x} \mapsto \sum \frac{1}{n!} \delta_{nx} = \left(f \mapsto \sum \frac{1}{n!} f(nx) \right)$$

$$\mu(\delta_{\delta_x})(x \mapsto e^x) = \sum \frac{1}{n!} e^{nx} = e^{e^x n} \quad \checkmark$$

$$\mu(\delta_{\delta_x})(x \mapsto e^{e^x}) = \sum \frac{1}{n!} e^{e^{nx}} \quad \times$$

The quantitative monad does not apply to all functions, but only to those whose convergence is exponentially bound, according to some Young function θ .

$$?_{\theta,m}(F) := \{f : F' \rightarrow \mathbb{C}, \forall m, \exists K, \forall z, |f(z)| \leq K e^{\theta(m||z||)}\}.$$

$$\bar{p} : !_{\theta_1} (!_{\theta_2}(E)) \rightarrow !_{(\theta_1 * e^{\theta_2})^*}(E)$$



Gannoun, Hachaichi, Ouerdiane, et Rezgui, Un théorème de dualité entre espaces de fonctions holomorphes à croissance exponentielles, 1999

Les fonctions parlent aux fonctions

$$!_{\theta,m}(F) := \{f : F \rightarrow \mathbb{C}, \forall m, \exists K, \forall z, |f(z)| \leq K e^{\theta(m||z||)}\}'.$$

issue: $||z|| : F$ needs to be normed.

- One single norm is too restrictive: we want to quantify over the convergence of the derivative of each function.
- The power of Fréchet spaces comes from their descriptions as a countable limit of Banach spaces:

$$F' := \lim_p N'_p$$

$$!_{\theta}F := \lim_{m,p} (?_{m,\theta} F_p)'$$

Les fonctions parlent aux fonctions

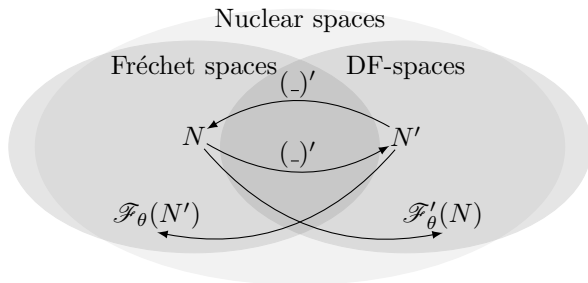
We have a quantitative and graded monad of Nuclear Dual of Fréchet spaces.

$$!_{\theta}F_p : \{f : F_p \rightarrow \mathbb{C}, \forall m, \exists K, \forall z, |f(z)| \leq Ke^{\theta(m||z||)}\}'$$

$$!_{\theta}F := (\lim_{m,p} (?_{m,\theta}F_p)')'$$

The space of Young functions is a semi-ring with a new duality operation $(-)^{\star}$.

$$\Theta : \{\theta, +, (- \times e^{-}), (-)^{\star}\}$$

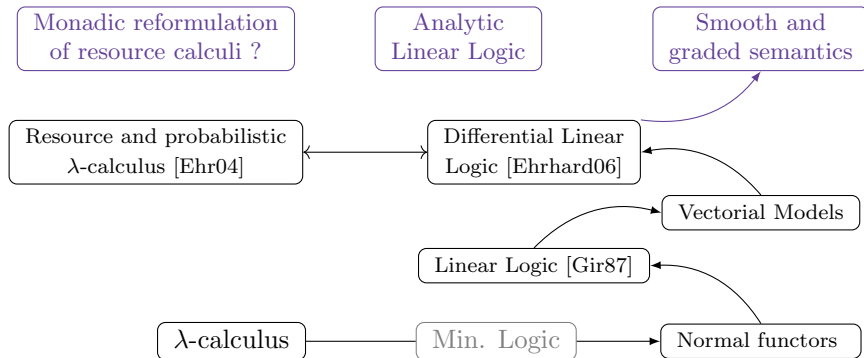


Recap

Programs	Logic	Semantics
<code>fun (x:A)-> (t:B)</code>	Proof of $A \vdash B$	$f : A \rightarrow B.$
Types	Formulas	Objects
Execution	Cut-elimination	Equality

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Differentiable Programming

Monadic reformulation
of resource calculi ?

Analytic
Linear Logic

Smooth and
graded semantics

Resource and probabilistic
 λ -calculus [Ehr04]

Differential Linear
Logic [Ehrhard06]

Vectorial Models

Linear Logic [Gir87]

λ -calculus

Min. Logic

Normal functors

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Differentiable Programming

Differential Operators

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Analytic
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 λ -calculus [Ehr04]

Differential Linear
Logic [Ehrhard06]

Vectorial Models

Automatic
Differentiation [80s]

Linear Logic [Gir87]

Dialectica [Göd58]

λ -calculus

Min. Logic

Normal functors

Perspectives

Quantitative Semantics: Approximating functions by Polynomials

- ▶ Taylor: Orthogonal Basis = $\{X^n\}$
 - ▶ recurrence : $T_{n+1} = XT_n$
 - ▶ composition $T_n \circ T_m = T_{nm}$
- ▶ Other Bases ? Chebychev ?
 - ▶ recurrence $T_{n+2} = 2T_{n+1} - T_n$
 - ▶ composition $T_n \circ T_m = T_{nm}$
- ▶ Characterization of approximation and orthogonality ?

Grading with partial differential operators

Grading by Linear Partial Differential Equations with constant coef.

$$\llbracket !_{\mathbf{D}} A \rrbracket := D((C^\infty(\llbracket A \rrbracket, \mathbb{K}))' \quad \llbracket ?_{\mathbf{D}} A \rrbracket := D^{-1}(C^\infty(\llbracket A' \rrbracket, \mathbb{K}))$$

parameters of the equations

solutions of the equations

$$D(-) = f \quad \phi \circ D = -$$

$$\begin{array}{c} \frac{\vdash \Gamma}{\vdash \Gamma, \text{cst}_1 : ?_{Id} A} w \quad \frac{\vdash \Gamma, f : ?_{D_1} A, g : ?_{D_2} A}{\vdash \Gamma, f.g : ?_{D_1 \circ D_2} A} c \quad \frac{\vdash \Gamma, f : ?_{D_1} A}{\vdash \Gamma, f * E_{D_2} : ?_{D_1 \circ D_2} A} d \\ \frac{}{\vdash \delta_0 : !_{Id} A} \bar{w} \quad \frac{\vdash \Gamma, \phi : !_{D_1} A \quad \vdash \Delta, \psi : !_{D_2} A}{\vdash \Gamma, \Delta, \psi * \phi : !_{D_1 \circ D_2} A} \bar{c} \quad \frac{\vdash \Gamma, \phi : !_{D_1} A}{\vdash \Gamma, \phi \circ D_2 : !_{D_1 \circ D_2} A} \bar{d} \end{array}$$

Monoid: (D, \circ, Id)



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Grading with partial differential operators

Grading by Linear Partial Differential Equations with constant coef.

$$\llbracket !_{\mathbf{D}} A \rrbracket := D((C^\infty(\llbracket A \rrbracket, \mathbb{K}))') \quad \llbracket ?_{\mathbf{D}} A \rrbracket := D^{-1}(C^\infty(\llbracket A' \rrbracket, \mathbb{K}))$$

parameters of the equations

solutions of the equations

For D an LPDOcc: $(\phi \circ D) * \psi = (\phi * \psi) \circ D$ $D(E_D * f) = f$

$$\frac{\vdash \Gamma}{\vdash \Gamma, cst_1 : ?_{Id} A} w \quad \frac{\vdash \Gamma, f : ?_{D_1} A, g : ?_{D_2} A}{\vdash \Gamma, f.g : ?_{D_1 \circ D_2} A} c \quad \frac{\vdash \Gamma, f : ?_{D_1} A}{\vdash \Gamma, f * E_{D_2} : ?_{D_1 \circ D_2} A} d$$

$$\frac{}{\vdash \delta_0 : !_{Id} A} \bar{w} \quad \frac{\vdash \Gamma, \phi : !_{D_1} A \quad \vdash \Delta, \psi : !_{D_2} A}{\vdash \Gamma, \Delta, \psi * \phi : !_{D_1 \circ D_2} A} \bar{c} \quad \frac{\vdash \Gamma, \phi : !_{D_1} A}{\vdash \Gamma, \phi \circ D_2 : !_{D_1 \circ D_2} A} \bar{d}$$

Monoid: (D, \circ, Id)



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The computational content of differentiation.

The codereliction of differential proof nets: In terms of polarity in linear logic [23], the $\forall\multimap$ -free constraint characterizes the formulas of intuitionistic logic that can be built only from positive connectives (\oplus , \otimes , 0 , 1 , $!$) and the why-not connective (“?”). In this framework, Markov’s principle expresses that from such a $\forall\multimap$ -free formula A (e.g. $? \oplus_x (?A(x) \otimes ?B(x))$) where the presence of “?” indicates that the proof possibly used weakening (efq or throw) or contraction (catch), a linear proof of A purged from the occurrences of its “?” connective can be extracted (meaning for the example above a proof of $\oplus_x (A(x) \otimes B(x))$). Interestingly, the removal of the “?”, i.e. the steps from $?P$ to P , correspond to applying the codereliction rule of differential proof nets [24].

Differentiation : $(?P = (P \multimap \perp) \Rightarrow \perp) \rightarrow ((P \multimap \perp) \multimap \perp) \equiv P$



Hugo Herbelin, “An intuitionistic logic that proves Markov’s principle”, LICS ’10 .

This can also be witnessed by identifying the computational content of Dialectica as a CPS style differential λ -calculus.[PMP,K 22]

Open questions

- ▶ \bar{p} and p do not interact well: cut-elimination ?
- ▶ More intricate differential operators semi-rings? Higher-order methods ?
Can we embed approximate resolution methods in the sequent calculus ?
- ▶ Can we express resolution methods in differential λ -calculi ?
- ▶ Can we make the categorical semantics of differentiation closer to the one of type theory ?

Conclusion

Take away

- ▶ The semantics of λ -calculus is not as much about discrete structures than about approximating continuous ones.
- ▶ The notion of linear type $- \multimap -$ has been influential in functional programming. Let's now make use of the distribution type !, which internalizes external transformations on programs.
- ▶ Functional analysis and functional programming might enrich each other: the former gives the latter **new concepts**, the latter gives the former **new structures**.

Thank you for listening!