

Fossacs 2019

Higher-Order Distributions for Linear Logic

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Inria

logo_oxfor

Differentiating Programs - Differentiating Functions

- ▶ Differentiation in Theoretical Computer Science : Automatic Differentiation, Incremental Computing, **Differential Linear Logic**... [Discrete]
- ▶ Differentiation in Mathematics : Differential Geometry, Numerical Analysis, Functional analysis ... [continuous]

Differentiation in Computer Science the same as Differentiation in Mathematics ?

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- ▶ [K18] : Models of Differential Linear Logic with Distributions and Differential Equations, without Higher-Order. $C^\infty(\mathbb{R}^n, \mathbb{R})$
- ▶ **Today : going to Higher Order.** $C^\infty(E, \mathbb{R})$?

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From mathematics to computer science.



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From models for physics to models for computing.



Higher-Order

Curry-Howard-Lambek

The syntax mirrors the semantics.

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

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λ -calculus

Coherence spaces [Girard87]

Linear maps $f : A \multimap B$

Non-linear maps $f : !A \multimap B$

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Linear proofs $f : A \vdash B$

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

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Vectorial Models [Ehrhard02/05]

Power series $f = \sum_n f_n$

Differentiation $D_0 : f \mapsto f_1$

Differential Linear Logic [Ehrhard&Regnier06]

Linear logic

Usual Implication

A linear implication

$$A \Rightarrow B = !A \multimap B$$
$$\mathcal{C}^\infty(A, B) \simeq \mathcal{L}(!A, B)$$

A proof is linear when it uses only once its hypothesis A.

Linear logic

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Exponential

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A focus on linearity

- Higher-Order is about *Seely's isomorphism*.

$$\mathcal{C}^\infty(A \times B, C) \simeq \mathcal{C}^\infty(A, \mathcal{C}^\infty(B, C))$$
$$\mathcal{L}(!(A \times B), C) \simeq \mathcal{L}(!A, \mathcal{L}(!B, C))$$
$$!(A \times B) \simeq !A \hat{\otimes} !B$$

- Classicality is about a linear involutive negation :

$$A^\perp := A \multimap \perp$$
$$A^{\perp\perp} \simeq A$$
$$A' := \mathcal{L}(A, \mathbb{R})$$
$$A \simeq A''$$

Just a glimpse at Differential Linear Logic

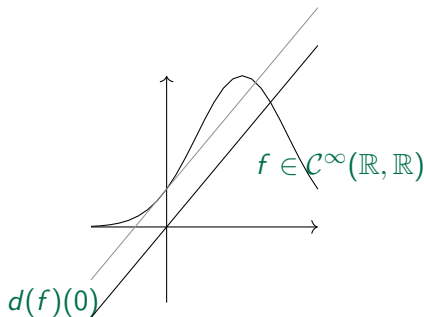
Differential Linear Logic

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

A linear proof is in particular non-linear.

$$\frac{f : !A \vdash B}{D_0(f) : A \vdash B} \bar{d}$$

From a non-linear proof we can extract a linear proof



Getting a smooth model of classical Differential Linear Logic ?

Smoothness

Spaces : E is a **locally convex** and **Hausdorff** topological vector space.

Functions: $f \in C^\infty(\mathbb{R}^n, \mathbb{R})$ is infinitely and everywhere differentiable.

These two requirements work as opposite forces.

- ✓ Handling smooth functions : some **completeness**.
- ✓ Interpreting the involutive linear negation $(E^\perp)^\perp \simeq E$: **Reflexive spaces**.

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Convenient differential category Blute, Ehrhard Tasson Cah. Geom. Diff. (2010)



Mackey-complete spaces and Power series, K. and Tasson, MSCS 2016.

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A model of LL with Schwartz' epsilon product, Dabrowski and K., 2018.



A logical account for PDEs, K., LICS18 [A polarized solution, no higher-order]



Higher-Order Distributions, Lemay and K., Fossacs19

Exponential : from ressources to distributions

- ▶ Linear Logic has long been interpreted in terms of discrete models and resource consumption.

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$$\begin{aligned} !A \multimap \perp &= A \Rightarrow \perp \\ \mathcal{L}(!E, \mathbb{R}) &\simeq \mathcal{C}^\infty(E, \mathbb{R}) \\ (!E)'' &\simeq \mathcal{C}^\infty(E, \mathbb{R})' \\ !E &\simeq \mathcal{C}^\infty(E, \mathbb{R})' \end{aligned}$$

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- ▶ The space of **distributions with compact support** $\mathcal{E}'(\mathbb{R}^n) := \mathcal{C}^\infty(\mathbb{R}^n, \mathbb{R})'$, whose elements are for example :

$$\phi_f : g \in \mathcal{C}^\infty(\mathbb{R}^n, \mathbb{R}) \mapsto \int fg.$$

$$\delta_x : g \mapsto g(x)$$

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- ▶ LL and Distribution Theory enjoy the same computing principle **same computing principles** : *Seely's isomorphisms are Kernel theorems*.

$$!A \otimes !B \simeq !(A \times B) \quad \mathcal{C}^\infty(E, \mathbb{R})' \hat{\otimes} \mathcal{C}^\infty(F, \mathbb{R})' \simeq \mathcal{C}^\infty(E \times F, \mathbb{R})' .$$

Which category of tvs should interpret formulas ?

Reflexive spaces enjoy poor stability properties.

- ▶ It is typically *not* preserved by \otimes .
- ▶ Nor by $\mathcal{L}(-, -)$.

Reflexivity takes many forms :

- ▶ It depends of the topology $E'_\beta, E'_c, E'_w, E'_\mu$ on the dual.
- ▶ The dual is not reflexive : one cannot close by bidual as with biorthogonals.

Monoidal closedness does not extends easily to the topological case :

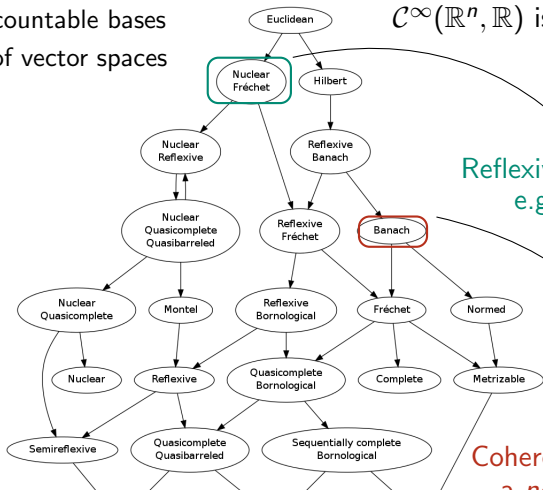
- ▶ Many possible topologies on \otimes : $\otimes_\beta, \otimes_\pi, \otimes_\varepsilon$.
- ▶ $\mathcal{L}_B(E \otimes_B F, G) \simeq \mathcal{L}_B(E, \mathcal{L}_B(F, G))$
 \Leftrightarrow "Grothendieck problème des topologies".

Topological models of DiLL

[Ehr02] [Ehr05] [DE08]

countable bases
of vector spaces

$C^\infty(\mathbb{R}^n, \mathbb{R})$ is not finite dimensional

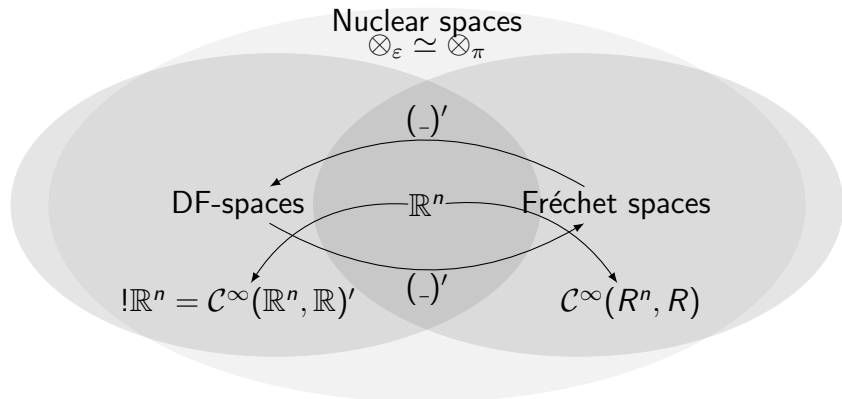


Reflexive and complete :
e.g. $C^\infty(\mathbb{R}^n, \mathbb{R})$

Coherent Banach spaces [Girard9]
a norm is too restrictive

Polarized model of Smooth differential Linear Logic [K.18]

Typical Nuclear Fréchet spaces are spaces of [smooth, holomorphic, rapidly decreasing ...] functions.



What about $\mathcal{C}^{\infty}(!\mathbb{R}^n, \mathbb{R})$ or $!!\mathbb{R}^n$?

Constructing some notion of smoothness which leaves stable the class of reflexive topological vector space.

We tackle this issue through the space of distribution

Consider E a topological vector space.

- ▶ Define an order on linear injections $f : \mathbb{R}^n \hookrightarrow E$ by $f \leq g := \exists \iota : \mathbb{R}^n \hookrightarrow \mathbb{R}^m, f = g \circ \iota$.
- ▶ Define the action of a distribution on E with respect to these linear injections:

$$\mathcal{E}'(E) := \varinjlim_{f: \mathbb{R}^n \rightarrow E} \mathcal{E}'(\mathbb{R}^n)$$

directed under the inclusion maps defined as

$$S_{f,g} : \mathcal{E}'(\mathbb{R}^n) \rightarrow \mathcal{E}'(\mathbb{R}^m), \phi \mapsto (h \mapsto \phi(h \circ \iota_{n,m}))$$

This is similar to work on C^∞ -algebras [KainKrieglMichor87], which we need to refine to obtain reflexivity.

A good inductive limit

Because the distributions spaces with which we build the inductive limit are extremely regular, we have

- ▶ $\mathcal{E}'(E)$ is always reflexive.
 - \rightsquigarrow weakly quasi-complete : $E = E''$ algebraically.
 - \rightsquigarrow barrelled $E \simeq E''$ topologically.
- ▶ $\mathcal{E}'(E)$ is the dual of a projective limit of spaces of functions :

$$\mathcal{E}(E) := \varprojlim_{f: \mathbb{R}^n \rightarrow E} \mathcal{E}_f(\mathbb{R}^n)$$

$\phi \in \mathcal{E}'(E)$ acts on $\mathbf{f} = (\mathbf{f}_f)_{f: \mathbb{R}^n \rightarrow E}$.

where $\mathbf{f}_f \in \mathcal{C}^\infty(\mathbb{R}^n, \mathbb{R})$.

The Kernel Theorem lifts to Higher-Order :

$$\mathcal{E}(E) \hat{\otimes} \mathcal{E}(F) \simeq \mathcal{E}(E \oplus F)$$

Reflexivity is enough for the structural morphisms

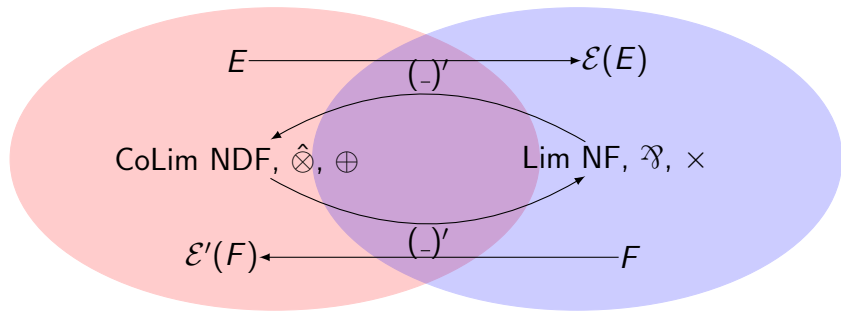
Because we worked with reflexive spaces at the beginning, we can build natural transformations :

$$d_E : \left\{ \begin{array}{l} !(E) \rightarrow E'' \simeq E \\ \phi \mapsto \underbrace{(\ell)}_{E \multimap \mathbb{R}} \in E' \mapsto \underbrace{\phi[(\ell \circ f)_{f: \mathbb{R}^n \multimap E}]}_{\mathbb{R}} \in \mathcal{E}(E) \end{array} \right.$$

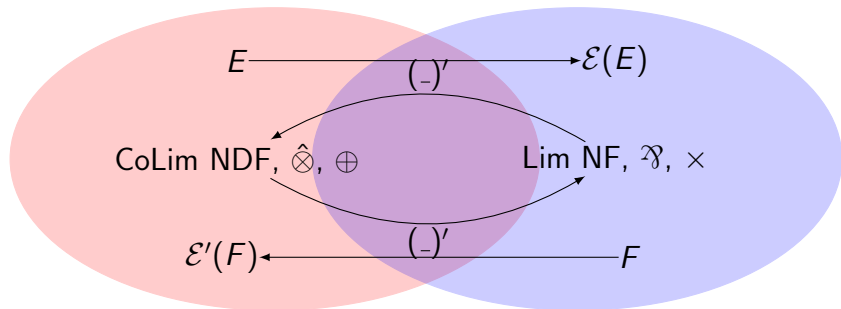
$$\bar{d}_E : \left\{ \begin{array}{l} E \rightarrow !E \simeq (\mathcal{E}(E))' \\ x \mapsto ((\mathbf{f}_f)_{f: \mathbb{R}^n \multimap E'}) \mapsto D_0 \mathbf{f}_f(f^{-1}(x)) \\ \text{where } f \text{ is injective such that } x \in \text{Im}(f) . \end{array} \right.$$

And interpretations for (co)-weakening and (co)-contraction follow from the Kernel Theorem.

We have obtain polarized model of Differential Linear Logic :



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... without promotion

We don't have a Cartesian Closed Category

This definition gives us functoriality only on isomorphisms :

$$! : \begin{cases} \mathbf{REFL}_{iso} \rightarrow \mathbf{REFL}_{iso} \\ E \mapsto \mathcal{E}'(E) \\ \ell : E \multimap F \mapsto !\ell \in \mathcal{E}(F') \end{cases}$$

where

$$(!\ell)(\phi)(\mathbf{g}) = \phi(\underbrace{(\mathbf{g}\ell \circ f)}_{\mathbb{R}^n \hookrightarrow F})_{f: \mathbb{R}^n \hookrightarrow E}.$$

No category with smooth functions as maps.

We have however a good candidate to make a co-monad of our functor.

$$\mu_E : \begin{cases} !E \rightarrow !!E \\ \phi \mapsto \left((\mathbf{g}_g)_g \in \mathcal{E}(!E) \simeq \lim C_g^\infty(\mathbb{R}^m) \right) \mapsto \mathbf{g}_g(g^{-1}(\phi)) \end{cases}$$

Conclusion

What we have : A Higher-Order exponential extending the notion of distributions, which interpret classical Differential Linear Logic without promotion.

$$\mathcal{E}'(E) := \varinjlim_{f: \mathbb{R}^n \rightarrow E} \mathcal{E}'_f(\mathbb{R}^n)$$

Perspectives :

- ▶ Linearity / Non-linearity , Solution / Parameter, Positive / Negative :
 - \rightsquigarrow give a categorical structure to the several interactions at stakes.
- ▶ Lifting this exponential to a co-monad:
 - \rightsquigarrow finer handling of indexations.
- ▶ Constructing exponentials via methods from Numerical Analysis :
 - $\rightsquigarrow !E = \overline{\langle \delta_x, x \in E \rangle}$ [BET12]
 - \rightsquigarrow Cut-elimination through Numerical Schemes.

Computing in Higher-Dimension - Computing Solutions

- ▶ If we wanted only smoothness and no reflexivity, we could have used :

$$!E = \overline{\langle \delta_x, x \in E \rangle}$$

By Frölicher and Kriegl, as used by Blute, Ehrhard and Tasson.

That's a *discretisation scheme*.

- ▶ In [K18] we showed that *cut-elimination is the resolution of certain class of differential equations* for which we have an explicit one-step resolution .
 - ↪ generalize to partial differential equations with no explicit solution.

Let's embed numerical schemes into cut-elimination.