

Séminaire Cosynus, LIX

Typing Differentiable Programming

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Differentiable programming

Definition: programming with differential transformations.

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[Abadi Plotkin POPL20]

[Brunel Mazza Pagani POPL20]

[Elliot ICFP18]

[Wang and al. ICFP 19]

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”De préférence typé”

Our work centers on finding a good type system for differentiable programming, typing a higher order differential transformation.

Lecun VS Logicians

Gödel Dialectica Transformation is Differentiable Programming.

Lecun VS Linear Logicians

Differential Linear Logic types a language expressing both forward and backward differentiation.

Curry-Howard-Lambek

Programs

Term

 $\lambda x^A.t^B$

Type

Execution

Logic

Proof

$$\frac{\vdots}{A \vdash B}$$

Formulas

Cut - elimination

Categories

Morphisms

 $f : A \rightarrow B$

Objects

Equality



In a future far far away : type theory allows to reason on basic computer algebra algorithms

Syntactical models (Pédrot)

Programs

Term

$\lambda x^A. t^B$

Type

Execution

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Proof

$$\frac{\vdots}{A \vdash B}$$

Formulas

Cut - elimination

Dialectica

Other Proofs

$$\frac{\vdots}{A \vdash B}$$

Other Formulas

Equivalence

In a future far far away : type theory allows to reason on basic computer algebra algorithms

Smooth models (K.)

Programs

Term

 $\lambda x^A . t^B$

Type

*Execution***Logic**

Proof

$$\frac{\vdots}{A \vdash B}$$

Formulas

*Cut - elimination***Analysis**

Smooth maps

 $f : A \rightarrow B$

Spaces

Equality



In a future far far away : type theory allows to reason on basic computer algebra algorithms

Preliminaries

- ▶ Automatic Differentiation.
- ▶ Linear Logic.
- ▶ Differential λ -calculus.

Automatic Differentiation

How does one compute the differentiation of an algebraic expression, computed as a sequence of elementary operations ?

$$\begin{array}{l} \text{E.g. : } z = y + \cos(x)^2 \\ x_1 = x_0^2 \\ x_2 = \cos(x_0) \\ z = y + x_2 \end{array} \quad \begin{array}{l} x'_1 = 2x_0x'_0 \\ x'_2 = -x'_0 \sin(x_0) \\ z' = y' + 2x_2x'_2 \end{array}$$

The computation of the final results requires the computation of the derivative of all partial computation. But in which order ?

Forward Mode differentiation: $(x_1, x'_1) \rightarrow (x_2, x'_2) \rightarrow (z, z')$.

Reverse Mode differentiation: $x_1 \rightarrow x_2 \rightarrow z \rightarrow z' \rightarrow x'_2 \rightarrow x'_1$
while keeping formal the unknown derivative.

AD from a higher-order functional point of view

$$\begin{aligned}D_u(f \circ g)(v) &= D_{g(u)}f(D_u f(v)) \\ D_u(f \circ g) &= D_{g(u)}f \circ D_u(f)\end{aligned}$$

► **Forward Mode differentiation:**

$$g(u) \rightarrow D_u g \rightarrow f(g(u)) \rightarrow D_{g(u)} f \rightarrow D_{g(u)} f \circ D_u(f).$$

► **Reverse Mode differentiation:**

$$g(u) \rightarrow f(g(u)) \rightarrow D_{g(u)} f \rightarrow D_u g \rightarrow D_{g(u)} f \circ D_u(f)$$

The choice of an algorithm is due to complexity considerations:

- Forward mode for $f : \mathbb{R} \rightarrow \mathbb{R}^n$.
- Reverse mode for $f : \mathbb{R}^n \rightarrow \mathbb{R}$

Linear logic

Usual Implication

A call-by-name translation

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

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

Linear logic

Usual implication

Linear Implication

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Linear logic

Usual implication

Linear implication

A call-by-name translation

$$A \Rightarrow B = !A \multimap B$$
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Exponential

Smooth Semantics

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

Differential λ -calculus [Ehrhard Regnier. 2004]

Inspired by denotational models of Linear Logic in vector spaces of sequences, it introduces a differentiation of λ -terms.

$D(\lambda x.t)$ is the linearization of $\lambda x.t$, it substitute x linearly, and then it remains a term t' where x is free.

Syntax:

$$\begin{aligned} \Lambda^d : S, T, U, V ::= 0 \mid s \mid s+T \\ \Lambda^s : s, t, u, v ::= x \mid \lambda x.s \mid sT \mid \mathbf{D}s \cdot t \end{aligned}$$

Operational Semantics:

$$\begin{aligned} (\lambda x.s)T &\rightarrow_{\beta} s[T/x] \\ \mathbf{D}(\lambda x.s) \cdot t &\rightarrow_{\beta_D} \lambda x. \frac{\partial s}{\partial x} \cdot t \end{aligned}$$

where $\frac{\partial s}{\partial x} \cdot t$ is the linear substitution of x by t in s .

The linear substitution ...

... which is not exactly a substitution

$$\frac{\partial y}{\partial x} \cdot T = \begin{cases} T & \text{if } x = y \\ 0 & \text{otherwise} \end{cases} \quad \frac{\partial 0}{\partial x} \cdot T = 0$$

$$\frac{\partial}{\partial x}(\lambda y \cdot s) \cdot T = \lambda y \cdot \frac{\partial s}{\partial x} \cdot T \quad \frac{\partial}{\partial x}(s + U) \cdot T = \frac{\partial s}{\partial x} \cdot T + \frac{\partial U}{\partial x} \cdot T$$

Differentiating composition:

$$\frac{\partial}{\partial x}(su) \cdot v = \left(\frac{\partial s}{\partial x} \cdot T\right)u + \dots$$

If x is linear in u , it is not linear in su

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Differentiating composition:

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But x can be free in v . In that case, we do what we would have done in differential geometry :

The linear substitution ...

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Differentiating composition:

$$\frac{\partial}{\partial x}(su) \cdot v = \left(\frac{\partial s}{\partial x} \cdot T\right)u + (Ds \cdot \left(\frac{\partial u}{\partial x} \cdot v\right)u)$$

Remember : We reverse the notations.

$$\frac{\partial f \circ g}{\partial x} v = D_{(g(v))} f \left(\frac{\partial g}{\partial x}(v) \right)$$

The linear substitution ...

... which is not exactly a substitution

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Differentiating composition:

$$\frac{\partial}{\partial x}(su) \cdot v = \left(\frac{\partial s}{\partial x} \cdot T\right)u + \underbrace{(\mathbb{D}s \cdot)}_{\text{Linear application}} \underbrace{\left(\frac{\partial u}{\partial x} \cdot v\right)}_{\text{Linear substitution}} u$$

Dialectica : Gödel doing Deep Learning

A Dialectica Transformation

Gödel Dialectica transformation [1958] : a translation from intuitionistic arithmetic to primitive recursive arithmetic.

$$A \rightsquigarrow \exists u : \mathbb{W}(A), \forall x : \mathbb{C}(A), A^D[u, x]$$

DePaiva [1991]: the linearized Dialectica translation operates on Linear Logic (types) and λ -calculus (terms).

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[Pedrot, CLS-LICS2014]

A linearized Dialectica translation preserving β -equivalence, via the introduction of an "abstract multiset constructor" on types on the target.

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A linearized Dialectica translation preserving β -equivalence, via the introduction of an "abstract multiset constructor" on types on the target.

\rightsquigarrow Dialectica as a program translation ... whose abstract multiset is *not smooth enough*

Pédrot Dialectica Transformation

At the source : λ -calculus typed with minimal logic.

At the target : λ -calculus with pairs and an \mathfrak{M} operation.

$$\begin{aligned} \mathbb{W}(\alpha) &:= \alpha_{\mathbb{W}} \\ \mathbb{C}(\alpha) &:= \alpha_{\mathbb{C}} \\ \mathbb{W}(A \Rightarrow B) &:= (\mathbb{W}(A) \Rightarrow \mathbb{W}(B)) \times (\mathbb{W}(A) \Rightarrow \mathbb{C}(B) \Rightarrow \mathfrak{M} \mathbb{C}(A)) \\ \mathbb{C}(A \Rightarrow B) &:= \mathbb{W}(A) \times \mathbb{C}(B) \end{aligned}$$

$$\begin{aligned} x_x &:= \lambda\pi. \{\pi\} & x^\bullet &:= x \\ x_y &:= \lambda\pi. \emptyset \text{ if } x \neq y & (\lambda x.t)^\bullet &:= (\lambda x.t^\bullet, \lambda x\pi.t_x \pi) \\ (\lambda x.t)_y &:= \lambda\pi. (\lambda x.t_y) \pi.1 \pi.2 & (t u)^\bullet &:= (t^\bullet.1) u^\bullet \end{aligned}$$

$$(t u)_y := \lambda\pi. (t_y (u^\bullet, \pi)) \otimes ((t^\bullet.2) u^\bullet \pi \ggg u_y)$$

Pédrot Dialectica Transformation

At the source : λ -calculus typed with minimal logic.

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Soundness [Ped14]

If $\Gamma \vdash t : A$ in the source then we have in the target

- ▶ $\mathbb{W}(\Gamma) \vdash t^\bullet : \mathbb{W}(A)$
- ▶ $\mathbb{W}(\Gamma) \vdash t_x : \mathbb{C}(A) \Rightarrow \mathfrak{M}\mathbb{C}(X)$ provided $x : X \in \Gamma$.

Tracking differentiation in Dialectica

$$\frac{}{\Gamma \vdash \emptyset : \mathfrak{M} A} \quad \frac{\Gamma \vdash m_1 : \mathfrak{M} A \quad \Gamma \vdash m_2 : \mathfrak{M} A}{\Gamma \vdash m_1 \otimes m_2 : \mathfrak{M} A}$$

$$\frac{\Gamma \vdash t : A}{\Gamma \vdash \{t\} : \mathfrak{M} A} \quad \frac{\Gamma \vdash m : \mathfrak{M} A \quad \Gamma \vdash f : A \Rightarrow \mathfrak{M} B}{\Gamma \vdash m \ggg f : \mathfrak{M} B}$$

$$x_x \quad := \quad \lambda \pi. \{\pi\}$$

$$x^\bullet \quad := \quad x$$

$$x_y \quad := \quad \lambda \pi. \emptyset \quad \text{if } x \neq y$$

$$(\lambda x. t)^\bullet \quad := \quad (\lambda x. t^\bullet, \lambda x \pi. t_x \pi)$$

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Differential λ -calculus

$$x_x := \lambda \pi. \{\pi\} \quad x^\bullet := x$$

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$$(\lambda x. t)_y \quad := \quad \lambda \pi. (\lambda x. t_y) \pi.1 \pi.2 \qquad (t u)^\bullet \quad \equiv \quad (\lambda x. (tx)^\bullet) u^\bullet$$

$$(t u)_y := \lambda \pi. (t_y (u^\bullet, \pi)) \otimes ((t^\bullet.2) u^\bullet \pi \gg= u_y)$$

Backpropagation

Differential λ -calculus in a hurry

$$D(\lambda x.s) \cdot t \rightarrow_{\beta_D} \lambda x. \frac{\partial s}{\partial x} \cdot t$$

$$\frac{\partial y}{\partial x} \cdot T = \begin{cases} T & \text{if } x = y \\ 0 & \text{otherwise} \end{cases} \quad \frac{\partial}{\partial x}(sU) \cdot T = \left(\frac{\partial s}{\partial x} \cdot T\right)U + (Ds \cdot \left(\frac{\partial U}{\partial x} \cdot T\right))U$$

$$\frac{\partial}{\partial x}(\lambda y.s) \cdot T = \lambda y. \frac{\partial s}{\partial x} \cdot T \quad \frac{\partial}{\partial x}(Ds \cdot u) \cdot T = D\left(\frac{\partial s}{\partial x} \cdot T\right) \cdot u + Ds \cdot \left(\frac{\partial u}{\partial x} \cdot T\right)$$

$$\frac{\partial 0}{\partial x} \cdot T = 0 \quad \frac{\partial}{\partial x}(s + U) \cdot T = \frac{\partial s}{\partial x} \cdot T + \frac{\partial U}{\partial x} \cdot T$$

Dialectica is Differentiation

The linearized Dialectica Translation weakens to a transformation from λ -calculus to Differential λ -calculus.

Differential calculus is typed with minimal logic and does not distinguish a specific types on which the formal sum $*$ applies :

$$\llbracket \emptyset \rrbracket := 0 \quad \llbracket \mathfrak{M} A \rrbracket = A \quad \llbracket t \otimes u \rrbracket := \llbracket t \rrbracket + \llbracket u \rrbracket \quad \llbracket \{t\} \rrbracket := \llbracket t \rrbracket.$$

Proposition

Consider two λ -terms t and u . Then $\llbracket t_x \rrbracket u \equiv \frac{\partial t}{\partial x} \cdot u$ and $((\lambda x.t) \bullet .2)u \equiv Dt \cdot u$.

Dialectica enriched with real functions

We now enrich both our source and target λ -calculi with a type of reals \mathbb{R} . We assume furthermore that the source contains functions symbols $\varphi, \psi, \dots : \mathbb{R} \rightarrow \mathbb{R}$ with derivative φ', ψ', \dots

$$\mathbb{W}(\mathbb{R}) := \mathbb{R} \qquad \mathbb{C}(\mathbb{R}) := 1$$

$$\varphi^\bullet := (\varphi, \lambda\alpha\pi. \{() \mapsto \varphi'(\alpha)\}) \quad \varphi_x := \lambda\pi. \emptyset$$

The soundness theorem is then adapted trivially.

Soundness Theorem

The following equation holds in the target.

$$(\varphi_1 \circ \dots \circ \varphi_n)^\bullet \cdot 2 \alpha () \equiv \{() \mapsto (\varphi_1 \circ \dots \circ \varphi_n)'(\alpha)\}$$

Dialectica is Backpropagation

When one distinguishes a *specific types for the codomain of functions*, on which the sums operate, we observe a cut-elimination mimicking the dynamic of backward differentiation.

$$A, B := \alpha \mid A \Rightarrow B \mid A \times B \mid A^\perp \mid \text{Tr}(A)$$

$$t, u := x \mid (t)u \mid \lambda x.t \mid (t, u) \mid t \mid u \otimes v \mid \emptyset.$$

Types at the source : Minimal Logic and a type of Traces).

$$\frac{\Gamma \vdash t : A}{\Gamma \vdash \{t\} : \text{Tr}(A)}$$

$$\frac{\Gamma \vdash t : \text{Tr}(A) \quad \Gamma \vdash u : \text{Tr}(A)}{\Gamma \vdash t \otimes u : \text{Tr}(A)}$$

$$\frac{}{\Gamma \vdash \emptyset : \text{Tr}(A)}$$

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Two mutually inductively defined translations :

$$\begin{array}{ll}
 x_x & := \lambda \pi. \{\pi\} & x^\bullet & := x \\
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$$(t u)_y := \lambda \pi. (t_y (u^\bullet, \pi)) \otimes ((t^\bullet.2) u^\bullet \pi \gg= u_y)$$

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 (\lambda x.t)_y & := \lambda \pi. (\lambda x.t_y) \pi.1 \pi.2 & (t u)^\bullet & := (t^\bullet.1) u^\bullet
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Two typed differential transformations

When $\Gamma \vdash t : !A \multimap B$ and writing $Dt = (t^\bullet.2)$ we have:

$$\Gamma \vdash Dt : A \Rightarrow (B^\perp \Rightarrow \text{Tr}(A^\perp))$$

$$\Gamma \vdash t_y : A \times B^\perp \Rightarrow \text{Tr}(Y^\perp)$$

Dialectica is Backpropagation

We reuse the arguments of Brunel, Mazza and Pagani:
Backpropagation is encoded through the contravariance of the differential arguments, which is typed by a linear dual.

Consider $f : \mathbb{R}^n \rightarrow \mathbb{R}^m$.

$$\vec{D}(f) : \begin{cases} \mathbb{R}^n \times \mathbb{R}^m \times x \rightarrow \mathbb{R}^n \times \mathbb{R}^m \\ (a, x) \mapsto (f(a), D_a f \cdot x) \end{cases}$$

$$\overleftarrow{D}(f) : \begin{cases} \mathbb{R}^n \times \mathbb{R}^{m^\perp} \rightarrow \mathbb{R}^m \times \mathbb{R}^{n^\perp} \\ (a, x) \mapsto (f(a), (v \mapsto v \cdot (D_a f \cdot x))) \end{cases}$$

- ▶ As in differential λ-calculus, the use of two separate differential transformation allows to go higher-order.

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Consider $f : E \rightarrow F$.

$$\overleftarrow{D}(f) : \begin{cases} E \times F' \rightarrow F \times E' \\ (a, \ell) \mapsto (f(a), (v \in F \mapsto (v \cdot (D_a f \cdot x)))) \end{cases}$$

- ▶ As in differential λ -calculus, the use of two separate differential transformation allows to go higher-order.

Lessons from Dialectica

- ▶ As in differential λ -calculus, the use of two distinct transformations allows to handle the differentiation of higher-order functions.
- ▶ As in [BMP20], encoding partial substitutions by *Linear duals* allow the encoding of backpropagation.
- ▶ This gives us a differential translation which can be enriched over dependant or positive types.
- ▶ Hint: call-by-name agrees with backpropagation.

\rightsquigarrow towards a finer, internal handling of automatic differentiation as a reduction strategy.

Automatic Differentiation as a choice of reduction strategy

Refining λ -calculus with operations from distribution theory.

Just a glimpse at Differential Linear Logic

$$A, B := A \otimes B \mid !A \wp B \mid \perp \mid A \oplus B \mid 0 \mid A \times B \mid \top \mid !A \mid !A$$

Exponential rules of DiLL₀

$$\frac{\vdash \Gamma, ?A, ?A}{\vdash \Gamma, ?A} c$$

$$\frac{\vdash \Gamma}{\vdash \Gamma, ?A} w$$

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

$$\frac{\vdash \Gamma, !A, \quad \vdash \Delta, !A}{\vdash \Gamma, \Delta, !A} \bar{c}$$

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

$$\frac{\vdash \Gamma, A}{\vdash \Gamma, !A} \bar{d}$$

$$\frac{\vdash ?\Gamma, A}{\vdash ?\Gamma, !A} p$$

\rightsquigarrow *A particular point of view on differentiation induced by duality.*



Normal functors, power series and λ -calculus. Girard, APAL(1988)



Differential interaction nets, Ehrhard and Regnier, TCS (2006)

Exponentials are distributions

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

functions

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

distributions

A typical distribution is the dirac operator:

$$\delta : \begin{cases} E \rightarrow \mathcal{C}^\infty(E, \mathbb{R})' \\ x \mapsto (\phi \mapsto \phi(x)) \end{cases}$$

Exponential rules of DiLL₀

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

$$\frac{\vdash \Gamma}{\vdash \Gamma, cst_0 : ?A} w$$

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

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

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

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

$$\frac{\vdash ?\Gamma, v : A}{\vdash ?\Gamma, \delta_v : !A} p$$

A few operations typed by DILL

The composition of linear functions:

$$\frac{\Gamma \vdash f : A \multimap B \quad \Delta \vdash g : B \multimap C}{\Gamma, \Delta \vdash g \circ f : A \multimap C} \text{ cut}$$

The composition of non-linear functions:

$$\frac{\frac{\Gamma \vdash f : !A \multimap B}{\Delta \vdash (x \mapsto \delta_{g(x)}) : !A \multimap !B} \text{ P} \quad \Delta \vdash g : !B \multimap C}{\Gamma, \Delta \vdash g \circ f = (x \mapsto \delta_{f(x)}g) : !A \multimap C} \text{ cut}$$

The Differentiation of non-linear functions:

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

Let's translate this into a term language typed by DILL.

A few operations typed by DILL

The chain rule is encoded in the interaction of diracs δ_x with differential arguments D_{ut} .

The composition of non-linear functions:

$$\frac{\frac{\Gamma \vdash f : !A \multimap B}{\Delta \vdash (x \mapsto \delta_{g(x)}) : !A \multimap !B} \text{P} \quad \Delta \vdash g : !B \multimap C}{\Gamma, \Delta \vdash g \circ f = (x \mapsto \delta_{f(x)}g) : !A \multimap C} \text{cut}$$

The Differentiation of non-linear functions:

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

Let's translate this into a term language typed by DILL.

A few operations typed by DiLL

The chain rule is encoded in the interaction of diracs δ_x with differential arguments $D_u t$.

The Chain rule:

$$\frac{\frac{\Gamma \vdash f : !A \multimap B}{\Delta \vdash (x \mapsto \delta_{g(x)}) : !A \multimap !B} \text{p} \quad \Delta \vdash g : !B \multimap C}{\Gamma, \Delta \vdash g \circ \delta_f : !A \multimap C} \text{cut} \quad \frac{\vdash \Delta', v : A}{\vdash \Gamma, D_0(-)(v) : !A} \bar{d}}{\Gamma, \Delta, \Delta' \vdash D_0(g \circ f)(v) : c} \text{cut}$$

Let's translate this into a term language typed by DiLL.

A minimal language allowing to express automatic differentiation

Two class of terms:

$$u, v := x \mid t^\perp \mid u * v \mid \emptyset \mid u \otimes v \mid 1 \mid \delta_u \mid D_u(t) \mid \downarrow t$$

$$t, s := u^\perp \mid t \cdot s \mid w_1 : N \mid \lambda x.t \mid dx.t \mid \uparrow u$$

A function $\lambda x.t$ can be match to two kind of arguments: diracs δ_u or differential operators $D_u t$.

$$(\lambda x.t)\delta_u \rightarrow t[u/x]$$

$$(\lambda x.t)D_t u \rightarrow \dots$$

The differentiation $\lambda x.t$ of must be inductively defined on t :

$$(\lambda x.(t)u)D_w s \rightarrow \uparrow(\downarrow((\lambda x.t)D_w s)u * \downarrow(t((\lambda x.u)D_w s)))$$

Differentiating an application $(t)u$ is symmetric in t and u .

$$(\lambda x.\uparrow\delta_t)D_u s \rightarrow (\lambda z.\uparrow(D_z((\lambda x.t)D_u s))((\lambda x.t)(u)))$$

The abstraction $\lambda x.\uparrow\delta_t$ will be composed with another abstraction and differentiation must take that into account.

Forward / Backward Differentiation as CBV/CBN

Then the differentiation of $(\lambda y.s) \circ (\lambda x.t)$ at a point $u = \delta_w$ according to a vector r computes as follows:

$$\begin{aligned}
 (\lambda x.((\lambda y.s)\delta_t))D_u r &\rightarrow \uparrow(\downarrow((\lambda x.(\lambda y.s))D_u r)\delta_t * \downarrow((\lambda y.s)((\lambda x.\delta_t)D_u r))) \\
 &\rightarrow^* \uparrow(\downarrow(\uparrow\emptyset) * \downarrow((\lambda y.s)((\lambda x.\delta_t)D_u r))) \text{ as } x \text{ is free in } s \\
 &\rightarrow^* (\lambda y.s)((\lambda x.\delta_t)D_u r) \text{ by involutivity of the shifts} \\
 &\rightarrow (\lambda y.s)(\lambda z.\uparrow(D_z((\lambda x.t)D_u r))((\lambda x.t)(u))) \\
 &\rightarrow ((\lambda y.s)(\lambda z.\uparrow(D_z((\lambda x.t)D_u r))))((t[w/x])) \text{ as } u = \delta_w \\
 &\rightarrow^* (\lambda y.s)D_v((\lambda x.t)D_u r) \text{ if } (t[w/x] \rightarrow^* \delta_v)
 \end{aligned}$$

The value of $t[w/x]$ is computed first-hand. Whether we proceed with the computation of the derivative of the first function $((\lambda x.t)D_u r)$ or to the derivative of the second $((\lambda y.s)D_v((\lambda x.t)D_u r))$ depends of the evaluation strategy.

Higher-order addition and Higher-order multiplication

Additions are done on the domain, through convolution (ie higher order addition).

$$\begin{aligned}\phi * \psi &:= f \mapsto \phi(x \mapsto \psi(y \mapsto f(x + y))) \\ \delta_u * \delta_v &\rightarrow \delta_{u*v}\end{aligned}$$

Multiplications are done on the codomain, through contractions (ie higher order multiplication).

$$\begin{aligned}f \cdot g &:= x \mapsto f(x) \cdot g(x) \\ (\lambda y.t) \cdot (\lambda z.s) &\rightarrow \lambda x.(t[x/y]) \cdot (s[x/z])\end{aligned}$$

Distinguishing Linear and Non-Linear Maps

$$\frac{\vdash \mathbb{N}, t : M, x^\perp : (!P)^\perp \mid}{\vdash \mathbb{N}, \lambda x.t : (!P)^\perp \wp M \mid} (\lambda)$$
$$\frac{\vdash \mathbb{N}, x^\perp : P^\perp, t : M}{\vdash \mathbb{N}, dx.t : (!P)^\perp \wp M} (d)$$

Interpreting Dialectica in DILL

$$[\mathfrak{M} A] := !![A]$$

$$[\lambda x.t] := \lambda x.[t]$$

$$[\emptyset] := \uparrow\emptyset$$

$$[u \otimes v] := \uparrow(\downarrow[u] * \downarrow[v])$$

$$[x] := x$$

$$[(t, u)] := ([t], [u])$$

$$[\{t\}] := \uparrow(\delta_{\delta_{[t]}})$$

$$[m \gg= f] := (dx.[f]x)[m]$$

A translation on top of Dialectica

If $\Gamma \vdash t : A$ in the target of Dialectica, then $:\mathbb{L}(\Gamma) \vdash [t] : \mathbb{L}(A)$ and if $t \equiv u$ in the target of Dialectica then $[t] \equiv [u]$ in our calculus.

A semantical point of view : if $\chi : C^\infty(E, F) \simeq \mathcal{L}(!E, F)$ then $(\delta_{\delta_e}) \gg= f := \chi(f)(\delta_e)$.

Conclusion

What we have:

- ▶ Dialectica is a reverse-mode differential transformation.
- ▶ Differential Linear Logic gives a type-system for a higher-order functional language, in which forward and reverse mode differentiation identity to reduction strategies.

What we would like to have:

- ▶ Higher-Order models.
- ▶ A merge between the two : an endo-transformation handling a rich type theory as well as forward or reverse differential transformation.
- ▶ A lighter use of shifts.

More on Dialectica

Monadic laws

$$\{t\} \gg= f \equiv f t \quad t \gg= (\lambda x. \{x\}) \equiv t$$

$$(t \gg= f) \gg= g \equiv t \gg= (\lambda x. f x \gg= g)$$

Monoidal laws

$$t \otimes u \equiv u \otimes t \quad \emptyset \otimes t \equiv t \otimes \emptyset \equiv t$$

$$(t \otimes u) \otimes v \equiv t \otimes (u \otimes v)$$

Distributivity laws

$$\emptyset \gg= f \equiv \emptyset \quad t \gg= \lambda x. \emptyset \equiv \emptyset$$

$$(t \otimes u) \gg= f \equiv (t \gg= f) \otimes (u \gg= f)$$

$$t \gg= \lambda x. (f x \otimes g x) \equiv (t \gg= f) \otimes (t \gg= g)$$