Computability in Europe 2023 Special session on Proof Theory

∂ is for Dialectica

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Gödel's Dialectica Transformation

▶ Gödel <u>Dialectica transformation</u> [1958] : a translation from intuitionistic arithmetic to a finite type extension of primitive recursive arithmetic.

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

- \triangleright De Paiva [1991]: the linearized Dialectica translation operates on Linear Logic (types) and λ-calculus (terms).
- ▶ Pedrot [2014] A computational Dialectica translation preserving β -equivalence, via the introduction of an "abstract multiset constructor" on types on the target.

Gödel's Dialectica

- 1. $(F \wedge G)' = (\exists yv) (zw) [A (y, z, x) \wedge B (v, w, u)].$
- 2. $(F \lor G)' = (\exists yvt) (zw) [t=0 \land A (y, z, x) \cdot \lor \cdot t=1 \land B (v, w, u)].$
- 3. $[(s) F]' = (\exists Y) (sz) A (Y (s), z, x)$.
- 4. $[(\exists s) F]' = (\exists sy) (z) A (y, z, x)$.
- 5. $(F \supset G)' = (\exists VZ) (yw) [A (y, Z (yw), x) \supset B (V (y), w, u)].$



Kurt Gödel (1958). Über eine bisher noch nicht benützte Erweiterung des finiten Standpunktes. Dialectica.

Gödel's Dialectica

- ► Validates semi-classical axioms:
 - ightharpoonup Markov's principle: $\neg\neg\exists xA\to\exists xA$ when A is decidable.
 - ▶ Independent of premises : $(A \to \exists xB) \to (\exists x.(A \to B))$
- ► Numerous applications :
 - ► Soudness results
 - ▶ Proof mining

A further distinguishing feature of the D-interpretation is its nice behavior with respect to modus ponens. In contrast to cut-elimination, which entails a global (and computationally infeasible) transformation of proofs, the D-interpretation extracts constructive information through a purely local procedure; when proofs of φ and $\varphi \to \psi$ are combined to yield a proof of ψ , witnessing terms for the antecedents of this last inference are combined to yield a witnessing term for the conclusion. As a result of this modularity, the interpretation of a theorem can be readily obtained from the interpretations of the lemmata used in its proof.



Jeremy Avigad and Solomon Feferman (1999). Gödel's functional ("Dialectica") interpretation

A peek into Dialectica interpretation of functions

$$(A \to B)_D = \exists fg \forall xy (A_D(x, gxy) \to B_D(fx, y))$$

Usual explanation: least unconstructive prenexation.

- ▶ Start from $\exists x, \forall u, A_D[x, u] \rightarrow \exists y, \forall v, B_D[y, v].$
- ▶ Obvious prenexation : $\forall x (\forall u, A_D[x, u] \rightarrow \exists y, \forall v, B_D[y, v])$
- ▶ Weak form of IP : $\forall x \exists y (\forall u, A_D[x, u] \rightarrow \forall v, B_D[y, v])$
- ▶ Prenexation : $\forall x \exists y, \forall v, \exists u (A_D[x, u] \rightarrow B_D[y, v]).$
- $\qquad \qquad \mathsf{Markov}: \, \forall x, \exists y, \forall v, \exists u (A_D[x,u] \to B_D[y,v])$
- ▶ Axiom of choice : $\exists f, \exists g, \forall u, \forall v, (A_D(u, guv) \rightarrow B_D[fu, v]).$

Dynamic behaviour: agrees to a chain rule.

Mathematical meaning: it's some kind of approximation.



Ulrich Kohlenbach, Applied Proof Theory: Proof Interpretations and their Use in Mathematics, 2008

Outline of the talk

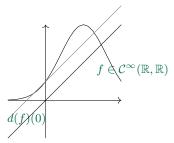
• The Historical Dialectica

- Differentiation and Differentiable Programming.
- Factorizing Dialectica through differential linear logic.
- Dialectica acting on λ -terms.
- Applications and related work.

Differentiable Programming

Differentiation

▶ Differentiation is finding the best linear approximation to a function at a point.

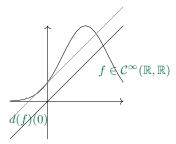


Chain Rule : $D_0(f \circ g) = D_{g(0)}f \circ D_0g$

- ▶ Differentiation is a mathematical operation which needs to be fitted to logical and computer science use.
 - Algorithmic Differentiation: differentiating sequences of many-valued functions efficiently.
 - ▶ Differential Linear Logic : Differentiating proofs and λ -terms

Differentiation

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Dialectica verifies the chain rule

Composing the Dialectica interpretation of arrows:

$$(A \Rightarrow B)_{D}[\phi_{1}; \psi_{1}, u_{1}; v_{1}] := A_{D}(u_{1}, \psi_{1} u_{1} v_{1}) \Rightarrow B_{D}(\phi_{1} u_{1}, v_{1})$$

$$(B \Rightarrow C)_{D}[\phi_{2}; \psi_{2}, u_{2}; v_{2}] := B_{D}(u_{2}, \psi_{2} u_{2} v_{2}) \Rightarrow C_{D}(\phi_{2} u_{2}, v_{2})$$

$$(A \Rightarrow C)_{D}[\phi_{3}; \psi_{3}, u_{3}; v_{3}] := A_{D}(u_{3}, \psi_{3} u_{3} v_{3}) \Rightarrow C_{D}(\phi_{3} u_{3}, v_{3})$$

The Dialectica interpretation amounts to the following equations:

$$u_3 = u_1$$
 $\psi_3, u_3, v_3 = \psi_1, u_1, v_1$
 $v_3 = v_2$ $\phi_2 u_2 = \phi_1, u_1$
 $u_2 = \phi_1 u_1$ $v_1 = \psi_2(u_2, v_2)$

which can be simplified to:

$$\phi_3(u_3) = \phi_2(\phi_1(u_3))$$
 composition of functions $\psi_3(u_3, v_3) = \psi_1(u_3, \psi_2(\phi_1 u_3, v_3))$ composition of their differentials

Thanks to T. Powell for noticing typos here.

But verifying the chain rule does not make you differentiation!

▶ More modern presentations of Dialectica.

▶ More Computer Science Friendly presentations of Differentiation.

Linearity must enter the game.

Curry-Howard for semantics

Programs	\mathbf{Logic}	Semantics		
fun $(x:A) \rightarrow (t:B)$	Proof of $A \vdash B$	$f: A \to B$.		
Types	Formulas	Objects		
Execution	Cut-elimination	Equality		
Dialectica				
Differential λ -calculus	Differential Linear Logic	Differential Categories		

 $Dialectica\ is\ Backward\ Differentiation\ in\ Logic$

And now for something completely different: Automatic Differentiation

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

E.g. :
$$z = y + cos(x^2)$$
 $x_1 = x_0^2$ $x_1' = 2x_0x_0'$ $x_2 = cos(x_1)$ $x_2' = -x_0'sin(x_0)$ $z = y + x_2$ $z' = y' + 2x_2x_2'$

Derivative of a sequence of instruction



sequence of instruction \times sequence of derivatives

Forward Mode differentiation [Wengert, 1964] $(x_1, x_1') \rightarrow (x_2, x_2') \rightarrow (z, z')$.

Reverse Mode differentiation: [Speelpenning, Rall, 1980s]

 $x_1 \to x_2 \to z \to z' \to x'_2 \to x'_1$ while keeping formal the unknown derivative.

Curry-Howard for semantics

The syntax mirrors the semantics.

Programs	\mathbf{Logic}	Semantics
fun $(x:A) \rightarrow (t:B)$	Proof of $A \vdash B$	$f: A \to B$.
Types	Formulas	Objects
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- Programs acts on programs.
 - ▶ Functions are higher-order: they act not only on \mathbb{R}^n , but also on $\mathcal{C}^{\infty}(\mathbb{R}^n, \mathbb{R})$.
- Programs are typed.
 - $Add: \mathcal{C}^{\infty}(\mathbb{R}^n, \mathbb{R}) \times \mathcal{C}^{\infty}(\mathbb{R}^n, \mathbb{R}) \to \mathcal{C}^{\infty}(\mathbb{R}^n, \mathbb{R})$
- ▶ Everything is interpreted in Categories.
 - ▶ Objects are Data
 - ► Functions are Programs
 - ► Transformations are functorial:

$$\mathcal{F}(p_1; p_2) = \mathcal{F}(p_1); \mathcal{F}(p_2)$$
$$\mathcal{F}(f_2 \circ f_1) = \mathcal{F}(f_2) \circ \mathcal{F}(f_1)$$

Back to AD: I hate graphs

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

- Forward Mode differentiation : $g(u) \to D_u g \to f(g(u)) \to D_{g(u)} f \to D_{g(u)} f \circ D_u(g)$.
- ▶ Reverse Mode differentiation: $g(u) \rightarrow f(g(u)) \rightarrow D_{g(u)}f \rightarrow D_{u}(g) \rightarrow D_{g(u)}f \circ D_{u}(g)$

The choice of an algorithm is due to complexity considerations:

- ▶ Forward mode for $f \circ g : \mathbb{R} \to \mathbb{R}^n$.
- ▶ Reverse mode for $f \circ g : \mathbb{R}^n \to \mathbb{R}$
- → Differentiable programming is a new research area triggered by the advances of deep learning algorithms on neural networks, it tries to attach two very old domains: lambda-calculus and automatic differentiation, with correctness and modularity goals in mind.

AD from a functorial point of view

$$\mathbf{D}_u(f \circ g) = \mathbf{D}_{\underline{g(u)}} f \circ \mathbf{D}_u(g)$$

Non-functorial !!!

How to make differentiation functorial? Make it act on pairs!

$$f:E\Rightarrow F$$

Forward Mode differentiation:

$$f: E \Rightarrow E \leadsto \overrightarrow{D}f: E \Rightarrow E \multimap F.$$

$$\overrightarrow{D}(f): \begin{cases} E \Rightarrow E \multimap F \\ u \mapsto v \mapsto D_u(f)(v) \end{cases}$$

Functorial forward differentiation:

$$(f, \overrightarrow{D}(f)): \begin{cases} E \times E \to F \times F \\ (a, x) \mapsto (f(a), (D_a f \cdot x)) \end{cases}$$

Reverse AD from a functorial point of view

How to make **reverse** differentiation functorial?

Make it act on pairs with linear duals!

Reverse functorial differentiation

Linear Dual $A^{\perp} \equiv A \multimap \perp \equiv \mathcal{L}(A, \mathbb{R})$

▶ Reverse Mode differentiation:

$$\begin{split} g(u) &\to f(g(u)) \to D_{g(u)} f \to D_{g(u)} f \circ D_u(g) \\ f &: E \Rightarrow F \leadsto \overleftarrow{D} f : E \Rightarrow F^\perp \Rightarrow E^\perp. \\ \overleftarrow{D}(f) &: \begin{cases} E \Rightarrow F^\perp \multimap E^\perp \\ u \mapsto \ell \mapsto \ell \circ D_u(f) \end{cases} \end{split}$$

[Mazza, Pagani, POPL2020]

► Reverse functorial differentiation

$$(f, \overleftarrow{D}(f)) : (E \Rightarrow F) \times (E \Rightarrow F^{\perp} \Rightarrow E^{\perp})$$

Reverse functorial differentiation

▶ Reverse Mode differentiation:

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[Mazza, Pagani, POPL2020]

▶ Reverse functorial differentiation :

$$(f, \overleftarrow{D}(f)) : (E \Rightarrow F) \times (E \Rightarrow F^{\perp} \Rightarrow E^{\perp})$$

Types!

Programs and variable are typed by logical formulas which describe their behavior

$$A \leadsto \exists x : \mathbb{W}(A), \forall \underline{u} : \mathbb{C}(A), A_D[x, u]$$

Witness and counter types:

$$\mathbb{C}(A \Rightarrow B) = \mathbb{C}(A) \times \mathbb{C}(B)$$

$$\mathbb{W}(A \Rightarrow B) = (\mathbb{W}(A) \Rightarrow \mathbb{W}(B)) \times (\mathbb{W}(A) \Rightarrow \mathbb{C}(B) \Rightarrow \mathbb{C}(A))$$

Reverse Mode differentiation

Functorial:
$$(h, \overleftarrow{D}h): (A \Rightarrow B) \times (A \Rightarrow B^{\perp} \multimap A^{\perp})$$

However:

- ► Having the same type does not mean you're the same program
- Some french (linear) logicians have a strong opinion on what proof differentiation should.

Types!

Programs and variable are typed by logical formulas which describe their behavior

$$A \leadsto \exists x : \mathbb{W}(A), \forall u : \mathbb{C}(A), A_D[x, u]$$
local opponent

Witness and counter for implication types:

$$\mathbb{C}(A \Rightarrow B) = \mathbb{C}(A) \times \mathbb{C}(B)$$

$$\mathbb{W}(A\Rightarrow B) = \overbrace{(\mathbb{W}(A)\Rightarrow\mathbb{W}(B))}^{\text{function}} \times \left(\mathbb{W}(A)\Rightarrow\underbrace{\mathbb{C}(B)\Rightarrow\mathbb{C}(A)}_{\text{reverse derivative}}\right)$$

Reverse Mode differentiation:

Functorial:
$$(h, \overleftarrow{D}h): (A \Rightarrow B) \times (A \Rightarrow B^{\perp} \multimap A^{\perp})$$

However:

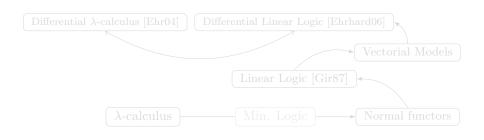
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A Linear Logic Refinement

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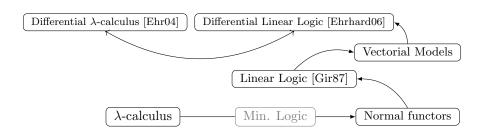


Doing to proofs everything we do to functions

Curry-Howard for semantics

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Doing to proofs everything we do to functions

Linear Logic

Usual Implication

←

Linear and Non Linear Arrows

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

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

- ► Notions of ressources which have made their way into programmation through linear types.
- ► The dynamics of linearity gets encoded through the rules of the ! connective, and its dual ?.

$$A, B := A \otimes B | A \Im B | A \oplus B | A \& B | !A | ?A$$

Linear Logic

Usual implication

←

Linear and Non Linear Arrows

Linear Implication

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Linear Logic

Usual implication Linear Arrows

⋆ Linear Implication

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

Exponential -

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

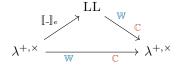
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$$A,B := A \otimes B|A \Im B|A \oplus B|A \& B|!A|?A$$

Dialectica factorizes through Linear Logic

The call by name arrow

$$A \Rightarrow B := !A \multimap B := ((!A) \otimes B^{\perp})^{\perp}$$



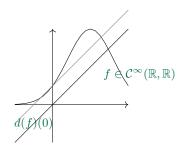


Valeria de Paiva, 1989, A dialectica-like model of linear logic.

Differential Linear Logic

$$\begin{array}{c} \vdash \ell : A \multimap B \\ \vdash \ell : !A \multimap B \end{array} \mathbf{d} \\ \text{A linear proof} \\ \text{is in particular non-linear.} \end{array}$$

$$\frac{\vdash f : !A \multimap B}{\vdash D_0 f : A \multimap B} \frac{\bar{d}}{\bar{d}}$$
From a non-linear proof
we can extract a linear proof





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

Exponential rules of Differential Linear Logic

$$\frac{\vdash \Gamma}{\vdash \Gamma, cst_1 : ?A} w \qquad \frac{\vdash \Gamma, f : ?A, g : ?A}{\vdash \Gamma, f.g : ?A} c \qquad \frac{\vdash \Gamma, \ell : A}{\vdash \Gamma, \ell : ?A} d$$

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

$$\frac{?\Gamma \vdash x : A}{?\Gamma \vdash \delta_{\pi} : !A} p$$

Differentiation in Differential Linear Logic

The only thing you need to know:

$$\frac{ \begin{array}{c|c} \vdash \Gamma, v : A \\ \hline \vdash \Gamma, D_0(_)(v) : !A \end{array}}{ \begin{array}{c} \vdash \Gamma, v : A \\ \hline \vdash \Gamma, D_0(_)(v) : !A \end{array}} \frac{\bar{d}}{\bar{c}}$$

Dialectica factorizes through Differential Linear Logic

Witnesses are functorial reverse derivative

$$\mathbb{W}(A \Rightarrow B) = (\mathbb{W}(A) \Rightarrow \mathbb{W}(B)) \times (\mathbb{W}(A) \Rightarrow \mathbb{C}(B) \Rightarrow \mathbb{C}(A))$$

If $\Gamma \vdash A$ in LL, then $\mathbb{W}(\Gamma) \vdash \mathbb{W}(A)$ in classical DILL.

$$\frac{ \frac{}{\vdash A, A^{\perp}} \text{ ax}}{\vdash A, !A^{\perp}} \frac{\text{ax}}{\bar{d}} \quad \frac{}{\vdash ?A, !A^{\perp}} \text{ ax}}{\frac{\vdash ?A, A, !A^{\perp}}{\Gamma \vdash ?A, A}} \frac{\text{ax}}{\bar{c}} \quad \frac{\pi}{\Gamma \vdash ?A} \text{ cur}$$

Dialectica factorizes through Differential Linear Logic

 $[\![A \times B]\!]_e := A \& B$ $[\![A + B]\!]_e := A \oplus B$

IDILL: Intuitionnistic Differential Linear Logic? Oh no ...

$$A \leadsto \exists x : \mathbb{W}(A), \forall \underline{u} : \mathbb{C}(A), A_D[x, u]$$

Let's say x, u, f, g are λ -terms.

The computational Dialectica : a reverse Differential λ -calculus

"Behind every successful proof there is a program", Gödel's wife

A computational Dialectica

Making Dialectica act on λ -terms instead of formulas.

λ -terms with an extra type allowing for sums

$$\frac{\Gamma \vdash m_1 : \mathfrak{M} A \qquad \Gamma \vdash m_2 : \mathfrak{M} A}{\Gamma \vdash m_1 : \mathfrak{M} A \qquad \Gamma \vdash m_2 : \mathfrak{M} A}$$

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

$$\begin{array}{ccc} \mathbb{W}(A\Rightarrow B) &:= & (\mathbb{W}(A)\Rightarrow \mathbb{W}(B)) \\ & & \times (\mathbb{C}(B)\Rightarrow \mathbb{W}(A)\Rightarrow \mathfrak{M}\,\mathbb{C}(A)) \\ \mathbb{C}(A\Rightarrow B) &:= & \mathbb{W}(A)\times \mathbb{C}(B) \end{array}$$

Pédrot's Dialectica Transformation

Soundness [Ped14]

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

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

A global and a local transformation

```
\begin{array}{llll} x^{\bullet} & := & x & (\lambda x. \, t)^{\bullet} & := & (\lambda x. \, t^{\bullet}, \lambda \pi x. \, t_{x} \, \pi) \\ x_{x} & := & \lambda \pi. \, \{\pi\} & (\lambda x. \, t)_{y} & := & \lambda \pi. \, (\lambda x. \, t_{y}) \, \pi.1 \, \pi.2 \\ x_{y} & := & \lambda \pi. \, \varnothing \text{ if } x \neq y & (t \, u)^{\bullet} & := & (t^{\bullet}.1) \, u^{\bullet} \\ & & & & & & & & & & \\ (t \, u)_{y} := \lambda \pi. \, (t_{y} \, (u^{\bullet}, \pi)) \circledast \, ((t^{\bullet}.2) \, \pi \, u^{\bullet} \gg = u_{y}) \end{array}
```

Flashback: Differential λ -calculus [Ehrhard, Regnier 04]

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:

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

Operational Semantics:

$$\begin{array}{c} (\lambda x.s)T \to_{\beta} s[T/x] \\ \mathrm{D}(\lambda x.s) \cdot t \to_{\beta_D} \lambda x.\frac{\partial s}{\partial x} \cdot t \end{array}$$

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

Linearity in Linear Logic

Linearity is about resources: A proof/program is *linear* iff it uses only once its hypotheses/argument.

$$\begin{array}{ll} \textbf{Linear} & \textbf{Non-linear} \\ A \vdash A \lor B & A \vdash A \land A \\ \lambda f \lambda x. f xx & \lambda x. \lambda f. f xx \end{array}$$

Differentiation is about making a λ -term linear :

 \leadsto about making a $\lambda\text{-term}$ have a linear usage of its arguments.

$$\lambda x \lambda f. fxx \rightsquigarrow ?$$

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$$D(\lambda x \lambda f. fxx) \cdot v := \lambda x. \lambda f. vx + ?$$

Linearity in Linear Logic

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Differentiation is about making a λ -term linear :

 \leadsto about making a $\lambda\text{-term}$ have a linear usage of its arguments.

$$D(\lambda x \lambda f. fxx) \cdot v := \lambda x. \lambda f. vx + \lambda x. \lambda f. Dxv$$

The linear substitution ...

... which is not exactly a substitution

$$\frac{\partial y}{\partial x} \cdot t = \left\{ \begin{array}{l} t \ if \ x = y \\ 0 \ otherwise \end{array} \right. \qquad \frac{\partial}{\partial x} (tu) \cdot s = \left(\frac{\partial t}{\partial x} \cdot s \right) u + \left(\mathrm{D}t \cdot \left(\frac{\partial u}{\partial x} \cdot s \right) \right) u$$

$$\frac{\partial}{\partial x} (\lambda y.s) \cdot t = \lambda y. \frac{\partial s}{\partial x} \cdot t \qquad \frac{\partial}{\partial x} (\mathrm{D}s \cdot u) \cdot t = \mathrm{D}\left(\frac{\partial s}{\partial x} \cdot t \right) \cdot u + \mathrm{D}s \cdot \left(\frac{\partial u}{\partial x} \cdot t \right)$$

$$\frac{\partial}{\partial x} \cdot t = 0 \qquad \qquad \frac{\partial}{\partial x} (s + u) \cdot t = \frac{\partial s}{\partial x} \cdot t + \frac{\partial u}{\partial x} \cdot t$$

 $\frac{\partial s}{\partial x} \cdot t$ represents s where x is linearly (i.e. one time) substituted by t.

The linear substitution ...

The computational Dialectica

$$\frac{\partial y}{\partial x} \cdot t = \left\{ \begin{array}{l} t \ if \ x = y \\ 0 \ otherwise \end{array} \right. \qquad \frac{\partial}{\partial x} (tu) \cdot s = \left(\frac{\partial t}{\partial x} \cdot s \right) u + (\mathrm{D}t \cdot (\frac{\partial u}{\partial x} \cdot s)) u$$

$$x_y \cdot \pi = \left\{ \begin{array}{l} \pi \ if \ x = y \\ \emptyset \ otherwise \end{array} \right. \qquad (t \ u)_y := \lambda \pi \cdot (t_y \ (u^{\bullet}, \pi)) \circledast ((t^{\bullet}.2) \pi \ u^{\bullet} \gg u_y)$$

$$\frac{\partial}{\partial x} (\lambda y.s) \cdot t = \lambda y. \frac{\partial s}{\partial x} \cdot t \qquad \frac{\partial}{\partial x} (\mathrm{D}s \cdot u) \cdot t = \mathrm{D}(\frac{\partial s}{\partial x} \cdot t) \cdot u + \mathrm{D}s \cdot (\frac{\partial u}{\partial x} \cdot t)$$

$$\frac{\partial}{\partial x} \cdot t = 0 \qquad \qquad \frac{\partial}{\partial x} (s + u) \cdot t = \frac{\partial s}{\partial x} \cdot t + \frac{\partial u}{\partial x} \cdot t$$

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x_{x} := \lambda \pi. \{\pi\} \qquad x^{\bullet} := x
x_{y} := \lambda \pi. \varnothing \quad \text{if } x \neq y \qquad (\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 \qquad (t \ u)^{\bullet} := (t^{\bullet}. 1) \ u^{\bullet}
(t \ u)_{y} := \lambda \pi. (t_{y} (u^{\bullet}, \pi)) \circledast ((t^{\bullet}. 2) u^{\bullet} \pi \gg u_{y})
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(t \ u)_{y} := \lambda \pi. (t_{y} (u^{\bullet}, \pi)) \circledast ((t^{\bullet}. 2) u^{\bullet} \pi \gg u_{y})
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$$x_{x} := \lambda \pi \cdot \frac{\partial x}{\partial x} \cdot \pi \qquad x^{\bullet} := x$$

$$x_{y} := \lambda \pi \cdot \frac{\partial x}{\partial y} \cdot \pi \quad \text{if } x \neq y \qquad (\lambda x. t)^{\bullet} := (\lambda x. t^{\bullet}, \lambda x \pi. t_{x} \pi)$$

$$(\lambda x. t)_{y} := \lambda \pi. (\lambda x. t_{y}) \pi. 1 \pi. 2 \qquad (t \ u)^{\bullet} := \equiv (\lambda x. (tx)^{\bullet}) u^{\bullet}$$

$$(t \ u)_{y} := \lambda \pi. (t_{y} (u^{\bullet}, \pi)) \circledast ((t^{\bullet}. 2) u^{\bullet} \pi \gg u_{y})$$

That's reverse differentiation

- ▶ (_)•.2 obeys the chain rule, (_)• is the functorial differentiation.
- \triangleright t_x is contravariant in x, representing a reverse linear substitution.

Theorem [K. Pédrot 22]

$$\llbracket u \gg t_x [\Gamma \leftarrow \overrightarrow{r}] \rrbracket \equiv_{\beta,\eta} \lambda z. (\llbracket u \rrbracket ((\partial x.t [\Gamma \leftarrow \overrightarrow{r}])z))$$

$$x_{x} := \lambda \pi \cdot \frac{\partial x}{\partial x} \cdot \pi \qquad x^{\bullet} := x$$

$$x_{y} := \lambda \pi \cdot \frac{\partial x}{\partial y} \cdot \pi \quad \text{if } x \neq y \qquad (\lambda x. t)^{\bullet} := (\lambda x. t^{\bullet}, \lambda x \pi. t_{x} \pi)$$

$$(\lambda x. t)_{y} := \lambda \pi. (\lambda x. t_{y}) \pi. 1 \pi. 2 \qquad (t u)^{\bullet} \equiv (\lambda x. (tx)^{\bullet}) u^{\bullet}$$

That's reverse differentiation

- ▶ (_)•.2 obeys the chain rule, (_)• is the functorial differentiation.
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Theorem [K. Pédrot 22]

$$\llbracket u \gg t_x [\Gamma \leftarrow \overrightarrow{r^{\bullet}}] \rrbracket \equiv_{\beta,\eta} \lambda z. (\llbracket u \rrbracket ((\partial x.t [\Gamma \leftarrow \overrightarrow{r'}])z))$$

Dialectica is differentiation in categories

That's already known through lenses!

What's categorical differentiation?

To cook a good differential category, one needs:

► A category of regular/continuous/non-linear functions

$$\mathbb{C}(A,B) = !A \multimap B .$$

▶ A category of linear functions, in which differentiation embeds

$$\mathcal{L}(A,B) = A \multimap B.$$

► Something which linearizes :

$$\bar{d}:A\to !A$$

▶ A notion of <u>duality</u>, if one wants to encode <u>reverse</u>. differentiation.

→ Basically, one wants a categorical model of DiLL.

Dialectica categories

Categories representing specific relations

Consider a category C. **Dial**(C) is constructed as follows:

- ▶ Objects : relations $\alpha \subseteq U \times X$, $\beta \subseteq V \times Y$.
- ▶ Maps from α to β :

$$(f:U\to V, F:U\times Y\to X)$$

► Composition : the chain rule !

Consider

$$\begin{array}{ccc} (f,F): & \alpha \subseteq (A,X) & \to & \beta \subseteq (B,Y) \\ \text{and} & (g,G): & \beta \subseteq (B,Y) & \to & \gamma \subseteq (C,Z) \end{array}$$

two arrows of the Dialectica category. Then their composition is defined as

$$(g,G)\circ (f,F):=(g\circ f,(a,z)\mapsto F(a,G(f(a),z))).$$

Dialectica categories through Differential Categories In a *-autonomous differential category:

$$\partial: Id \otimes ! \to !$$

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

from $f: A \to B$ one constructs:

$$\overleftarrow{D}(f) \in \mathcal{L}(!A \otimes B^{\perp}, A^{\perp}).$$

Dialectica categories factorize through differential categories

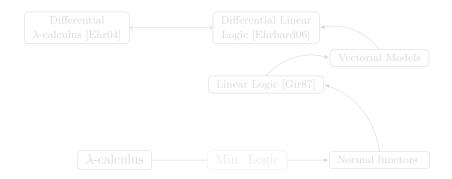
If \mathcal{L} is a model of D1LL such that $\mathcal{L}_!$ has finite limits:

$$\begin{cases} \mathcal{L}_{!} & \to & \mathscr{D}(\mathcal{L}_{!}) \\ A & \mapsto & A \times A^{\perp} \\ f & \mapsto & (f, \overleftarrow{D}(f)) \end{cases}$$

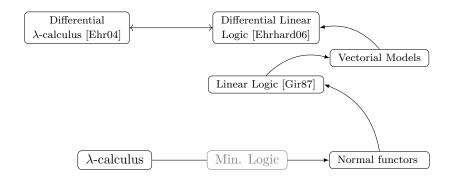
We have an obvious forgetful functor:

$$\mathcal{U}: \left\{ egin{array}{ll} \mathscr{D}(\mathscr{L}_!) &
ightarrow \mathscr{L}_! \ lpha \subseteq A imes X &
ightarrow A \ (f,F) &
ightarrow f \end{array}
ight.$$

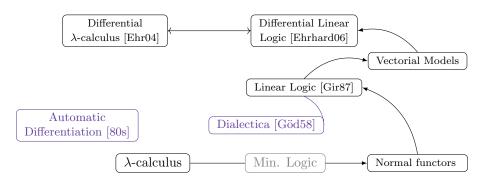
Programs	\mathbf{Logic}	Semantics
fun $(x:A) \rightarrow (t:B)$	Proof of $A \vdash B$	$f: A \to B$.
Types	Formulas	Objects
Execution	Cut-elimination	Equality



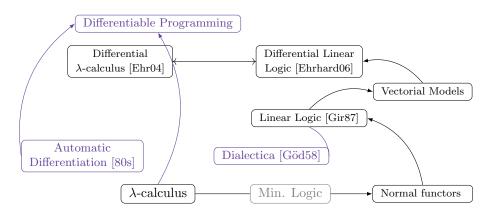
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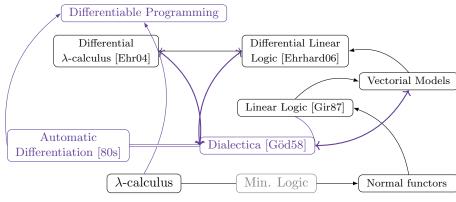
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Programs	\mathbf{Logic}	Semantics
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Types	Formulas	Objects
Execution	Cut-elimination	Equality



 $A\ good\ point\ for\ logicians:\ G\"{o}del\ invented\ Dialectica\ 40\ years\ before\ reverse\ differentiation\ was\ put\ to\ light$

Conclusion and applications

Take home message:

Dialectica is functorial reverse differentiation, extracting intensional local content from proofs.

A new semantical correspondance between computations and mathematics : intentional meaning of program is local behaviour of functions.

Program	Proof	Function
Quantitative	Resources	Linearity
${f Control}$	Classical Principles	Differentiation

Related work and potential applications:

- ▶ Markov's principle and delimited continuations on positive formulas.
- ▶ Proof mining and backpropagation.
- ▶ Bar Induction and Taylor Exponentiation.

Dialectica is differentiation ...

... We knew it already!

The codereliction of differential proof nets: In terms of polarity in linear logic [23], the \forall ----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 ----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:
$$(?P = (P \multimap \bot) \Rightarrow \bot) \rightarrow ((P \multimap \bot) \multimap \bot) \equiv P)$$

Hugo Herbelin, "An intuitionistic logic that proves Markov's principle", LICS '10 .

Differentiation and delimited continuations

Herbelin Lics'10

Markov's principle is proved by allowing catch and throw operations on hereditary positive formulas.

$$\frac{a: \neg \neg T \vdash_{\alpha:T} a: \neg \neg T}{a: \neg \neg T \vdash_{\alpha:T} a(\lambda b. \text{throw}_{\alpha} b): \bot} \xrightarrow{\text{THROW}} \xrightarrow{b: T \vdash_{\alpha:T} \text{throw}_{\alpha} b: \bot} \xrightarrow{\rightarrow_I} \xrightarrow{\rightarrow_E} \xrightarrow{a: \neg \neg T \vdash_{\alpha:T} a(\lambda b. \text{throw}_{\alpha} b): \bot} \xrightarrow{b: T \vdash_{\alpha:T} \lambda b. \text{throw}_{\alpha} b: \neg T} \xrightarrow{\rightarrow_E} \xrightarrow{a: \neg \neg T \vdash_{\alpha:T} \text{efq a}(\lambda b. \text{throw}_{\alpha} b): T} \xrightarrow{CATCH} \xrightarrow{a: \neg \neg T \vdash \text{catch}_{\alpha} \text{efq a}(\lambda b. \text{throw}_{\alpha} b): T} \xrightarrow{\rightarrow_I} \xrightarrow{\rightarrow_I}$$

Figure 3. Proof of MP

Proof Mining

Extracting quantitative information from proofs.

Effective moduli from ineffective uniqueness proofs. An unwinding of de La Vallée Poussin's proof for Chebycheff approximation*

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Abstract

We consider uniqueness theorems in classical analysis having the form

$$(+) \forall u \in U, v_1, v_2 \in V_u (G(u, v_1) = 0 = G(u, v_2) \rightarrow v_1 = v_2),$$

where U,V are complete separable metric spaces, V_u is compact in V and $G:U\times V\to \mathbb{R}$ is a constructive function.

If (+) is proved by arithmetical means from analytical assumptions

$$(++) \forall x \in X \exists y \in Y_x \forall z \in Z(F(x, y, z) = 0)$$

only (where X, Y, Z are complete separable metric spaces, $Y_x \subset Y$ is compact and

 $F: X \times Y \times Z \to \mathbb{R}$ constructive), then we can extract from the proof of $(++) \to (+)$ an effective modulus of uniqueness. i.e.

$$(+++) \forall u \in U, v_1, v_2 \in V_u, k \in \mathbb{N}(|G(u, v_1)|, |G(u, v_2)| \le 2^{-\Phi u k} \to d_V(v_1, v_2) \le 2^{-k}).$$

Proof Mining

Markov's principle and the independence of premises are necessary for most of **mathematical analysis proofs** :

Proof mining allows to refine these proofs by taking away thes principles as guaranteed by (some variant of) Dialectica's transformation.

Conjecture

Does it differentiate the function $(\epsilon \to \eta)$ in :

$$\forall u, v_1 v_2, \forall \epsilon > 0, \exists \eta > 0, \|G(u, v_1) - G(u, v_2)\| < \eta \to d_V(v_1, v_2) < \epsilon$$

?

Is proof mining (based on) reverse differentiation applied to proofs?

What else can we explain by differentiation?

Thank you for Listening!