

Higher order differentiation and Taylor expansion

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Linearity

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→ differential calculus (first order)

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→ differential calculus (first order)
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Needs an operational semantics for counting the uses of resources

Operational semantics

CBV

$$\bar{2} (I_1 I_2) 0 \rightsquigarrow \bar{2} I_2 0 \rightsquigarrow I_2 (I_2 0) \rightsquigarrow I_2 0 \rightsquigarrow 0$$

I_1 used once

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$$\bar{2}(I_1 I_2)0 \rightsquigarrow (I_1 I_2)((I_1 I_2)0) \rightsquigarrow I_2((I_1 I_2)0) \rightsquigarrow (I_1 I_2)0 \rightsquigarrow I_2 0 \rightsquigarrow 0$$

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We will stick to CBN (e.g., Krivine's machine)

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$$P[X_1 + X_2] \sim P[X_i]$$

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$$P\{X_1/x\} + P\{X_2/x\} \sim P\{X_i/x\} = P[X_i]$$

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P bilinear: has 2 (linear) occurrences of x

$$P\{(X_1 + X_2)/x\} \sim P[X_1 + X_2, X_1 + X_2] \sim P[X_i, X_j]$$
$$P\{X_1/x\} + P\{X_2/x\} \sim P\{X_i/x\} = P[X_i, X_j]$$

Sum as non deterministic choice

Remark

Can keep sums without committing choices, e.g. instead of

$$\begin{aligned} X_1 + X_2 &\rightsquigarrow X_i \\ P[X_1 + X_2] &\sim P[X_i] \end{aligned}$$

have

$$P[X_1 + X_2] \sim P[X_1] + P[X_2]$$

Linearity in lambda-calculus

Lambda-calculus application is not linear:

$$(P_1 + P_2)Q = P_1Q + P_2Q$$

but

$$P(Q_1 + Q_2) \neq PQ_1 + PQ_2$$

(e.g. composition $f \circ g$ in an additive category is only left linear)

Differentiation

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Example:

if $P = x^2 = x \cdot x$ then

$$\frac{\partial P}{\partial x} \cdot Q = Q \cdot x + x \cdot Q \quad (= 2x \cdot Q)$$

Differential lambda-calculus

Syntax

$$T = x \mid \lambda x T \mid T_1 T_2 \mid 0 \mid T_1 + T_2 \mid DT_1 . T_2$$

Associativity, commutativity

Linearity

$$\lambda x (P_1 + P_2) = \lambda x P_1 + \lambda x P_2$$

$$(P_1 + P_2)Q = P_1Q + P_2Q$$

$$D(P_1 + P_2) . Q = DP_1 . Q + DP_2 . Q$$

$$DP . (Q_1 + Q_2) = DP . Q_1 + DP . Q_2$$

Notation

- ▶ Differential calculus: if $f : \mathbb{R}^n \mapsto \mathbb{R}$ then $Df(x) : \mathbb{R}^n \mapsto \mathbb{R}$ linear

$$Df : \mathbb{R}^n \rightarrow (\mathbb{R}^n \multimap \mathbb{R})$$

$$Df(x).u = \text{diff of } f \text{ at } x \text{ in direction } u$$

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$$\rightsquigarrow DP . Q \text{ as a linearization of } PQ$$

Partial derivative = linear substitution

$$\frac{\partial x}{\partial x} \cdot Q = Q \qquad \frac{\partial y}{\partial x} \cdot Q = 0$$

$$\frac{\partial \lambda_y P}{\partial x} \cdot Q = \lambda_y \frac{\partial P}{\partial x} \cdot Q \quad (x \neq y)$$

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The chain rule

If $F : \mathbb{R}^p \mapsto \mathbb{R}$ and $U : \mathbb{R}^n \mapsto \mathbb{R}^p$ are differentiable then (with standard notations):

$$\frac{\partial F \circ U}{\partial x_i}(\vec{x}) = DF(U(\vec{x})) \cdot \frac{\partial U}{\partial x_i}(\vec{x})$$

If P_1 doesn't depend on x so that $\frac{\partial P_1}{\partial x} \cdot Q = 0$ then (with l-diff notations):

$$\frac{\partial P_1 P_2}{\partial x} \cdot Q = \left(DP_1 \cdot \left(\frac{\partial P_2}{\partial x} \cdot Q \right) \right) P_2$$

Dynamics

Reduction

$$(\lambda x P)Q \rightsquigarrow P\{Q/x\}$$
$$D\lambda x P . Q \rightsquigarrow \lambda x \frac{\partial P}{\partial x} \cdot Q$$

Taylor expansion

Differential calculus

$$f(x) = \sum_{n \geq 0} \frac{1}{n!} D^n f(0) \cdot x^n$$

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Differential *lambda*-calculus

$$PQ = \sum_{n \geq 0} \frac{1}{n!} (D^n P \cdot Q^n) 0$$

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$$P\{Q/x\} = \sum_{n \geq 0} \frac{1}{n!} \left(\frac{\partial^n P}{\partial x^n} \cdot Q^n \right) \{0/x\}$$

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And now let's move to the *full Taylor expansion*

Resource calculus

Syntax

$$s = x \mid \lambda x s \mid \langle s \rangle B$$

$$B = 1 \mid s_1 \dots s_n$$

+ linearity conditions

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Remark

$$\langle s \rangle B = (D^n s . B)0 \quad \text{where } \deg(B) = n$$

Linear substitution

$$\frac{\partial x}{\partial x} \cdot t = t \qquad \frac{\partial y}{\partial x} \cdot t = 0$$

$$\frac{\partial \lambda y s}{\partial x} \cdot t = \lambda y \frac{\partial s}{\partial x} \cdot t$$

$$\frac{\partial \langle s \rangle B}{\partial x} \cdot t = \left\langle \frac{\partial s}{\partial x} \cdot t \right\rangle B + \langle s \rangle \frac{\partial B}{\partial x} \cdot t$$

$$\frac{\partial s_1 \dots s_n}{\partial x} \cdot t = \sum_{i=1}^n s_1 \dots \left(\frac{\partial s_i}{\partial x} \cdot t \right) \dots s_n$$

Reduction

$$\langle \lambda x s \rangle s_1 \dots s_n \rightsquigarrow 0 \quad \text{if } \deg_x(s) \neq n$$

$$\rightsquigarrow \frac{\partial^n s}{\partial x^n} \cdot s_1 \dots s_n = \sum_{\sigma \in \mathfrak{S}_n} s[s_{\sigma 1} \dots s_{\sigma n}]$$

Uniformity

Uniform resource term = lambda-term approximant

Approximations

$$\lambda x x x \rightsquigarrow \lambda x \langle x \rangle x^n$$

$$\lambda f f (\lambda g g \lambda z z) \rightsquigarrow \lambda f \langle f \rangle 1$$

$$\rightsquigarrow \lambda f \langle f \rangle \lambda g \langle g \rangle 1$$

$$\rightsquigarrow \lambda f \langle f \rangle \lambda g \langle g \rangle 1 . \lambda g \langle g \rangle \lambda z z$$

$$\rightsquigarrow \lambda f \langle f \rangle \lambda g \langle g \rangle 1 . \lambda g \langle g \rangle \lambda z z . \lambda g \langle g \rangle (\lambda z z)^2$$

Reduction vs beta-reduction

Theorem

If $S \rightarrow_{\beta} S_0$ (normal form) and $s_0 \in \mathcal{T}(S_0)$ then there is a unique $s \in \mathcal{T}(S)$ s.t. $s \rightsquigarrow s_0$.

$$\begin{array}{ccc} S & \xrightarrow{*} & s \in \mathcal{T}(S) \\ \beta \downarrow & & \downarrow \\ S_0 & \xrightarrow{*} & s_0 \in \mathcal{T}(S_0) \end{array}$$

Full Taylor expansion

Let S be a lambda-term;

$$S^* = \sum_{s \in \mathcal{T}(S)} \frac{1}{m(s)} s$$

$m(s)$ is the *multiplicity coefficient* of s :

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Theorem

$$(PQ)^* = \sum_{n \geq 0} \frac{1}{n!} \langle P^* \rangle Q^{*n}$$