

Automatic Differentiation With Higher Infinitesimals

or: Computational Smooth
Infinitesimal Analysis in Weil Algebra

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Slides available at: <https://bit.ly/casc-smooth>

Today's Topic in short:

To Implement computational
Higher Infinitesimal Analysis,
utilising the connection between
Automatic Differentiation
and
Nilpotent Infinitesimal Analysis,
applying Gröbner basis technique

So what?

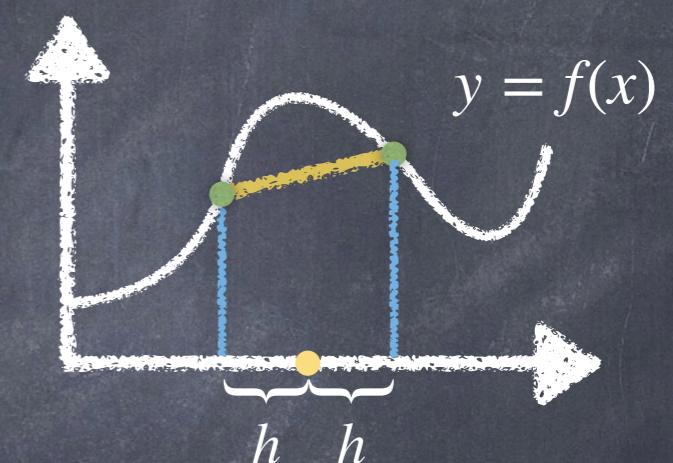
- **Automatic Differentiation** is a collection of methods to efficiently compute higher derivatives of the given function exploiting Chain Rule.
- **Nilpotent Infinitesimal Analysis:** an analysis using nilpotent infinitesimal such that $d^2 = 0$, $d \neq 0$. (Analysis in the era of Newton or Gauß)
 - Although such an infinitesimal real contradicts classically, it can be justified with topos theory and used to develop differential geometry (cf: **Synthetic Differential Geometry** by Lawvere et al.).
 - Such infinitesimals can be algebraically formulated as **Weil algebras**.
- Our Result: Combining **Gröbner basis** and Automatic Differentiation, we propose an algorithm to achieve infinitesimal analysis with general, higher infinitesimals associated with Weil algebras.
- We first explore the exact connection b/w AD and Nilpotent Analysis.

Automatic Differentiation (Forward-mode)

Three ways of differentiation

- **Automatic Differentiation** calculates differential coefficients efficiently and precisely, using Chain Rule.
- **Numerical Differentiation** replaces limits with small reals (e.g. Euler method)
 - Efficient, but only an approximation.
- **Symbolic Differentiation** expresses functions as symbolic ASTs and then differentiate them symbolically.
 - Exact, but easy to explode exponentially.

$$(f \circ g)'(x) = (f' \circ g)(x) \cdot g'(x)$$



$$f'(x) \approx \frac{f(x+h) - f(x-h)}{2h}$$

Automatic Differentiation (Forward mode)

A coefficient ring \mathbb{R}

value

diff. coeff.

```
data AD a = AD { real :: a, infinitesimal :: a }
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$$\text{AD } f \ f' + \text{AD } g \ g' = \text{AD } (f + g) \ (f' + g')$$

Linearity

$$\text{AD } f \ f' * \text{AD } g \ g' = \text{AD } (f * g) \ (f' * g + f * g')$$

Leibniz Rule

$$\sin(\text{AD } f \ f') = \text{AD } (\sin f) \ (f' * \cos f)$$

Chain Rule

- Basic Idea: Storing the value and differential coefficient simultaneously.
 - We use a shorthand: $\text{AD } x \ y = x + yd$.
- For $F: \mathbb{R} \rightarrow \mathbb{R}$, its AD version $\hat{F}: \text{AD}_{\mathbb{R}} \rightarrow \text{AD}_{\mathbb{R}}$ returns $F(f(x)) + \frac{d}{dx} F(f(x))d$, regarding inputs as of form $f(x) + f'(x)d$ for some function f .
- The differential of F at $x = a$ will be the coefficient of d in $\hat{F}(a + d)$.
 - $a + d$ corresp. to the differential at $x = a$ of identity function $f(x) = x$.

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Why does this work?

We can justify with this
with the concepts of
 C^∞ -ring and Weil algebra!

Justification

- Forward Mode AD corresponds to the **dual number ring**
 $\mathbb{R}[d] = \mathbb{R}[X]/(X^2)$, which has the canonical C^∞ -ring structure
 - The law $d^2 = 0$ characterises the AD!

Def. (Lawvere)

A C^∞ -ring is a commutative ring where the product-preserving map $A(f): A^n \rightarrow A$ is defined for any C^∞ -function $f: \mathbb{R}^n \rightarrow \mathbb{R}$.

"Smooth multivariate function naturally lifts to A "

⇒ More Rigorous Def

- Example: the ring $C^\infty(M)$ of real-valued smooth func. on a manifold M .
- Specifically, $\mathbb{R}[d]$ is an instance of **Weil algebras**, which is the special class of C^∞ -rings with a nilpotent infinitesimal structure!

Power serieses form a C^∞ -ring

Thm 2 (Lawvere)

The ring of multivariate formal power series $\mathbb{R}[[\mathbb{X}]]$ has the structure of C^∞ -ring via Taylor expansion.

Idea

For $f: \mathbb{R}^n \xrightarrow{C^\infty} \mathbb{R}$, $g_1, \dots, g_n \in \mathbb{R}[[\mathbb{X}]]$, we let:

$$\mathbb{R}[[\mathbb{X}]](f)(\vec{g}) := \sum_{\alpha \in \mathbb{N}^n} \frac{X^\alpha}{\alpha!} D^\alpha(f \circ \langle g_1, \dots, g_n \rangle)(\mathbf{0}).$$

Weil algebra

Def.

\mathbb{R} -vector space W is called a **Weil algebra** if there exists an ideal $I \subseteq \mathbb{R}[X_1, \dots, X_n]$ such that

$$W \cong \mathbb{R}[\mathbb{X}]/I, \text{ and } (X_1, \dots, X_n)^k \subseteq I \text{ for some } k.$$

- Intuition: a real line **augmented with nilpotent infinitesimals**.
- Each variable X_i corresponds to an infinitesimal.
 - X_i is nilpotent in the quotient ring $\mathbb{R}[\mathbb{X}]/I$.
 - Specifically, it is finite dimensional as a vector-space; hence **I is zero-dimensional ideal!**

Weil ring is a C^∞ -ring

Thm (Lawvere)

Any Weil algebra $W = \mathbb{R}[\mathbb{X}]/I$ has the canonical C^∞ -structure.

This follows from the following lemma:

Lem 1

For any ring-theoretic ideal I on a C^∞ -ring A , A/I has the canonical C^∞ -structure induced by the quotient map:

$$(A/I)(f)([x_1]_I, \dots, [x_m]_I) := [A(f)(x_1, \dots, x_m)]_I$$

- Since I is zero-dimensional, $\mathbb{R}[[X]]/I \cong \mathbb{R}[X]/I$ is a C^∞ -ring.
- In particular, **dual number ring** $\mathbb{R}[d] \cong \mathbb{R}[X]/(X^2)$ is also a C^∞ -ring!

The C^∞ -structure of $\mathbb{R}[d]$

- We let $f(a + bd) = f(a) + bf'(a)d$ for the univariate case.
- What about multivariate case $f: \mathbb{R}^n \xrightarrow{C^\infty} \mathbb{R}$?
 - This can be calculated from the C^∞ -structure of the univariate formal power series ring by cutting-off the terms at the degree 1 and replacing X with d .

The C^∞ -structure of $\mathbb{R}[d]$

Letting $g_i = a_i + b_i X$ ($i = 1, \dots, n$),

$$\begin{aligned} & \mathbb{R}[[X]](f)(\vec{g}) \\ &= f(\vec{g}(0)) + \frac{d}{dx}(f \circ \langle g_1, \dots, g_n \rangle)(0)X + \dots \\ &= f(a_1, \dots, a_n) + (g'_1(0) + \dots + g'_n(0)) \cdot f'(g_1(0), \dots, g_n(0)) X + \dots \\ &= f(a_1, \dots, a_n) + (b_1 + \dots + b_n) \cdot f'(a_1, \dots, a_n) X + \dots \end{aligned}$$

Hence, by quotienting by $I = (X^2)$, we have

$$\mathbb{R}[d](f)(a_1 + b_1 d, \dots, a_n + b_n d) = f(\vec{a}) + (b_1 + \dots + b_n) f'(\vec{a}) \cdot d.$$

This coincides with the definition of AD when $n = 1$!

Summary

- (Forward Mode) Automatic Differentiation is a method to **compute differentials of composite functions of smooth maps efficiently and precisely** by storing the direct value and differential coefficient at some point simultaneously.
- This can be viewed as utilising the structure of C^∞ -ring of the **dual number ring** $\mathbb{R}[d] = \mathbb{R}[X]/(X^2)$.
- The dual number ring is an instance of Weil algebra which axiomatises the real line with nilpotent infinitesimal.

Dual Number Ring and Smooth Infinitesimal Analysis

Why Weil algebras?

- Q. Why we have referred to Weil algebra and C^∞ -rings, instead of just saying "Forward-mode AD can be justified by Taylor expansion of smooth maps"?
- A. To generalise Forward Mode AD to multivariate / higher-order differentials and **higher-order nilpotent infinitesimals**.

Higher-order diff. with Weil algebras

- A tensor of the dual number rings can be used to compute higher derivatives. Let:

$$\mathbb{R}[d_1, \dots, d_n] = \mathbb{R}[X_1, \dots, X_n]/(X_i^2 \mid i \leq n) \cong \underbrace{\mathbb{R}[d] \otimes_{\mathbb{R}} \dots \otimes_{\mathbb{R}} \mathbb{R}[d]}_{n\text{-copies}}$$

- Then, by an easy induction, we have:

$$\mathbb{R}[\vec{d}](f)(x + d_1 + \dots + d_n) = \sum_{0 \leq i \leq n} f^{(i)}(x) \sigma_n^i(\vec{d}),$$

where σ_k^i is the k -variate elementary symmetric expr. of degree i .

- In particular, we can get **k -th derivative of f** by looking at the coefficient of $d_1 \dots d_k$ of $\mathbb{R}[\vec{d}](f)(x + d_1 + \dots + d_n)$!
- It has all the information of derivatives up to n th
- Likewise, higher derivatives of multivariate functions can be computed by a finite tensor products of the dual number rings.

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Our Contribution

Further Generalisation

- The structure of a Weil algebra $W = \mathbb{R}[\mathbb{X}]/I$ is determined by an ideal I with $(X_1, \dots, X_n) \subseteq \sqrt{I}$.
- W is a finite-dimensional vector space; i.e, I is zero-dimensional.
 - We can employ Gröbner basis algorithms to compute their \mathbb{R} -algebra structure.
- For C^∞ -structures of infinitesimal, we can first calculate in the formal power series ring, cutoff at the appropriate degree and take the quotient map.
 - In this way, we can do the calculation in the space with general nilpotent infinitesimals on computers!

Weil algebra Test

Algorithm 1 (WeilTest)

- Input: a polynomial ideal $I \subseteq \mathbb{R}[\mathbf{X}]$
- Is $I \subseteq \mathbb{R}[\mathbf{X}]$ a zerodimensional ideal? No \rightarrow Not a Weil algebra
- Computer the Gröbner basis of \sqrt{I} with Gröbner algorithms
- $X_1, \dots, X_n \in \sqrt{I}$
 - No \rightarrow Not a Weil algebra
 - Yes \rightarrow Returns monomial basis, an upper bound of multidegree, and the multiplication table ([6])
- With these information, algebraic operations on Weil algebra W is easy.
- In our paper, we also returns the expression of nontrivial monomials as the linear combination of basis for the sake of efficiency.

C^∞ -structure of Weil algebra

- Lem 1: the C^∞ -structure of a Weil algebra is given by the quotienting that of $\mathbb{R}[[X]]$.

- By Thm. 2, the C^∞ -structure of $\mathbb{R}[[X]]$ is given by:

$$\mathbb{R}[[X]](f)(g_1, \dots, g_m) = \sum_{\alpha \in \mathbb{N}^n} \frac{X^\alpha}{\alpha!} D^\alpha (f \circ \langle g_1, \dots, g_m \rangle)(\mathbf{0}).$$

- We need an arbitrary partial derivatives of composite functions.
 - In general, AD calculating higher-order derivatives of multivariate functions is called **Tower-mode AD**.
 - Tower-mode AD and $\mathbb{R}[[X]]$ coincides modulo the factor of $\alpha! = \alpha_1! \cdot \dots \cdot \alpha_m!$.
- Here, we think about C^∞ -maps, derivatives are commutative.
 - We proposed, in [9], a variant of Tower AD optimised for this case (Appendix)

Computing C^∞ -structure of Weil algebra

Algorithm 2 (LiftWeil)

- Input: $f: \mathbb{R}^n \xrightarrow{C^\infty} \mathbb{R}$ a smooth map admitting Tower AD,
 Information of Weil algebra $W = \mathbb{R}[X_1, \dots, X_n]/I$ given by Alg. 1 (Especially,
 monomial basis $\{X^{\beta_1}, \dots, X^{\beta_k}\}$ and multidegree upper bound α of W), and
 an element $\vec{w} = (w_1, \dots, w_n) \in W^n$.
- Output: $W(f)(\vec{w}) \in W$: given by C^∞ -structure of W .
 1. Take the representative polynomials \vec{g} of \vec{w} .
 2. Calculate $h \in \mathbb{R}[X]$ given by cutting off $\mathbb{R}[[X]](f)(\vec{g})$ at the upper bound α (Using Tower AD and multiply by factorial).
 3. Calculates $\bar{h}^G = \sum_{i \leq k} c_i X^{\beta_i}$ and return (c_1, \dots, c_k) .

Application 1: More efficient higher derivatives

- "Tensor of dual numbers" approach needs n -many variables to calculate n th derivative.
 - The result much duplicate and redundant informations.
 - In particular, # of terms explodes to 2^n to calculate up to n th!
- We can save space and time using the following lemma:

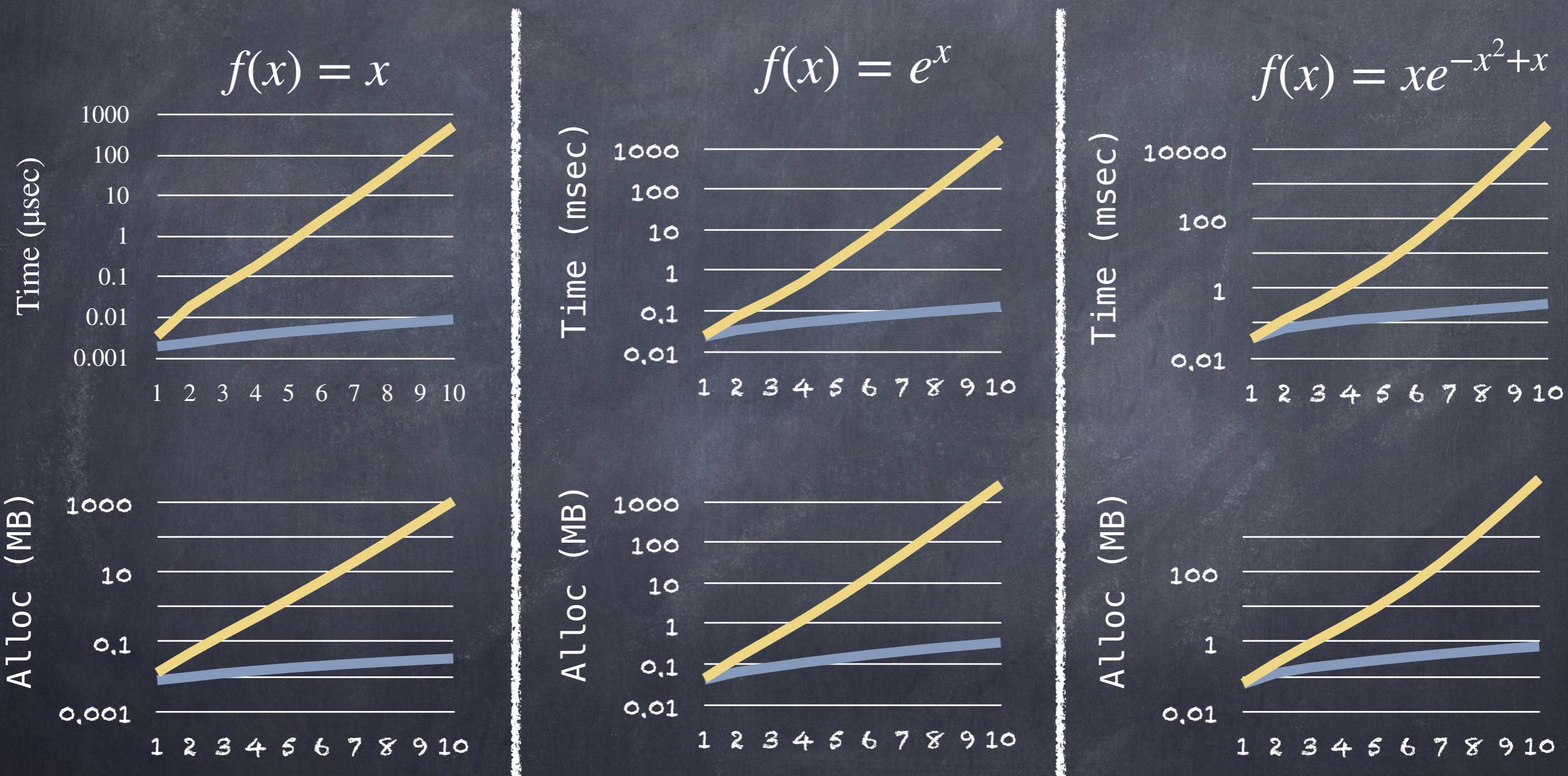
Lem

For $\varepsilon = (X) \in \mathbb{R}[X]/X^n$ and smooth $f: \mathbb{R} \rightarrow \mathbb{R}$, we have:

$$\mathbb{R}[\varepsilon](f)(x + \varepsilon) = \sum_{0 \leq i < n} \frac{1}{i!} f^{(i)}(x) \varepsilon^i$$

Higher derivatives: bench

— $\mathbb{R}[d] \otimes \dots \otimes \mathbb{R}[d]$ — $\mathbb{R}[X] / X^{n+1}$



Environment: virtual Linux environment on GitHub Actions (Standard_DS2_v2 Azure instance) with two Intel Xeon Platinum 8171M virtual CPUs (2.60GHz) and 7 GiB of RAM.

Application 2: Tensors

- The tensor $W_1 \otimes_{\mathbb{R}} W_2$ of Weil algebras W_1, W_2 is again a Weil (Lem 4):

$$\mathbb{R}[[X]]/I \otimes_{\mathbb{R}} \mathbb{R}[[Y]]/J \cong \mathbb{R}[[X, Y]]/\langle I, J \rangle$$

- The structure of tensors can be calculated with convolution (Alg. 4).
 - Using tensors we can combine arbitrary, multiple Weil algebras.
 - In particular, we can "package" higher-order ADs as Weil algebras, and "compose" them with tensors!

Ex.

Letting $\mathbb{R}[\varepsilon] = \mathbb{R}[x]/(x^{n+1})$, $\mathbb{R}[\delta] = \mathbb{R}[y]/(y^{m+1})$,

$$(\mathbb{R}[\varepsilon] \otimes_{\mathbb{R}} \mathbb{R}[\delta])(f)(x + \varepsilon, y + \delta) = \sum_{\substack{0 \leq i \leq n \\ 0 \leq j \leq m}} \frac{1}{i!j!} \partial x^i \partial y^j f(x, y).$$

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Summary

- ⦿ Forward Mode Automatic Differentiation corresponds to the nilpotent infinitesimal analysis of second-order.
 - ⦿ Justified by the theory of C^∞ -rings by Lawvere et al.
 - ⦿ In particular, Forward AD is an instance of Weil algebra which has nilpotent infinitesimals.
 - ⦿ These concepts came from Synthetic Differential Geometry.
- ⦿ Our Result: Combining AD and algorithms for zero-dimensional ideals, we achieved computational infinitesimal analysis with higher-order nilpotent infinitesimals on computers.
- ⦿ Future Works: Optimisation, Synthetic treatment of Differential Geometry on Computers

Existing Works

- ⦿ Nishimura and Osoekawa (2007) applies zero-dimensional algorithms to calculate **the fundamental relation of generators of limit of Weil algebras**.
 - ⦿ They focus on the utility application to develop the theory of Weil algebra in SDG and not interested in computing differentials of smooth functions.
- ⦿ Our interest is other direction – our central concern is computational infinitesimal analysis and smooth-ring computation.
 - ⦿ Our contribution clarifies the connection of ADs and SIAs, which has been already pointed out but not discussed thoroughly, and implement it on the computer.

Future Works

- Develop **Synthetic Differential Geometry** (by Lawvere et al.) on computers using higher infinitesimals.
 - In SDG, general Weil algebras such as $\mathbb{R}[d_1, \dots, d_n] = \mathbb{R}[x_1, \dots, x_n]/(d_i d_j \mid i, j \leq n)$ are used to develop theory.
 - This opens a door to synthetic treatment of differential-geometric objects on computer.
 - Challenge: although $C^\infty(M)$ is finitely-presented as a C^∞ -ring but, not so as an \mathbb{R} -algebra!
- Investigate the connection with other flavours of ADs, such as Reverse-Mode.

References

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Thank you!

Any Questions?

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- Future Works: Optimisation, Synthetic treatment of Differential Geometry on Computers

Appendices

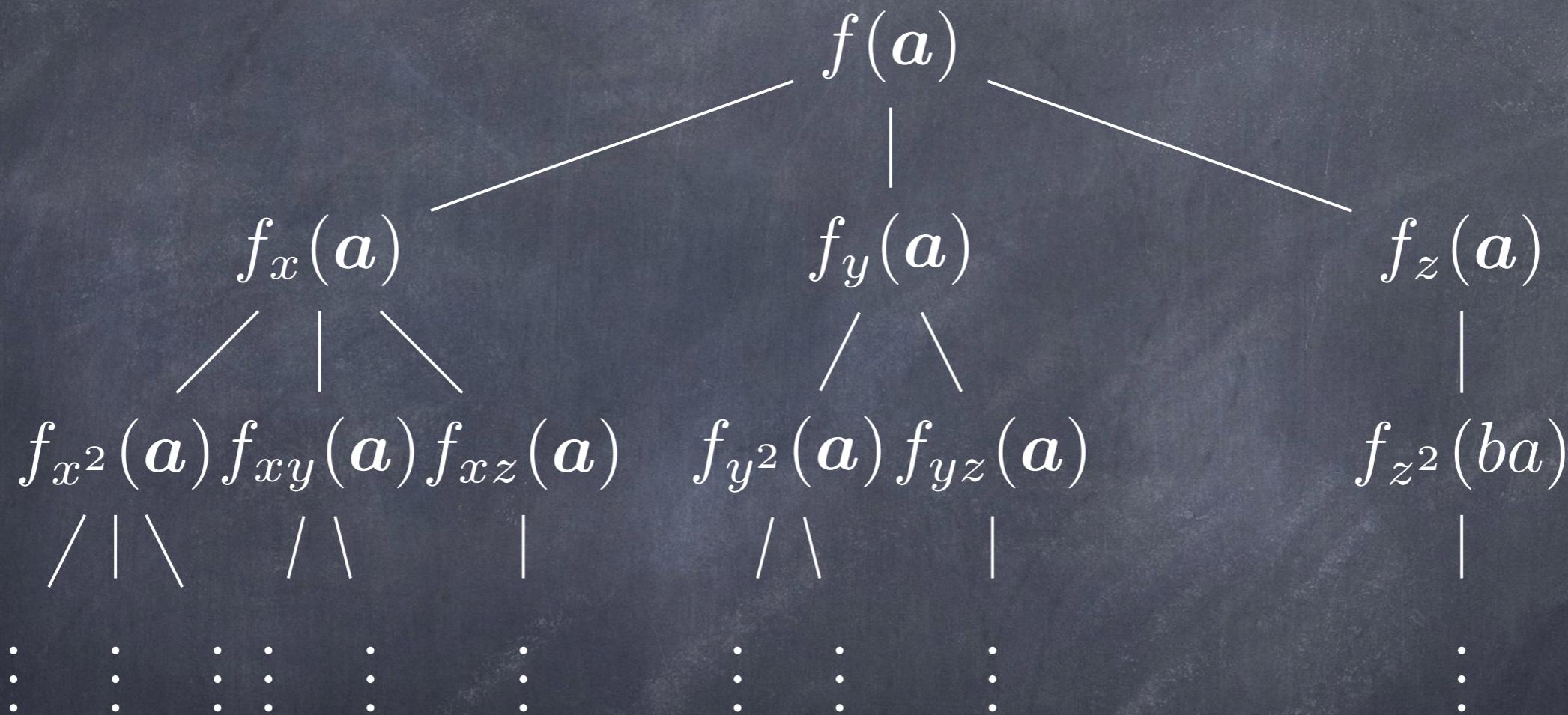
Categorically Rigorous Definition of smooth rings

Def. (Lawvere)

Let CartSp be the category of all finite-dimensional Euclidean spaces and C^∞ -maps between them.

A C^∞ -ring is a product-preserving functor $A : \text{CartSp} \rightarrow \text{Sets}$.

Succinct Tower AD



- Stores differentials as an infinite tree with decaying branching.
 - Goes right if no further differentiation will be taken for preceding variables.

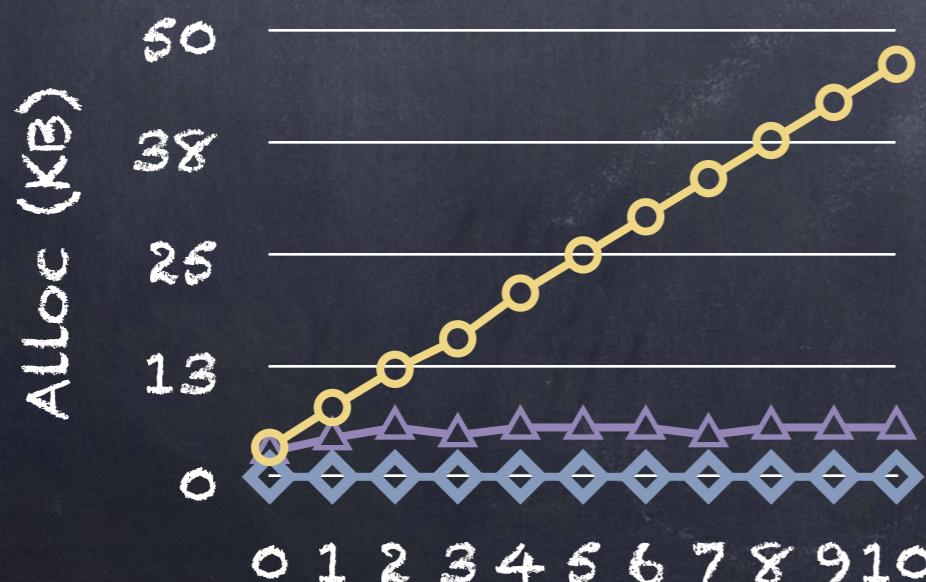
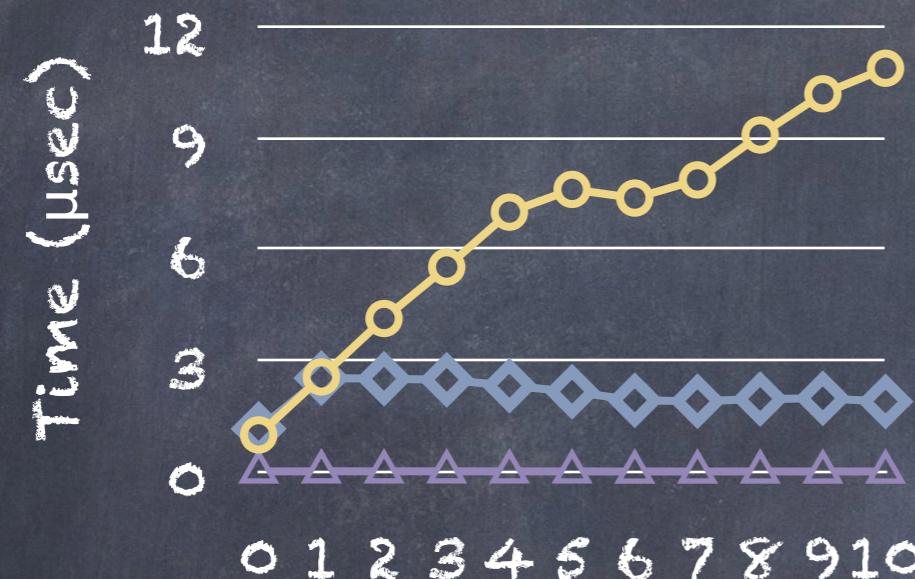
Tower bench: Univariate

○ Sparse

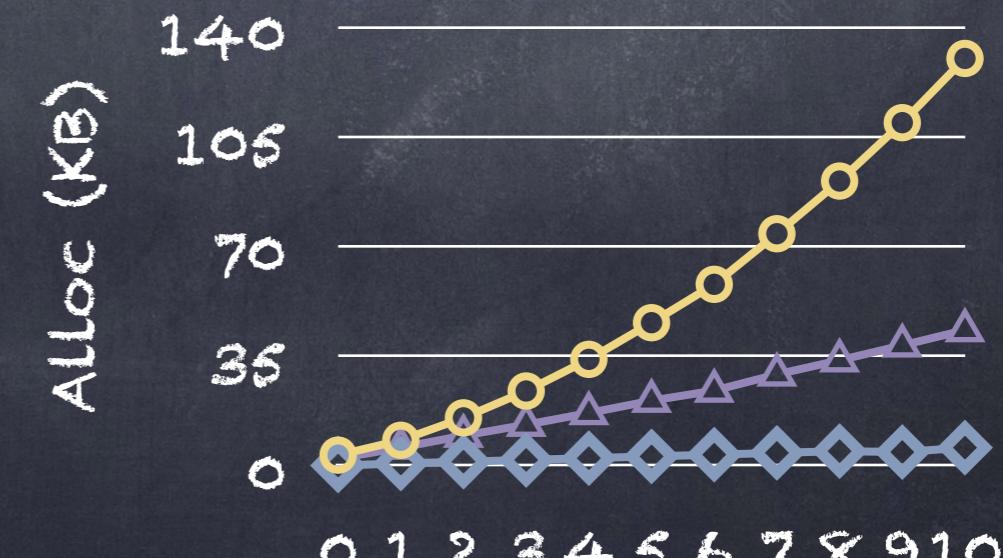
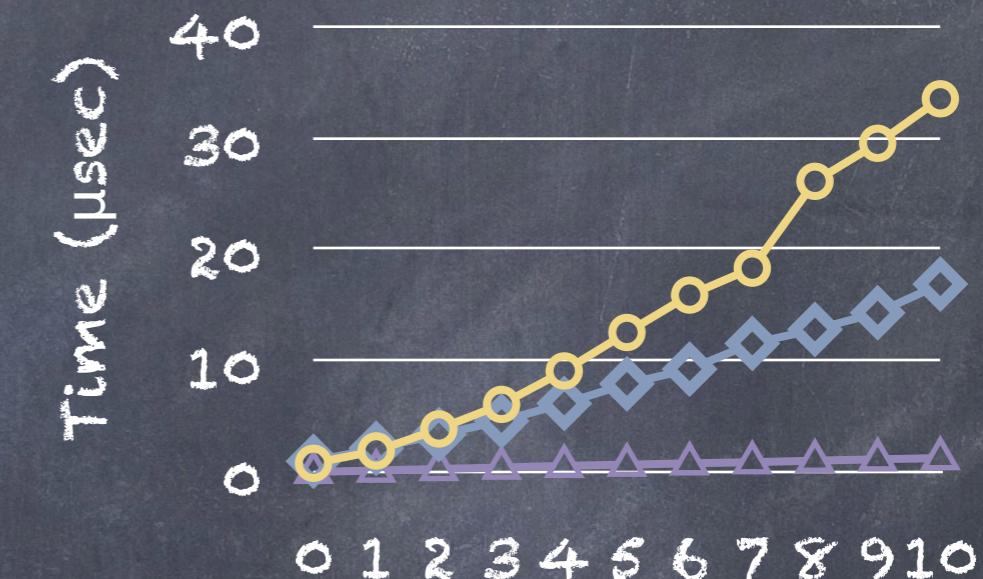
◇ diffs

△ Our Method

$$f(x) = x$$



$$f(x) = e^x$$



Tower bench: Multivariate

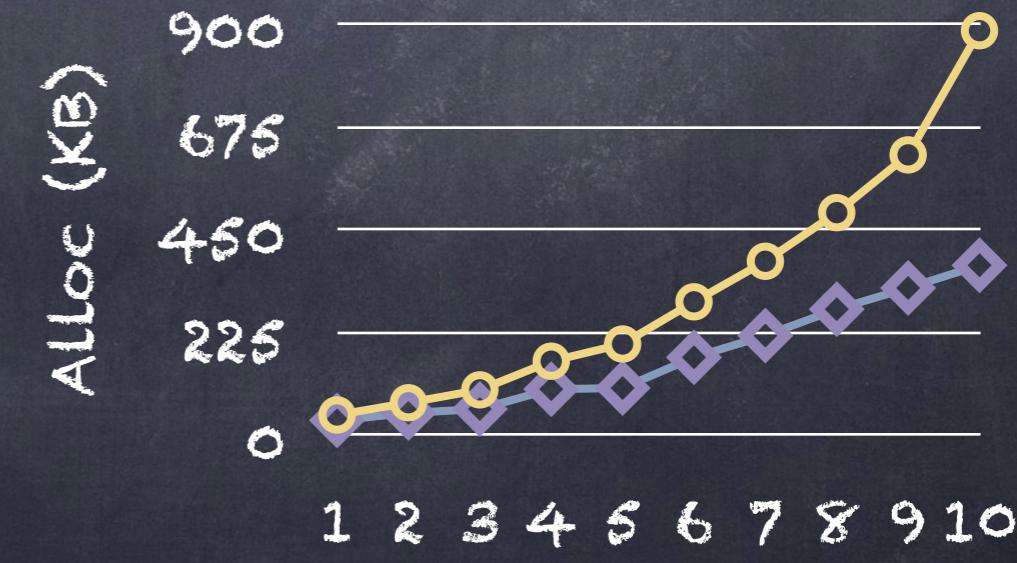
○ Sparse

$$f(x, y) = \sin x \cdot e^{y^2}$$



◊ Our Method

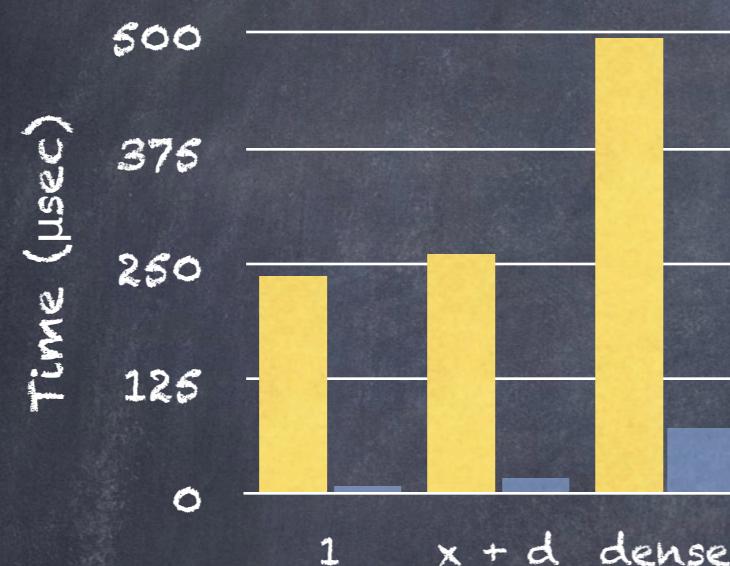
$$f(x, y, z) = \sin x \cdot e^{y^2+z}$$



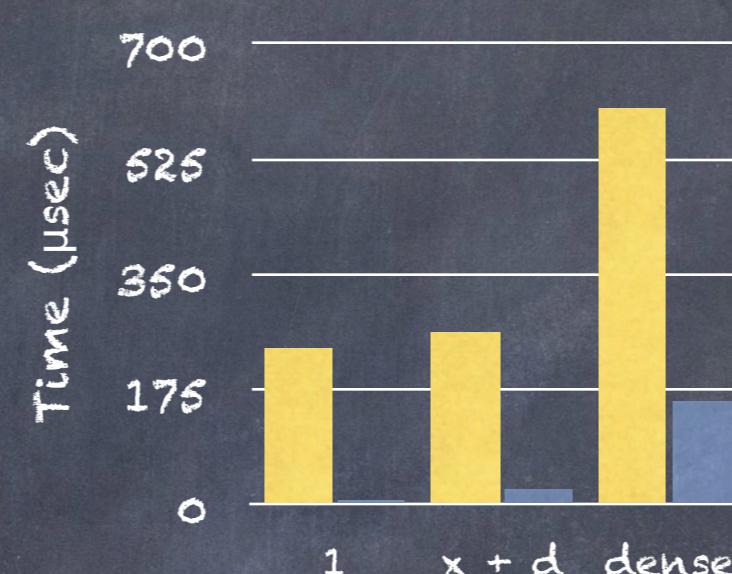
Tower bench: $\mathbb{R}[x, y]/(x^3 - y^2, y^3)$

Sparse Our Method

$$f(x) = x$$



$$f(x) = e^x$$



$$f(x, y, z) = \sin x \cdot e^{x^2+x}$$

