

Automatic Differentiation of Functional Programs or Lambda the Ultimate Calculus

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NEU
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Part of this talk covers joint work with Barak A. Pearlmutter.

The Essence

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need flow analysis

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need reflective transformation of closure bodies
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need **polyvariant** flow analysis

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$$\max_x \min_y f(x, y)$$

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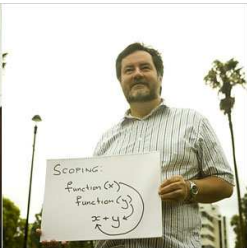
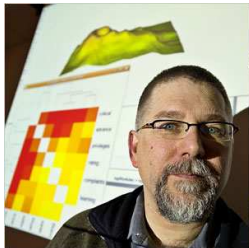
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The Essence of Forward-Mode AD

$$f(c + \varepsilon) = \frac{f(c)}{0!} + \frac{f'(c)}{1!}\varepsilon + \frac{f''(c)}{2!}\varepsilon^2 + \dots + \frac{f^{(i)}(c)}{i!}\varepsilon^i + \dots$$

Taylor, B. (1715). *Methodus Incrementorum Directa et Inversa*. London.

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To compute $\mathcal{D}f c$:

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$(\mathcal{D}f)$ is $\mathcal{O}(1)$ relative to f (both space and time).

$$a + bi$$

Hamilton, W. R. (1837). *Theory of conjugate functions, or algebraic couples; with a preliminary and elementary essay on algebra as the science of pure time*. Transactions of the Royal Irish Academy, **17**(1):293–422.

Arithmetic on Complex Numbers

$$a + bi$$

$$i^2 = -1$$

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$$(x_1 + x'_1\varepsilon) \times (x_2 + x'_2\varepsilon) = (x_1 \times x_2) + (x_1 \times x'_2 + x_2 \times x'_1)\varepsilon + (x'_1 + x'_2)\varepsilon^2$$

Clifford, W. K. (1873). *Preliminary Sketch of Bi-quaternions*. Proceedings of the London Mathematical Society, **4**:381–95.

Arithmetic on Dual Numbers

$$x + x'\varepsilon$$

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Arithmetic on Dual Numbers

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$$\langle x, x' \rangle$$

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Dynamic Overloading: SCMUTILS

```
(define-structure bundle primal tangent)
(define (primal p) (if (bundle? p) (bundle-primal p) p))
(define (tangent p) (if (bundle? p) (bundle-tangent p) 0))

(define +
  (let ((+ +))
    (lambda (x1 x2)
      (make-bundle (+ (primal x1) (primal x2))
                   (+ (tangent x1) (tangent x2))))))

(define *
  (let ((+ +) (* *))
    (lambda (x1 x2)
      (make-bundle (* (primal x1) (primal x2))
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(define ((D f) x) (tangent (f (make-bundle x 1))))
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(define (f x) (* 2 (* x (* x x))))
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Convenient

Dynamic Overloading: SCMUTILS

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Convenient but **slow**

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(define ((D f) x) (tangent (f (make-bundle x 1))))

(define (f x) (* 2 (* x (* x x))))

(D f)
(D (D f))
(D (lambda (x) ... (D (lambda (y) ...) ...) ...)) ...)
```

Convenient but **slow**

Dynamic Overloading: SCMUTILS

```
(define-structure bundle primal tangent)
(define (primal p) (if (bundle? p) (bundle-primal p) p))
(define (tangent p) (if (bundle? p) (bundle-tangent p) 0))

(define ((D f) x)
  (fluid-let ((+ (lambda (x1 x2)
                   (make-bundle (+ (primal x1) (primal x2))
                                   (+ (tangent x1) (tangent x2))))))
    (* (lambda (x1 x2)
         (make-bundle (* (primal x1) (primal x2))
                       (+ (* (primal x1) (tangent x2))
                           (* (tangent x1) (primal x2)))))))
      (tangent (f (make-bundle x 1))))))

(define (f x) (* 2 (* x (* x x))))

(D f)
(D (D f))
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```

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      (tangent (f (make-bundle x 1))))))

(define (f x) (* 2 (* x (* x x))))

(D f)
(D (D f))
(D (lambda (x) ... (D (lambda (y) ...) ...) ...)) ...)
```

Convenient but **slow**

Preprocessor: ADIFOR and TAPENADE

```
function f(x)
double precision x, f
f = 2.0d0*x*x*x
end
```

Preprocessor: ADIFOR and TAPENADE

```
function f(x)
double precision x, f
f = 2.0d0*x*x*x
end
```

```
function gf(x, gx, gresult)
double precision x, gx, gf, gresult
gf = 2.0d0*x*x*x
gresult = 6.0d0*x*x*gx
end
```

Preprocessor: ADIFOR and TAPENADE

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function f(x)
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gf = 2.0d0*x*x*x
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end
```

Fast

Preprocessor: ADIFOR and TAPENADE

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double precision x, f
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end
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```
function gf(x, gx, gresult)
double precision x, gx, gf, gresult
gf = 2.0d0*x*x*x
gresult = 6.0d0*x*x*gx
end
```

Fast but **inconvenient**

Preprocessor: ADIFOR and TAPENADE

```
function f(x)
double precision x, f
f = 2.0d0*x*x*x
end
```

AD_TOP = f

```
function gf(x, gx, gresult)
double precision x, gx, gf, gresult
gf = 2.0d0*x*x*x
gresult = 6.0d0*x*x*gx
end
```

Fast but **inconvenient**

Preprocessor: ADIFOR and TAPENADE

```
function f(x)
double precision x, f
f = 2.0d0*x*x*x
end
```

```
AD_TOP = f
AD_IVARS = x
AD_DVARS = f
```

```
function gf(x, gx, gresult)
double precision x, gx, gf, gresult
gf = 2.0d0*x*x*x
gresult = 6.0d0*x*x*gx
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AD_TOP = f
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```

```
AD_TOP = gf
AD_IVARS = x, gx
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Fast but **inconvenient**

Preprocessor: ADIFOR and TAPENADE

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function f(x)
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AD_TOP = f
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gf = 2.0d0*x*x*x
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```
AD_TOP = gf
AD_IVARS = x, gx
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```
function ggf(x, gx, gx, ggx, gresult, ggresult, gresult)
double precision x, gx, gx, ggx, ggf, gresult, gresult, ggresult
ggf = 2.0d0*x*x*x
gresult = 6.0d0*x*x*gx
gresult = 6.0d0*x*x*gx
ggresult = 6.0d0*x*x*ggx+12.0d0*x*gx*gx
end
```

Fast but **inconvenient**

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function f(x)
double precision x, f
f = 2.0d0*x*x*x
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```
AD_TOP = f
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function gf(x, gx, gresult)
double precision x, gx, gf, gresult
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AD_TOP = gf
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function ggf(x, gx, gx, ggx, gresult, ggresult, gresult)
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function f(x)
double precision x, f
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AD_TOP = f
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AD_DVARS = f
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```
function gf(x, gx, gresult)
double precision x, gx, gf, gresult
gf = 2.0d0*x*x*x
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AD_TOP = gf
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function ggf(x, gx, gx, ggx, gresult, ggresult, gresult)
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ggf = 2.0d0*x*x*x
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```

Fast but **inconvenient**

Preprocessor: ADIFOR and TAPENADE

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function f(x)
double precision x, f
f = 2.0d0*x*x*x
end
```

```
AD_TOP = f
AD_IVARS = x
AD_DVARS = f
```

```
function gf(x, gx, gresult)
double precision x, gx, gf, gresult
gf = 2.0d0*x*x*x
gresult = 6.0d0*x*x*gx
end
```

```
AD_TOP = gf
AD_IVARS = x, gx
AD_DVARS = gf, gresult
AD_PREFIX = h
```

```
function hgf(x, hx, gx, hgx, gresult, hgresult, hresult)
double precision x, hx, gx, hgx, hgf, hresult, gresult, hgresult
hgf = 2.0d0*x*x*x
hresult = 6.0d0*x*x*hx
gresult = 6.0d0*x*x*gx
hgresult = 6.0d0*x*x*hgx+12.0d0*x*gx*hx
end
```

Fast but **inconvenient**

Static Overloading: FADBAD++

```
double f(double x) {return 2*x*x*x;}  
double x;  
... f(x) ...
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double f(double x) {return 2*x*x*x;}  
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```
F<double> f(F<double> x) {return 2*x*x*x;}  
F<double> x;  
x.diff(0, 1);  
... f(x).d(0) ...
```

Static Overloading: FADBAD++

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```

```
F<F<double> > f(F<F<double> > x) {return 2*x*x*x;}  
F<F<double> > x;  
x.diff(0, 1);  
x.diff(0, 1).diff(0,1);  
... f(x).d(0).d(0) ...
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Slow

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x.diff(0, 1).diff(0,1);  
... f(x).d(0).d(0) ...
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Slow

Static Overloading: FADBAD++

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Slow and **inconvenient**

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Slow and **inconvenient**

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template <typename T>
T f(T x) {return 2*x*x*x;}
T x;
```

Slow and **inconvenient**

Static Overloading: FADBAD++

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double f(double x) {return 2*x*x*x;}  
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Slow and **inconvenient**

Differential geometry bundles points \mathbb{R}^n in a manifold with tangent vectors $\overline{\mathbb{R}^h}$

Our API for Functional Forward AD

$$\text{bundle} : \mathbb{R}^n \times \overline{\mathbb{R}^h} \rightarrow (\mathbb{R}^n \triangleright \overline{\mathbb{R}^h})$$

Differential geometry bundles points \mathbb{R}^n in a manifold with tangent vectors $\overline{\mathbb{R}^h}$

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$$j^* : (\mathbb{R}^n \rightarrow \mathbb{R}^m) \rightarrow ((\mathbb{R}^n \triangleright \overline{\mathbb{R}^h}) \rightarrow (\mathbb{R}^m \triangleright \overline{\mathbb{R}^m}))$$

Differential geometry bundles points \mathbb{R}^n in a manifold with tangent vectors $\overline{\mathbb{R}^h}$
 j^* maps a **function** to its *push forward*

Our API for Functional Forward AD

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Differential geometry bundles points \mathbb{R}^n in a manifold with tangent vectors $\overline{\mathbb{R}^h}$
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$$\begin{aligned}\text{bundle} & : \tau \times \overline{\tau} \rightarrow (\tau \triangleright \overline{\tau}) \\ \text{primal} & : (\tau \triangleright \overline{\tau}) \rightarrow \tau \\ \text{tangent} & : (\tau \triangleright \overline{\tau}) \rightarrow \overline{\tau} \\ \text{j*} & : (\tau_1 \rightarrow \tau_2) \rightarrow ((\tau_1 \triangleright \overline{\tau}_1) \rightarrow (\tau_2 \triangleright \overline{\tau}_2))\end{aligned}$$

Differential geometry bundles points \mathbb{R}^n in a manifold with tangent vectors $\overline{\mathbb{R}^n}$
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Generalize to arbitrary types

Our API for Functional Forward AD

bundle : $\tau \times \overline{\tau} \rightarrow (\tau \triangleright \overline{\tau})$
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Differential geometry bundles points \mathbb{R}^n in a manifold with tangent vectors $\overline{\mathbb{R}^n}$
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Generalize to arbitrary types

What is the tangent of a discrete value or a function?

Our API for Functional Forward AD

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What is the tangent of a discrete value or a function?

Can abbreviate $\tau \triangleright \overline{\tau}$ as $\overline{\tau}^{\rightarrow}$

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(define ((D f) x) (tangent ((j* f) (bundle x 1))))
```

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Convenient

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What is $(j^* j^*)$?

Convenient and **fast**

A property

A property

$x : \mathbb{R}^n$

A property

$$x : \mathbb{R}^n$$

$$\overline{x} : \mathbb{R}^n$$

A property

$$x : \mathbb{R}^n$$

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

A property

$$x : \mathbb{R}^n$$

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$$((\mathcal{J} f) x)[i, j] = \frac{\partial f(x)[i]}{\partial x[j]}$$

A property

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$$\overline{x} : \mathbb{R}^n$$

$$((\mathcal{J} f) \overline{x}) : \mathbb{R}^{m \times n}$$

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A property

$$x : \mathbb{R}^n$$

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$$((\mathcal{J} f) \overline{x}) : \mathbb{R}^n \xrightarrow{L} \mathbb{R}^m$$

A property

$$\begin{aligned}x &: \mathbb{R}^n & \bar{x} &: \mathbb{R}^n & f &: \mathbb{R}^n \rightarrow \mathbb{R}^m \\((\mathcal{J} f) x)[i,j] &= \frac{\partial f(x)[i]}{\partial x[j]} & ((\mathcal{J} f) x) &: \mathbb{R}^{m \times n} & ((\mathcal{J} f) x) &: \mathbb{R}^n \xrightarrow{L} \mathbb{R}^m \\(\text{primal } ((j^* f) (\text{bundle } x \bar{x}))) &= (f x)\end{aligned}$$

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$$\begin{aligned}x &: \mathbb{R}^n & \overline{x} &: \mathbb{R}^n & f &: \mathbb{R}^n \rightarrow \mathbb{R}^m \\((\mathcal{J} f) x)[i,j] &= \frac{\partial f(x)[i]}{\partial x[j]} & ((\mathcal{J} f) x) &: \mathbb{R}^{m \times n} & ((\mathcal{J} f) x) &: \mathbb{R}^n \xrightarrow{L} \mathbb{R}^m \\(\text{primal } ((j^* f) (\text{bundle } x \overline{x}))) &= (f x) \\(\text{tangent } ((j^* f) (\text{bundle } x \overline{x}))) &= ((\mathcal{J} f) x) \times \overline{x}\end{aligned}$$

A property

$$\begin{aligned}x &: \mathbb{R}^n & \bar{x} &: \mathbb{R}^n & f &: \mathbb{R}^n \rightarrow \mathbb{R}^m \\((\mathcal{J}f) x)[i,j] &= \frac{\partial f(x)[i]}{\partial x[j]} & ((\mathcal{J}f) x) &: \mathbb{R}^{m \times n} & ((\mathcal{J}f) x) &: \mathbb{R}^n \xrightarrow{L} \mathbb{R}^m \\(\text{primal } ((j^* f) (\text{bundle } x \bar{x}))) &= (f x) \\(\text{tangent } ((j^* f) (\text{bundle } x \bar{x}))) &= ((\mathcal{J}f) x) \times \bar{x} \\(\text{tangent } ((j^* f) (\text{bundle } x \bar{x}))) &= (((\mathcal{J}f) x) \bar{x})\end{aligned}$$

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$$\begin{aligned}x &: \mathbb{R}^n & \bar{x} &: \mathbb{R}^n & f &: \mathbb{R}^n \rightarrow \mathbb{R}^m \\((\mathcal{J}f) x)[i,j] &= \frac{\partial f(x)[i]}{\partial x[j]} & ((\mathcal{J}f) x) &: \mathbb{R}^{m \times n} & ((\mathcal{J}f) x) &: \mathbb{R}^n \xrightarrow{L} \mathbb{R}^m \\(\text{primal } ((j^* f) (\text{bundle } x \bar{x}))) &= (f x) \\(\text{tangent } ((j^* f) (\text{bundle } x \bar{x}))) &= ((\mathcal{J}f) x) \times \bar{x} \\(\text{tangent } ((j^* f) (\text{bundle } x \bar{x}))) &= (((\mathcal{J}f) x) \bar{x}) \\((j^* f) x) &= (\text{bundle } (f (\text{primal } x)) (((\mathcal{J}f) (\text{primal } x)) (\text{tangent } x)))\end{aligned}$$

A property

$$\begin{aligned}x &: \mathbb{R}^n & \bar{x} &: \mathbb{R}^n & f &: \mathbb{R}^n \rightarrow \mathbb{R}^m \\((\mathcal{J}f) x)[i,j] &= \frac{\partial f(x)[i]}{\partial x[j]} & ((\mathcal{J}f) x) &: \mathbb{R}^{m \times n} & ((\mathcal{J}f) x) &: \mathbb{R}^n \xrightarrow{L} \mathbb{R}^m \\(\text{primal } ((j^* f) (\text{bundle } x \bar{x}))) &= (f x) \\(\text{tangent } ((j^* f) (\text{bundle } x \bar{x}))) &= ((\mathcal{J}f) x) \times \bar{x} \\(\text{tangent } ((j^* f) (\text{bundle } x \bar{x}))) &= (((\mathcal{J}f) x) \bar{x}) \\((j^* f) x) &= (\text{bundle } (f (\text{primal } x)) ((\mathcal{J}f) (\text{primal } x)) (\text{tangent } x))) \\ \text{rearrangement function: } &(\forall i)(\exists j)(f x)[i] = x[j]\end{aligned}$$

A property

$$\begin{aligned}x &: \mathbb{R}^n & \bar{x} &: \mathbb{R}^n & f &: \mathbb{R}^n \rightarrow \mathbb{R}^m \\((\mathcal{J}f) x)[i,j] &= \frac{\partial f(x)[i]}{\partial x[j]} & ((\mathcal{J}f) x) &: \mathbb{R}^{m \times n} & ((\mathcal{J}f) x) &: \mathbb{R}^n \xrightarrow{L} \mathbb{R}^m \\(\text{primal } ((j^* f) (\text{bundle } x \bar{x}))) &= (f x) \\(\text{tangent } ((j^* f) (\text{bundle } x \bar{x}))) &= ((\mathcal{J}f) x) \times \bar{x} \\(\text{tangent } ((j^* f) (\text{bundle } x \bar{x}))) &= (((\mathcal{J}f) x) \bar{x}) \\((j^* f) x) &= (\text{bundle } (f (\text{primal } x)) ((\mathcal{J}f) (\text{primal } x)) (\text{tangent } x))) \\ \text{rearrangement function: } &(\forall i)(\exists j)(f x)[i] = x[j] \\ f &: \mathbb{R}^n \xrightarrow{L} \mathbb{R}^m\end{aligned}$$

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$$f : \mathbb{R}^n \xrightarrow{L} \mathbb{R}^m$$

0/1 matrix, every row has exactly one 1

A property

$$\begin{aligned}x &: \mathbb{R}^n & \bar{x} &: \mathbb{R}^n & f &: \mathbb{R}^n \rightarrow \mathbb{R}^m \\((\mathcal{J}f)x)[i,j] &= \frac{\partial f(x)[i]}{\partial x[j]} & ((\mathcal{J}f)x) &: \mathbb{R}^{m \times n} & ((\mathcal{J}f)x) &: \mathbb{R}^n \xrightarrow{L} \mathbb{R}^m \\(\text{primal } ((j^* f) (\text{bundle } x \bar{x}))) &= (f x) \\(\text{tangent } ((j^* f) (\text{bundle } x \bar{x}))) &= ((\mathcal{J}f)x) \times \bar{x} \\(\text{tangent } ((j^* f) (\text{bundle } x \bar{x}))) &= (((\mathcal{J}f)x) \bar{x}) \\((j^* f)x) &= (\text{bundle } (f (\text{primal } x)) ((\mathcal{J}f) (\text{primal } x)) (\text{tangent } x))) \\ \text{rearrangement function: } &(\forall i)(\exists j)(f x)[i] = x[j]\end{aligned}$$

$$f : \mathbb{R}^n \xrightarrow{L} \mathbb{R}^m$$

0/1 matrix, every row has exactly one 1

$$((\mathcal{J}f)x)$$

A property

$$\begin{aligned}x &: \mathbb{R}^n & \bar{x} &: \mathbb{R}^n & f &: \mathbb{R}^n \rightarrow \mathbb{R}^m \\((\mathcal{J} f) x)[i, j] &= \frac{\partial f(x)[i]}{\partial x[j]} & ((\mathcal{J} f) x) &: \mathbb{R}^{m \times n} & ((\mathcal{J} f) x) &: \mathbb{R}^n \xrightarrow{L} \mathbb{R}^m \\(\text{primal } ((j^* f) (\text{bundle } x \bar{x}))) &= (f x) \\(\text{tangent } ((j^* f) (\text{bundle } x \bar{x}))) &= ((\mathcal{J} f) x) \times \bar{x} \\(\text{tangent } ((j^* f) (\text{bundle } x \bar{x}))) &= (((\mathcal{J} f) x) \bar{x}) \\((j^* f) x) &= (\text{bundle } (f (\text{primal } x)) ((\mathcal{J} f) (\text{primal } x)) (\text{tangent } x))) \\ \text{rearrangement function: } &(\forall i)(\exists j)(f x)[i] = x[j]\end{aligned}$$

$$f : \mathbb{R}^n \xrightarrow{L} \mathbb{R}^m$$

0/1 matrix, every row has exactly one 1

$$((\mathcal{J} f) x) = \frac{\partial f(x)[i]}{\partial x[j]}$$

A property

$$\begin{aligned}x &: \mathbb{R}^n & \bar{x} &: \mathbb{R}^n & f &: \mathbb{R}^n \rightarrow \mathbb{R}^m \\((\mathcal{J}f)x)[i,j] &= \frac{\partial f(x)[i]}{\partial x[j]} & ((\mathcal{J}f)x) &: \mathbb{R}^{m \times n} & ((\mathcal{J}f)x) &: \mathbb{R}^n \xrightarrow{L} \mathbb{R}^m \\(\text{primal } ((j^* f) (\text{bundle } x \bar{x}))) &= (f x) \\(\text{tangent } ((j^* f) (\text{bundle } x \bar{x}))) &= ((\mathcal{J}f)x) \times \bar{x} \\(\text{tangent } ((j^* f) (\text{bundle } x \bar{x}))) &= (((\mathcal{J}f)x) \bar{x}) \\((j^* f)x) &= (\text{bundle } (f (\text{primal } x)) ((\mathcal{J}f) (\text{primal } x)) (\text{tangent } x))) \\ \text{rearrangement function: } &(\forall i)(\exists j)(f x)[i] = x[j]\end{aligned}$$

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0/1 matrix, every row has exactly one 1

$$((\mathcal{J}f)x) = \frac{\partial f(x)[i]}{\partial x[j]} = \begin{cases} 1 & \text{when } (f x)[i] = x[j] \\ 0 & \text{otherwise} \end{cases}$$

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0/1 matrix, every row has exactly one 1

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when f is a rearrangement function

$$((j* f) x) = (\text{bundle } (f (\text{primal } x)) (f (\text{tangent } x)))$$

A property

$$x : \mathbb{R}^n \quad \bar{x} : \mathbb{R}^n \quad f : \mathbb{R}^n \rightarrow \mathbb{R}^m$$
$$((\mathcal{J}f) x)[i,j] = \frac{\partial f(x)[i]}{\partial x[j]} \quad ((\mathcal{J}f) x) : \mathbb{R}^{m \times n} \quad ((\mathcal{J}f) x) : \mathbb{R}^n \xrightarrow{L} \mathbb{R}^m$$

$$(\text{primal } ((j* f) (\text{bundle } x \bar{x}))) = (f x)$$

$$(\text{tangent } ((j* f) (\text{bundle } x \bar{x}))) = ((\mathcal{J}f) x) \times \bar{x}$$

$$(\text{tangent } ((j* f) (\text{bundle } x \bar{x}))) = (((\mathcal{J}f) x) \bar{x})$$

$$((j* f) x) = (\text{bundle } (f (\text{primal } x)) ((\mathcal{J}f) (\text{primal } x)) (\text{tangent } x))$$

rearrangement function: $(\forall i)(\exists j)(f x)[i] = x[j]$

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A property

$$\begin{aligned}x &: \tau_1 & \bar{x} &: \tau_1 & f &: \tau_1 \rightarrow \tau_2 \\((\mathcal{J}f) x)[i,j] &= \frac{\partial f(x)[i]}{\partial x[j]} & ((\mathcal{J}f) x) &: \tau_1 \xrightarrow{L} \tau_2 \\(\text{primal } ((j* f) (\text{bundle } x \bar{x}))) &= (f x) \\(\text{tangent } ((j* f) (\text{bundle } x \bar{x}))) &= ((\mathcal{J}f) x) \times \bar{x} \\(\text{tangent } ((j* f) (\text{bundle } x \bar{x}))) &= (((\mathcal{J}f) x) \bar{x}) \\((j* f) x) &= (\text{bundle } (f (\text{primal } x)) (((\mathcal{J}f) (\text{primal } x)) (\text{tangent } x))) \\ \text{rearrangement function: } &(\forall i)(\exists j)(f x)[i] = x[j]\end{aligned}$$

$$f: \tau_1 \xrightarrow{L} \tau_2$$

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when f is a rearrangement function

$$((j* f) x) = (\text{bundle } (f (\text{primal } x)) (f (\text{tangent } x)))$$

What is the tangent of $\#t$?

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What if we take $\overline{\#t} = \#f$?

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when f is a rearrangement function

$((j * f) x) = (\text{bundle } (f (\text{primal } x)) (f (\text{tangent } x)))$

What is the tangent of #t?

What if we take $\overline{\#t} = \#f$?

when f is a rearrangement function

$((j * f) x) = (\text{bundle } (f (\text{primal } x)) (f (\text{tangent } x)))$

$f : (\#t \ x \ y) \mapsto (\#t \ x \ y)$ but $f : (\#f \ x \ y) \mapsto (\#f \ y \ x)$

What is the tangent of #t?

What if we take $\overline{\#t} = \#f$?

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f is a rearrangement function

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f is a rearrangement function

$((j* f) (\text{bundle } (\#t x y) (\#f \overline{x} \overline{y})))$

What is the tangent of #t?

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f is a rearrangement function

$((j* f) (\text{bundle } (\#t x y) (\#f \overline{x} \overline{y})))$

$= (\text{bundle } (\#t x y) (\#f \overline{y} \overline{x}))$

What is the tangent of #t?

What if we take $\overline{\#t} = \#f$?

when f is a rearrangement function

$((j* f) x) = (\text{bundle } (f (\text{primal } x)) (f (\text{tangent } x)))$

$f : (\#t \ x \ y) \mapsto (\#t \ x \ y)$ but $f : (\#f \ x \ y) \mapsto (\#f \ y \ x)$

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What is the tangent of #t?

What if we take $\overline{\#t} = \#f$?

when f is a rearrangement function

$((j* f) x) = (\text{bundle } (f (\text{primal } x)) (f (\text{tangent } x)))$

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$((j* f) (\text{bundle } (\#t x y) (\#f \overline{x} \overline{y})))$

$= (\text{bundle } (\#t x y) (\#f \overline{y} \overline{x}))$

What is the tangent of $\#t$?

What if we take $\overline{\#t} = \#f$?

when f is a rearrangement function

$((j^* f) x) = (\text{bundle } (f (\text{primal } x)) (f (\text{tangent } x)))$

$f : (\#t x y) \mapsto (\#t x y)$ but $f : (\#f x y) \mapsto (\#f y x)$

f is a rearrangement function

$((j^* f) (\text{bundle } (\#t x y) (\#f \overline{x} \overline{y})))$

$= (\text{bundle } (\#t x y) (\#f \overline{y} \overline{x}))$

Problem avoided if we take $\overline{\#t} = \#t$

What is $(j^* \ j^*)$?

What is $(j^* \quad j^*)$?

when f is a rearrangement function

$$((j^* f) x) = (\text{bundle } (f \text{ (primal } x)) \text{ (} f \text{ (tangent } x)))$$

What is $(j^* \ j^*)$?

when f is a rearrangement function

$$((j^* f) x) = (\text{bundle } (f \text{ (primal } x)) \ (f \text{ (tangent } x)))$$

bundle, primal, tangent, and j^* are rearrangement functions

What is $(j^* j^*)$?

when f is a rearrangement function

$$((j^* f) x) = (\text{bundle } (f \text{ (primal } x)) \text{ (} f \text{ (tangent } x)))$$

bundle, primal, tangent, and j^* are rearrangement functions

$$((j^* \text{ bundle}) x) = (\text{bundle } (\text{bundle } (\text{primal } x)) \text{ (bundle } (\text{tangent } x)))$$

$$((j^* \text{ primal}) x) = (\text{bundle } (\text{primal } (\text{primal } x)) \text{ (primal } (\text{tangent } x)))$$

$$((j^* \text{ tangent}) x) = (\text{bundle } (\text{tangent } (\text{primal } x)) \text{ (tangent } (\text{tangent } x)))$$

$$((j^* j^*) x) = (\text{bundle } (j^* (\text{primal } x)) \text{ (} j^* \text{ (tangent } x)))$$

A (Not So) Brief Tutorial on AD

$$z = g(f(x))$$

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$$\begin{aligned}z &= g(f(x)) \\ &= (f \circ g)(x)\end{aligned}$$

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$$\begin{aligned}z &= g(f(x)) \\ &= (f \circ g)(x)\end{aligned}$$

$$\begin{aligned}y &= f(x) \\ z &= g(y)\end{aligned}$$

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$$\begin{aligned}z &= g(f(x)) \\ &= (f \circ g)(x)\end{aligned}$$

$$\begin{aligned}y &= f(x) \\ z &= g(y)\end{aligned}$$

$$\frac{dz}{dx} = \frac{dz}{dy} \frac{dy}{dx}$$

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$$\begin{aligned}z &= g(f(x)) \\ &= (f \circ g)(x)\end{aligned}$$

$$\begin{aligned}y &= f(x) \\ z &= g(y)\end{aligned}$$

$$\frac{dz}{dx} = \frac{dz}{dy} \frac{dy}{dx}$$

$$\mathcal{D} (f \circ g) x = (\mathcal{D} g y) \times (\mathcal{D} f x)$$

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$$f = f_1 \circ \cdots \circ f_n$$

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$$f = f_1 \circ \cdots \circ f_n$$

$$\mathbf{x}_1 = f_1 \mathbf{x}_0$$

$$\vdots$$

$$\mathbf{x}_n = f_n \mathbf{x}_{n-1}$$

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$$f = f_1 \circ \cdots \circ f_n$$

$$\mathbf{x}_1 = f_1 \mathbf{x}_0$$

$$\vdots$$

$$\mathbf{x}_n = f_n \mathbf{x}_{n-1}$$

$$\mathcal{J} f \mathbf{x}_0 = (\mathcal{J} f_n \mathbf{x}_{n-1}) \times \cdots \times (\mathcal{J} f_1 \mathbf{x}_0)$$

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$$f = f_1 \circ \cdots \circ f_n$$

$$\mathbf{x}_1 = f_1 \mathbf{x}_0$$

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$$\mathcal{J} f \mathbf{x}_0 = (\mathcal{J} f_n \mathbf{x}_{n-1}) \times \cdots \times (\mathcal{J} f_1 \mathbf{x}_0)$$

$$(\mathcal{J} f \mathbf{x}_0)^\top = (\mathcal{J} f_1 \mathbf{x}_0)^\top \times \cdots \times (\mathcal{J} f_n \mathbf{x}_{n-1})^\top$$

A (Not So) Brief Tutorial on AD

$$\overline{\mathbf{X}}_n = \mathcal{J} f \mathbf{x}_0$$

A (Not So) Brief Tutorial on AD

$$\begin{aligned}\overline{\mathbf{X}}_n &= \mathcal{J} f \mathbf{x}_0 \\ &= (\mathcal{J} f_n \mathbf{x}_{n-1}) \times \cdots \times (\mathcal{J} f_2 \mathbf{x}_1) \times (\mathcal{J} f_1 \mathbf{x}_0)\end{aligned}$$

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$$\begin{aligned}\overline{\mathbf{X}}_n &= \mathcal{J} f \mathbf{x}_0 \\ &= (\mathcal{J} f_n \mathbf{x}_{n-1}) \times \cdots \times (\mathcal{J} f_2 \mathbf{x}_1) \times (\mathcal{J} f_1 \mathbf{x}_0)\end{aligned}$$

$$\overline{\mathbf{X}}_1 = (\mathcal{J} f_1 \mathbf{x}_0)$$

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$$\begin{aligned}\overline{\mathbf{X}}_n &= \mathcal{J} f \mathbf{x}_0 \\ &= (\mathcal{J} f_n \mathbf{x}_{n-1}) \times \cdots \times (\mathcal{J} f_2 \mathbf{x}_1) \times (\mathcal{J} f_1 \mathbf{x}_0)\end{aligned}$$

$$\begin{aligned}\overline{\mathbf{X}}_1 &= (\mathcal{J} f_1 \mathbf{x}_0) \\ \overline{\mathbf{X}}_2 &= (\mathcal{J} f_2 \mathbf{x}_1) \times \overline{\mathbf{X}}_1\end{aligned}$$

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$$\begin{aligned}\overline{\mathbf{X}}_n &= \mathcal{J} f \mathbf{x}_0 \\ &= (\mathcal{J} f_n \mathbf{x}_{n-1}) \times \cdots \times (\mathcal{J} f_2 \mathbf{x}_1) \times (\mathcal{J} f_1 \mathbf{x}_0)\end{aligned}$$

$$\overline{\mathbf{X}}_1 = (\mathcal{J} f_1 \mathbf{x}_0)$$

$$\overline{\mathbf{X}}_2 = (\mathcal{J} f_2 \mathbf{x}_1) \times \overline{\mathbf{X}}_1$$

⋮

$$\overline{\mathbf{X}}_n = (\mathcal{J} f_n \mathbf{x}_{n-1}) \times \overline{\mathbf{X}}_{n-1}$$

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$$\overline{\mathbf{X}_0} = (\mathcal{J} f \mathbf{x}_0)^\top$$

A (Not So) Brief Tutorial on AD

$$\begin{aligned}\overline{\mathbf{X}_0} &= (\mathcal{J} f \mathbf{x}_0)^\top \\ &= (\mathcal{J} f_1 \mathbf{x}_0)^\top \times \cdots \times (\mathcal{J} f_{n-1} \mathbf{x}_{n-2})^\top \times (\mathcal{J} f_n \mathbf{x}_{n-1})^\top\end{aligned}$$

A (Not So) Brief Tutorial on AD

$$\begin{aligned}\overline{\mathbf{X}_0} &= (\mathcal{J} f \mathbf{x}_0)^\top \\ &= (\mathcal{J} f_1 \mathbf{x}_0)^\top \times \cdots \times (\mathcal{J} f_{n-1} \mathbf{x}_{n-2})^\top \times (\mathcal{J} f_n \mathbf{x}_{n-1})^\top\end{aligned}$$

$$\overline{\mathbf{X}_{n-1}} = (\mathcal{J} f_n \mathbf{x}_{n-1})^\top$$

A (Not So) Brief Tutorial on AD

$$\begin{aligned}\overline{\mathbf{X}_0} &= (\mathcal{J} f \mathbf{x}_0)^\top \\ &= (\mathcal{J} f_1 \mathbf{x}_0)^\top \times \cdots \times (\mathcal{J} f_{n-1} \mathbf{x}_{n-2})^\top \times (\mathcal{J} f_n \mathbf{x}_{n-1})^\top\end{aligned}$$

$$\overline{\mathbf{X}_{n-1}} = (\mathcal{J} f_n \mathbf{x}_{n-1})^\top$$

$$\overline{\mathbf{X}_{n-2}} = (\mathcal{J} f_{n-1} \mathbf{x}_{n-2})^\top \times \overline{\mathbf{X}_{n-1}}$$

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$$\begin{aligned}\overline{\mathbf{X}}_0 &= (\mathcal{J} f \mathbf{x}_0)^\top \\ &= (\mathcal{J} f_1 \mathbf{x}_0)^\top \times \cdots \times (\mathcal{J} f_{n-1} \mathbf{x}_{n-2})^\top \times (\mathcal{J} f_n \mathbf{x}_{n-1})^\top\end{aligned}$$

$$\begin{aligned}\overline{\mathbf{X}}_{n-1} &= (\mathcal{J} f_n \mathbf{x}_{n-1})^\top \\ \overline{\mathbf{X}}_{n-2} &= (\mathcal{J} f_{n-1} \mathbf{x}_{n-2})^\top \times \overline{\mathbf{X}}_{n-1} \\ &\vdots \\ \overline{\mathbf{X}}_0 &= (\mathcal{J} f_1 \mathbf{x}_0)^\top \times \overline{\mathbf{X}}_1\end{aligned}$$

A (Not So) Brief Tutorial on AD

$$\begin{array}{lcl} \overline{\mathbf{X}}_1 & = & (\mathcal{J} f_1 \mathbf{x}_0) \\ \overline{\mathbf{X}}_2 & = & (\mathcal{J} f_2 \mathbf{x}_1) \times \overline{\mathbf{X}}_1 \\ & \vdots & \\ \overline{\mathbf{X}}_n & = & (\mathcal{J} f_n \mathbf{x}_{n-1}) \times \overline{\mathbf{X}}_{n-1} \end{array} \qquad \begin{array}{lcl} \overline{\mathbf{X}}_{n-1} & = & (\mathcal{J} f_n \mathbf{x}_{n-1})^\top \\ \overline{\mathbf{X}}_{n-2} & = & (\mathcal{J} f_{n-1} \mathbf{x}_{n-2})^\top \times \overline{\mathbf{X}}_{n-1} \\ & \vdots & \\ \overline{\mathbf{X}}_0 & = & (\mathcal{J} f_1 \mathbf{x}_0)^\top \times \overline{\mathbf{X}}_1 \end{array}$$

A (Not So) Brief Tutorial on AD

$$\begin{array}{ll} \overline{\mathbf{X}}_1 & = (\mathcal{J} f_1 \mathbf{x}_0) \\ \overline{\mathbf{X}}_2 & = (\mathcal{J} f_2 \mathbf{x}_1) \times \overline{\mathbf{X}}_1 \\ & \vdots \\ \overline{\mathbf{X}}_n & = (\mathcal{J} f_n \mathbf{x}_{n-1}) \times \overline{\mathbf{X}}_{n-1} \end{array} \quad \begin{array}{ll} \overline{\mathbf{X}}_{n-1} & = (\mathcal{J} f_n \mathbf{x}_{n-1})^\top \\ \overline{\mathbf{X}}_{n-2} & = (\mathcal{J} f_{n-1} \mathbf{x}_{n-2})^\top \times \overline{\mathbf{X}}_{n-1} \\ & \vdots \\ \overline{\mathbf{X}}_0 & = (\mathcal{J} f_1 \mathbf{x}_0)^\top \times \overline{\mathbf{X}}_1 \end{array}$$

A (Not So) Brief Tutorial on AD

$$\begin{array}{ll} \overline{\mathbf{X}}_1 & = (\mathcal{J} f_1 \mathbf{x}_0) \\ \overline{\mathbf{X}}_2 & = (\mathcal{J} f_2 \mathbf{x}_1) \times \overline{\mathbf{X}}_1 \\ \vdots & \\ \overline{\mathbf{X}}_n & = (\mathcal{J} f_n \mathbf{x}_{n-1}) \times \overline{\mathbf{X}}_{n-1} \end{array} \qquad \begin{array}{ll} \overline{\mathbf{X}}_{n-1} & = (\mathcal{J} f_n \mathbf{x}_{n-1})^\top \\ \overline{\mathbf{X}}_{n-2} & = (\mathcal{J} f_{n-1} \mathbf{x}_{n-2})^\top \times \overline{\mathbf{X}}_{n-1} \\ \vdots & \\ \overline{\mathbf{X}}_0 & = (\mathcal{J} f_1 \mathbf{x}_0)^\top \times \overline{\mathbf{X}}_1 \end{array}$$

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$$\overline{\mathbf{x}}_n = (\mathcal{J} f \mathbf{x}_0) \times \overline{\mathbf{x}}_0$$

A (Not So) Brief Tutorial on AD

$$\begin{aligned}\overline{\mathbf{x}}_n &= (\mathcal{J} f \mathbf{x}_0) \times \overline{\mathbf{x}}_0 \\ &= (\mathcal{J} f_n \mathbf{x}_{n-1}) \times \cdots \times (\mathcal{J} f_1 \mathbf{x}_0) \times \overline{\mathbf{x}}_0\end{aligned}$$

A (Not So) Brief Tutorial on AD

$$\begin{aligned}\overline{\mathbf{x}}_n &= (\mathcal{J} f \mathbf{x}_0) \times \overline{\mathbf{x}}_0 \\ &= (\mathcal{J} f_n \mathbf{x}_{n-1}) \times \cdots \times (\mathcal{J} f_1 \mathbf{x}_0) \times \overline{\mathbf{x}}_0\end{aligned}$$

$$\overline{\mathbf{x}}_1 = (\mathcal{J} f_1 \mathbf{x}_0) \times \overline{\mathbf{x}}_0$$

\vdots

$$\overline{\mathbf{x}}_n = (\mathcal{J} f_n \mathbf{x}_{n-1}) \times \overline{\mathbf{x}}_{n-1}$$

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$$\overline{\mathbf{x}_0} = (\mathcal{J} f \mathbf{x}_0)^\top \times \overline{\mathbf{x}_n}$$

A (Not So) Brief Tutorial on AD

$$\begin{aligned}\overline{\mathbf{x}_0} &= (\mathcal{J} f \mathbf{x}_0)^\top \times \overline{\mathbf{x}_n} \\ &= (\mathcal{J} f_1 \mathbf{x}_0)^\top \times \cdots \times (\mathcal{J} f_n \mathbf{x}_{n-1})^\top \times \overline{\mathbf{x}_n}\end{aligned}$$

A (Not So) Brief Tutorial on AD

$$\begin{aligned}\overline{\mathbf{x}_0} &= (\mathcal{J} f \mathbf{x}_0)^\top \times \overline{\mathbf{x}_n} \\ &= (\mathcal{J} f_1 \mathbf{x}_0)^\top \times \cdots \times (\mathcal{J} f_n \mathbf{x}_{n-1})^\top \times \overline{\mathbf{x}_n}\end{aligned}$$

$$\begin{aligned}\overline{\mathbf{x}_{n-1}} &= (\mathcal{J} f_n \mathbf{x}_{n-1})^\top \times \overline{\mathbf{x}_n} \\ &\vdots \\ \overline{\mathbf{x}_0} &= (\mathcal{J} f_1 \mathbf{x}_0)^\top \times \overline{\mathbf{x}_1}\end{aligned}$$

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$$\mathbf{y} = \mathbf{A} \times \mathbf{x}$$

A (Not So) Brief Tutorial on AD

$$\begin{aligned}\mathbf{y} &= \mathbf{A} \times \mathbf{x} \\ &= f(\mathbf{x})\end{aligned}$$

A (Not So) Brief Tutorial on AD

$$\begin{aligned}\mathbf{y} &= \mathbf{A} \times \mathbf{x} \\ &= f(\mathbf{x})\end{aligned}$$

$$\overline{\mathbf{x}}'_n = (\mathcal{J} f \mathbf{x}_0) \times \overline{\mathbf{x}}'_0$$

A (Not So) Brief Tutorial on AD

$$\begin{aligned}\mathbf{y} &= \mathbf{A} \times \mathbf{x} \\ &= f(\mathbf{x})\end{aligned}$$

$$\begin{aligned}\overline{\mathbf{x}}_n &= (\mathcal{J} f \mathbf{x}_0) \times \overline{\mathbf{x}}_0 \\ &= \overline{f'} \mathbf{x}_0 \overline{\mathbf{x}}_0\end{aligned}$$

A (Not So) Brief Tutorial on AD

$$\begin{aligned}\mathbf{y} &= \mathbf{A} \times \mathbf{x} \\ &= f(\mathbf{x})\end{aligned}$$

$$\begin{aligned}\overline{\mathbf{x}}_n &= (\mathcal{J} f \mathbf{x}_0) \times \overline{\mathbf{x}}_0 \\ &= \overline{f'} \mathbf{x}_0 \overline{\mathbf{x}}_0\end{aligned}$$

$$\overline{\mathbf{x}}_0 = (\mathcal{J} f \mathbf{x}_0)^\top \times \overline{\mathbf{x}}_n$$

A (Not So) Brief Tutorial on AD

$$\begin{aligned}\mathbf{y} &= \mathbf{A} \times \mathbf{x} \\ &= f(\mathbf{x})\end{aligned}$$

$$\begin{aligned}\overline{\mathbf{x}}_n &= (\mathcal{J} f \mathbf{x}_0) \times \overline{\mathbf{x}}_0 \\ &= \overline{f} \mathbf{x}_0 \overline{\mathbf{x}}_0\end{aligned}$$

$$\begin{aligned}\overleftarrow{\mathbf{x}}_0 &= (\mathcal{J} f \mathbf{x}_0)^\top \times \overleftarrow{\mathbf{x}}_n \\ &= \overleftarrow{f} \mathbf{x}_0 \overleftarrow{\mathbf{x}}_n\end{aligned}$$

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$$\overline{\mathbf{X}}_n[:,j] = \overline{f}' \mathbf{x}_0 \overline{\mathbf{e}}_j$$

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$$\begin{aligned}\overline{\mathbf{x}}_n[;j] &= \overline{f'} \mathbf{x}_0 \overline{\mathbf{e}}_j \\ &= (\mathcal{J} f \mathbf{x}_0)[;j]\end{aligned}$$

A (Not So) Brief Tutorial on AD

$$\begin{aligned}\overline{\mathbf{X}}_n[j] &= \overline{f} \mathbf{x}_0 \overline{\mathbf{e}}_j \\ &= (\mathcal{J} f \mathbf{x}_0)[j]\end{aligned}$$

$$\underline{\mathbf{X}}_0[i] = \underline{f} \mathbf{x}_0 \underline{\mathbf{e}}_i$$

A (Not So) Brief Tutorial on AD

$$\begin{aligned}\overline{\mathbf{X}}_n[j] &= \overline{f} \mathbf{x}_0 \overline{\mathbf{e}}_j \\ &= (\mathcal{J} f \mathbf{x}_0)[j]\end{aligned}$$

$$\begin{aligned}\overleftarrow{\mathbf{X}}_0[i] &= \overleftarrow{f} \mathbf{x}_0 \overleftarrow{\mathbf{e}}_i \\ &= (\mathcal{J} f \mathbf{x}_0)^\top[i]\end{aligned}$$

A (Not So) Brief Tutorial on AD

$$\begin{aligned}\overline{\mathbf{X}}_n[j] &= \overline{f} \mathbf{x}_0 \overline{\mathbf{e}}_j \\ &= (\mathcal{J} f \mathbf{x}_0)[j]\end{aligned}$$

$$\begin{aligned}\overleftarrow{\mathbf{X}}_0[i] &= \overleftarrow{f} \mathbf{x}_0 \overleftarrow{\mathbf{e}}_i \\ &= (\mathcal{J} f \mathbf{x}_0)^\top[i] \\ &= (\mathcal{J} f \mathbf{x}_0)[i;]\end{aligned}$$

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$$\mathbf{y} = \mathbf{B} \times (\mathbf{A} \times \mathbf{x})$$

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$$\begin{aligned}\mathbf{y} &= \mathbf{B} \times (\mathbf{A} \times \mathbf{x}) \\ &= (\mathbf{B} \times \mathbf{A}) \times \mathbf{x}\end{aligned}$$

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$$\begin{aligned}\mathbf{y} &= \mathbf{B} \times (\mathbf{A} \times \mathbf{x}) \\ &= (\mathbf{B} \times \mathbf{A}) \times \mathbf{x} \\ &= g(f(\mathbf{x}))\end{aligned}$$

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$$\begin{aligned}\mathbf{y} &= \mathbf{B} \times (\mathbf{A} \times \mathbf{x}) \\ &= (\mathbf{B} \times \mathbf{A}) \times \mathbf{x} \\ &= g(f(\mathbf{x})) \\ &= (f \circ g)(\mathbf{x})\end{aligned}$$

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$$\begin{aligned}\mathbf{y} &= \mathbf{B} \times (\mathbf{A} \times \mathbf{x}) \\ &= (\mathbf{B} \times \mathbf{A}) \times \mathbf{x} \\ &= g(f(\mathbf{x})) \\ &= (f \circ g)(\mathbf{x})\end{aligned}$$

$$(\mathcal{J} f_n \mathbf{x}_{n-1}) \times \cdots \times (\mathcal{J} f_1 \mathbf{x}_0) = (\overline{f_1}' \mathbf{x}_0) \circ \cdots \circ (\overline{f_n}' \mathbf{x}_{n-1})$$

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$$\begin{aligned}\mathbf{y} &= \mathbf{B} \times (\mathbf{A} \times \mathbf{x}) \\ &= (\mathbf{B} \times \mathbf{A}) \times \mathbf{x} \\ &= g(f(\mathbf{x})) \\ &= (f \circ g)(\mathbf{x})\end{aligned}$$

$$\begin{aligned}(\mathcal{J} f_n \mathbf{x}_{n-1}) \times \cdots \times (\mathcal{J} f_1 \mathbf{x}_0) &= (\overline{f_1} \mathbf{x}_0) \circ \cdots \circ (\overline{f_n} \mathbf{x}_{n-1}) \\ (\mathcal{J} f_1 \mathbf{x}_0)^\top \times \cdots \times (\mathcal{J} f_n \mathbf{x}_{n-1})^\top &= (\overline{f_n} \mathbf{x}_{n-1}) \circ \cdots \circ (\overline{f_1} \mathbf{x}_0)\end{aligned}$$

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$$\mathbf{x}_{i-1} = (\boxed{\mathbf{x}_{i-1}[1]} \cdots \boxed{\mathbf{x}_{i-1}[L_i]} \cdots \boxed{\mathbf{x}_{i-1}[R_i]} \cdots \boxed{\mathbf{x}_{i-1}[m]})$$
$$\mathbf{x}_i = f_i \mathbf{x}_{i-1} = (\boxed{\mathbf{x}_{i-1}[1]} \cdots \boxed{u_i \mathbf{x}_{i-1}[R_i]} \cdots \boxed{\mathbf{x}_{i-1}[R_i]} \cdots \boxed{\mathbf{x}_{i-1}[m]})$$

$$\mathbf{x}[L_i] := u_i \mathbf{x}[R_i]$$

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$$\mathbf{x}_{i-1} = (\boxed{\mathbf{x}_{i-1}[1]} \cdots \boxed{\mathbf{x}_{i-1}[L_i]} \cdots \boxed{\mathbf{x}_{i-1}[R_i]} \cdots \boxed{\mathbf{x}_{i-1}[m]})$$
$$\mathbf{x}_i = f_i \mathbf{x}_{i-1} = (\boxed{\mathbf{x}_{i-1}[1]} \cdots \boxed{u_i \mathbf{x}_{i-1}[R_i]} \cdots \boxed{\mathbf{x}_{i-1}[R_i]} \cdots \boxed{\mathbf{x}_{i-1}[m]})$$

$$\mathbf{x}[L_i] := u_i \mathbf{x}[R_i]$$

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$$\mathbf{x}_{i-1} = (\boxed{\mathbf{x}_{i-1}[1]} \cdots \boxed{\mathbf{x}_{i-1}[L_i]} \cdots \boxed{\mathbf{x}_{i-1}[R_i]} \cdots \boxed{\mathbf{x}_{i-1}[m]})$$
$$\mathbf{x}_i = f_i \mathbf{x}_{i-1} = (\boxed{\mathbf{x}_{i-1}[1]} \cdots \boxed{u_i \mathbf{x}_{i-1}[R_i]} \cdots \boxed{\mathbf{x}_{i-1}[R_i]} \cdots \boxed{\mathbf{x}_{i-1}[m]})$$

$$\mathbf{x}[L_i] := u_i \mathbf{x}[R_i]$$

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$$\mathbf{x}_{i-1} = (\boxed{\mathbf{x}_{i-1}[1]} \cdots \boxed{\mathbf{x}_{i-1}[L_i]} \cdots \boxed{\mathbf{x}_{i-1}[R_i]} \cdots \boxed{\mathbf{x}_{i-1}[S_i]} \cdots \boxed{\mathbf{x}_{i-1}[m]})$$
$$\mathbf{x}_i = f_i \mathbf{x}_{i-1} = (\boxed{\mathbf{x}_{i-1}[1]} \cdots \boxed{b_i(\mathbf{x}_{i-1}[R_i], \mathbf{x}_{i-1}[S_i])} \cdots \boxed{\mathbf{x}_{i-1}[R_i]} \cdots \boxed{\mathbf{x}_{i-1}[S_i]} \cdots \boxed{\mathbf{x}_{i-1}[m]})$$

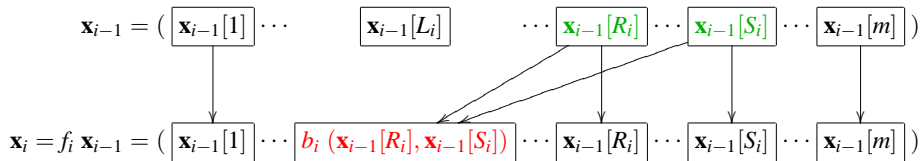
$$\mathbf{x}[L_i] := b(\mathbf{x}[R_i], \mathbf{x}[S_i])$$

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$$\mathbf{x}_{i-1} = (\boxed{\mathbf{x}_{i-1}[1]} \cdots \boxed{\mathbf{x}_{i-1}[L_i]} \cdots \boxed{\mathbf{x}_{i-1}[R_i]} \cdots \boxed{\mathbf{x}_{i-1}[S_i]} \cdots \boxed{\mathbf{x}_{i-1}[m]})$$
$$\mathbf{x}_i = f_i \mathbf{x}_{i-1} = (\boxed{\mathbf{x}_{i-1}[1]} \cdots \boxed{b_i(\mathbf{x}_{i-1}[R_i], \mathbf{x}_{i-1}[S_i])} \cdots \boxed{\mathbf{x}_{i-1}[R_i]} \cdots \boxed{\mathbf{x}_{i-1}[S_i]} \cdots \boxed{\mathbf{x}_{i-1}[m]})$$

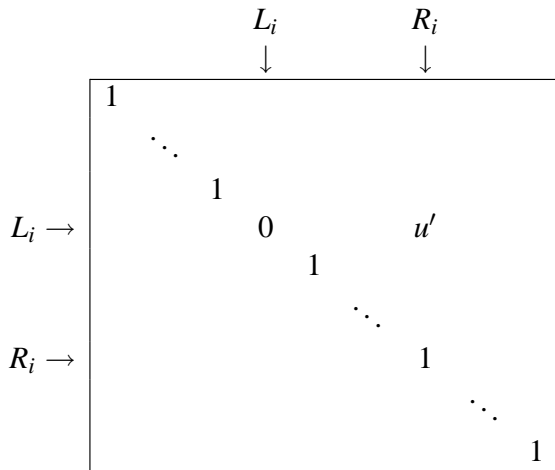
$$\mathbf{x}[L_i] := b(\mathbf{x}[R_i], \mathbf{x}[S_i])$$

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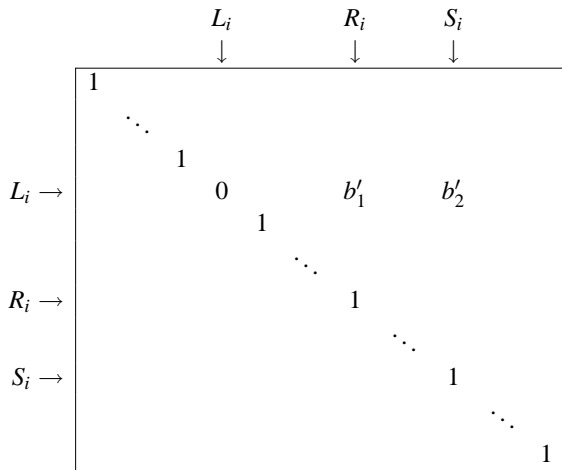
$$\mathbf{x}[L_i] := b(\mathbf{x}[R_i], \mathbf{x}[S_i])$$

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$$u' = \mathcal{D} u_i \mathbf{x}_{i-1}[R_i]$$

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$$b'_1 = \mathcal{D}_1 b_i(\mathbf{x}_{i-1}[R_i], \mathbf{x}_{i-1}[S_i])$$

$$b'_2 = \mathcal{D}_2 b_i(\mathbf{x}_{i-1}[R_i], \mathbf{x}_{i-1}[S_i])$$

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$$\overline{\mathbf{x}}_i' = \overline{f}_i \mathbf{x}_{i-1} \overline{\mathbf{x}}_{i-1}'$$

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$$\begin{aligned}\overline{\mathbf{x}}_i &= \overline{f}_i \mathbf{x}_{i-1} \overline{\mathbf{x}}_{i-1} \\ &= (\mathcal{J} f_i \mathbf{x}_{i-1}) \times \overline{\mathbf{x}}_{i-1}\end{aligned}$$

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$$\overline{\mathbf{x}}_i = \overline{f}_i \mathbf{x}_{i-1} \overline{\mathbf{x}}_{i-1}$$

A (Not So) Brief Tutorial on AD

$$\begin{aligned}\overline{\mathbf{x}}_i &= \overline{f}_i \mathbf{x}_{i-1} \overline{\mathbf{x}}_{i-1} \\ &= (\mathcal{J} f_i \mathbf{x}_{i-1}) \times \overline{\mathbf{x}}_{i-1}\end{aligned}$$

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$$\overline{\mathbf{x}_{i-1}} = \overline{f_i} \mathbf{x}_{i-1} \overline{\mathbf{x}_i}$$

A (Not So) Brief Tutorial on AD

$$\begin{aligned}\overline{\mathbf{x}}_{i-1} &= \overline{f}_i \mathbf{x}_{i-1} \overline{\mathbf{x}}_i \\ &= (\mathcal{J} f_i \mathbf{x}_{i-1})^\top \times \overline{\mathbf{x}}_i\end{aligned}$$

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$$\overline{\mathbf{x}_{i-1}} = \overline{f_i} \mathbf{x}_{i-1} \overline{\mathbf{x}_i}$$

A (Not So) Brief Tutorial on AD

$$\begin{aligned}\overline{\mathbf{x}}_{i-1} &= \overline{f_i} \mathbf{x}_{i-1} \overline{\mathbf{x}}_i \\ &= (\mathcal{J} f_i \mathbf{x}_{i-1})^\top \times \overline{\mathbf{x}}_i\end{aligned}$$

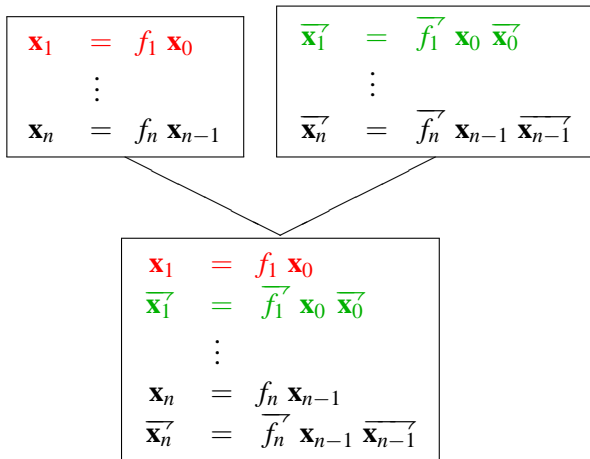
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$$\begin{aligned}\mathbf{x}_1 &= f_1 \mathbf{x}_0 \\ &\vdots \\ \mathbf{x}_n &= f_n \mathbf{x}_{n-1}\end{aligned}$$

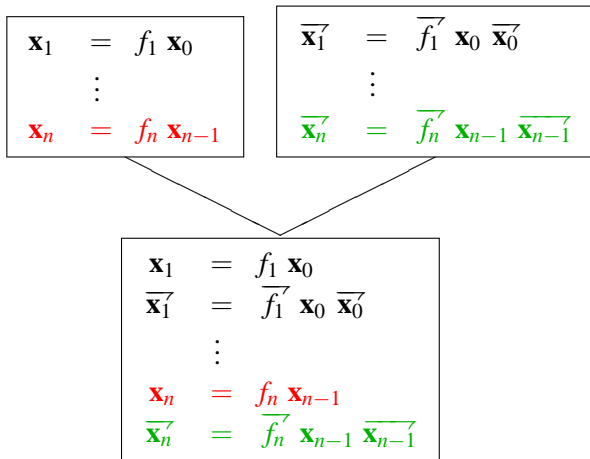
$$\begin{aligned}\overline{\mathbf{x}}_1' &= \overline{f_1}' \mathbf{x}_0 \overline{\mathbf{x}}_0' \\ &\vdots \\ \overline{\mathbf{x}}_n' &= \overline{f_n}' \mathbf{x}_{n-1} \overline{\mathbf{x}}_{n-1}'\end{aligned}$$

$$\begin{aligned}\mathbf{x}_1 &= f_1 \mathbf{x}_0 \\ \overline{\mathbf{x}}_1' &= \overline{f_1}' \mathbf{x}_0 \overline{\mathbf{x}}_0' \\ &\vdots \\ \mathbf{x}_n &= f_n \mathbf{x}_{n-1} \\ \overline{\mathbf{x}}_n' &= \overline{f_n}' \mathbf{x}_{n-1} \overline{\mathbf{x}}_{n-1}'\end{aligned}$$

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$$\begin{aligned}\mathbf{x}_1 &= f_1 \mathbf{x}_0 \\ \overline{\mathbf{x}}_1' &= \overline{f_1}' \mathbf{x}_0 \overline{\mathbf{x}}_0' \\ &\vdots \\ \mathbf{x}_n &= f_n \mathbf{x}_{n-1} \\ \overline{\mathbf{x}}_n' &= \overline{f_n}' \mathbf{x}_{n-1} \overline{\mathbf{x}}_{n-1}'\end{aligned}$$

$$\begin{aligned}(\mathbf{x}_1, \overline{\mathbf{x}}_1') &= ((f_1 \mathbf{x}_0), (\overline{f_1}' \mathbf{x}_0 \overline{\mathbf{x}}_0')) \\ &\vdots \\ (\mathbf{x}_n, \overline{\mathbf{x}}_n') &= ((f_n \mathbf{x}_{n-1}), (\overline{f_n}' \mathbf{x}_{n-1} \overline{\mathbf{x}}_{n-1}'))\end{aligned}$$

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$$\left. \begin{array}{l} \mathbf{x}_1 = f_1 \mathbf{x}_0 \\ \vdots \\ \mathbf{x}_n = f_n \mathbf{x}_{n-1} \end{array} \right\} \rightsquigarrow \left\{ \begin{array}{l} \overrightarrow{\mathbf{x}}_1 = \overrightarrow{f}_1 \overrightarrow{\mathbf{x}}_0 \\ \vdots \\ \overrightarrow{\mathbf{x}}_n = \overrightarrow{f}_n \overrightarrow{\mathbf{x}}_{n-1} \end{array} \right.$$

$$\begin{aligned} \overrightarrow{\mathbf{x}} &\equiv (\mathbf{x}, \overline{\mathbf{x}}) \\ \overrightarrow{f} \overrightarrow{\mathbf{x}} &\equiv ((f \mathbf{x}), (\overline{f} \mathbf{x} \overline{\mathbf{x}})) \end{aligned}$$

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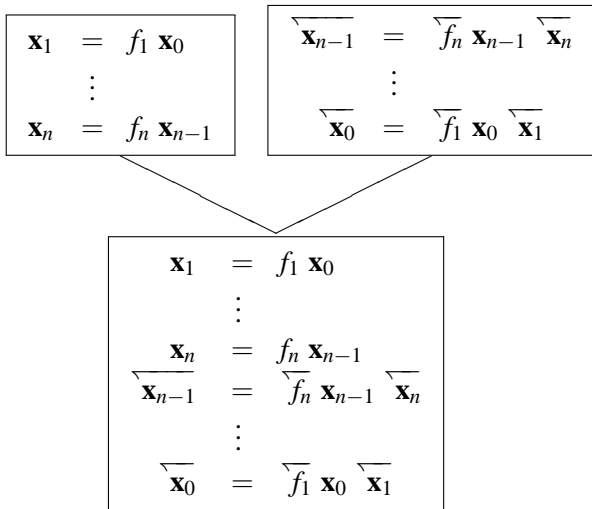
$$\begin{aligned}x_{L_i} &:= u_i x_{R_i} &\rightsquigarrow& \overrightarrow{x_{L_i}} := \overrightarrow{u_i} \overrightarrow{x_{R_i}} \\x_{L_i} &:= b_i (x_{R_i}, x_{S_i}) &\rightsquigarrow& \overrightarrow{x_{L_i}} := \overrightarrow{b_i} (\overrightarrow{x_{R_i}}, \overrightarrow{x_{S_i}})\end{aligned}$$

$$\overrightarrow{x} \equiv (x, \overline{x'})$$

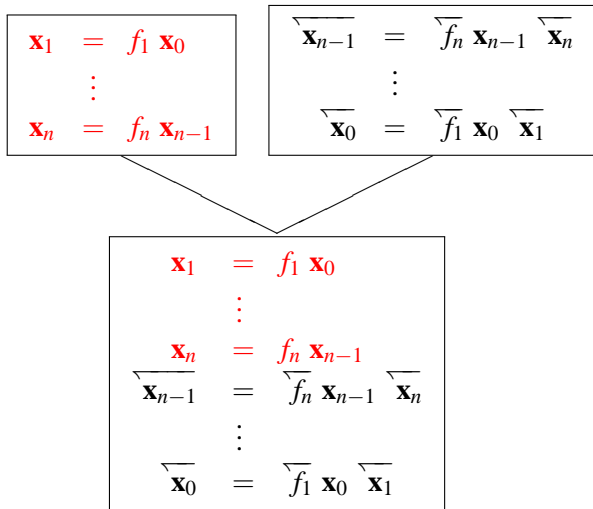
$$\overrightarrow{u} \overrightarrow{x} \equiv ((u x), ((\mathcal{D} u x) \times \overline{x'}))$$

$$\overrightarrow{b} (\overrightarrow{x_1}, \overrightarrow{x_2}) \equiv ((b(x_1, x_2)), ((\mathcal{D}_1 b(x_1, x_2)) \times \overline{x'_1}) + ((\mathcal{D}_2 b(x_1, x_2)) \times \overline{x'_2})))$$

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$$\mathbf{x}_1 = f_1 \mathbf{x}_0$$

$$\vdots$$

$$\mathbf{x}_n = f_n \mathbf{x}_{n-1}$$

$$\overline{\mathbf{x}_{n-1}} = \overline{f_n} \mathbf{x}_{n-1} \overline{\mathbf{x}_n}$$

$$\vdots$$

$$\overline{\mathbf{x}_0} = \overline{f_1} \mathbf{x}_0 \overline{\mathbf{x}_1}$$

$$\mathbf{x}_1 = f_1 \mathbf{x}_0$$

$$\vdots$$

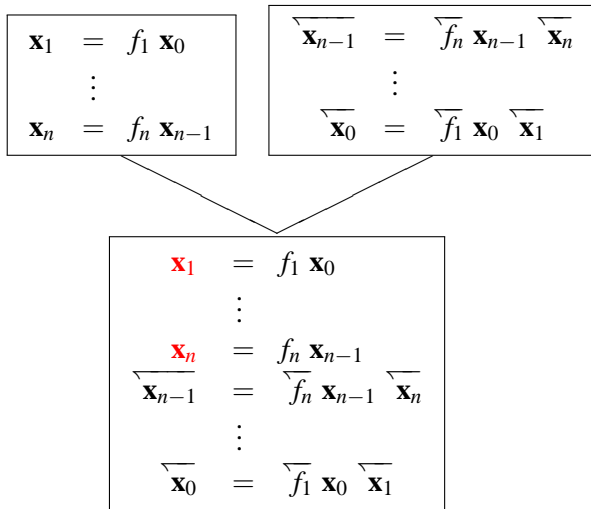
$$\mathbf{x}_n = f_n \mathbf{x}_{n-1}$$

$$\overline{\mathbf{x}_{n-1}} = \overline{f_n} \mathbf{x}_{n-1} \overline{\mathbf{x}_n}$$

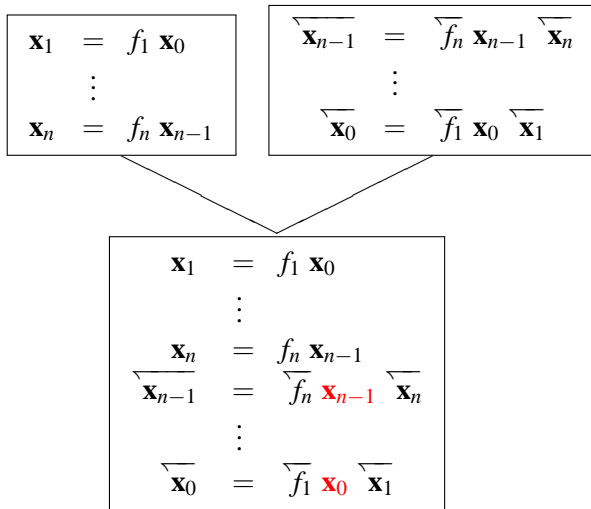
$$\vdots$$

$$\overline{\mathbf{x}_0} = \overline{f_1} \mathbf{x}_0 \overline{\mathbf{x}_1}$$

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$$\mathbf{x}_1 = f_1 \mathbf{x}_0$$

$$\overline{\mathbf{x}}_1 = \lambda \overline{\mathbf{x}} \quad \overline{\mathbf{x}}_0 (\overline{f}_1 \mathbf{x}_0 \overline{\mathbf{x}})$$

⋮

$$\mathbf{x}_n = f_n \mathbf{x}_{n-1}$$

$$\overline{\mathbf{x}}_n = \lambda \overline{\mathbf{x}} \quad \overline{\mathbf{x}}_{n-1} (\overline{f}_1 \mathbf{x}_0 \overline{\mathbf{x}})$$

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$$\begin{aligned}\mathbf{x}_1 &= f_1 \mathbf{x}_0 \\ \overline{\mathbf{x}}_1 &= \lambda \overline{\mathbf{x}} \ \overline{\mathbf{x}}_0 (\overline{f_1} \mathbf{x}_0 \ \overline{\mathbf{x}}) \\ &\vdots \\ \mathbf{x}_n &= f_n \mathbf{x}_{n-1} \\ \overline{\mathbf{x}}_n &= \lambda \overline{\mathbf{x}} \ \overline{\mathbf{x}}_{n-1} (\overline{f_n} \mathbf{x}_0 \ \overline{\mathbf{x}})\end{aligned}$$

$$\begin{aligned}(\mathbf{x}_1, \overline{\mathbf{x}}_1) &= ((f_1 \mathbf{x}_0), (\lambda \overline{\mathbf{x}} \ \overline{\mathbf{x}}_0 (\overline{f_1} \mathbf{x}_0 \ \overline{\mathbf{x}}))) \\ &\vdots \\ (\mathbf{x}_n, \overline{\mathbf{x}}_n) &= ((f_n \mathbf{x}_{n-1}), (\lambda \overline{\mathbf{x}} \ \overline{\mathbf{x}}_{n-1} (\overline{f_n} \mathbf{x}_0 \ \overline{\mathbf{x}})))\end{aligned}$$

A (Not So) Brief Tutorial on AD

$$\left. \begin{array}{l} \mathbf{x}_1 = f_1 \mathbf{x}_0 \\ \vdots \\ \mathbf{x}_n = f_n \mathbf{x}_{n-1} \end{array} \right\} \rightsquigarrow \left\{ \begin{array}{l} \overleftarrow{\mathbf{x}}_1 = \overleftarrow{f}_1 \overleftarrow{\mathbf{x}}_0 \\ \vdots \\ \overleftarrow{\mathbf{x}}_n = \overleftarrow{f}_n \overleftarrow{\mathbf{x}}_{n-1} \end{array} \right.$$

$$\begin{aligned} \overleftarrow{\mathbf{x}} &\equiv (\mathbf{x}, \overline{\mathbf{x}}) \\ \overleftarrow{f} \overleftarrow{\mathbf{x}} &\equiv ((f \mathbf{x}), (\lambda \overline{\mathbf{x}} \overline{\mathbf{x}} (\overline{f} \mathbf{x} \overline{\mathbf{x}}))) \end{aligned}$$

A (Not So) Brief Tutorial on AD

$$\overleftarrow{f} \mathbf{x} \equiv \mathbf{begin} \ \bar{x} := \lambda \overleftarrow{\mathbf{x}} \ \bar{x} (\overleftarrow{f} \ \mathbf{x} \ \overleftarrow{\mathbf{x}}); \\ (f \ \mathbf{x}) \ \mathbf{end}$$

A (Not So) Brief Tutorial on AD

$$\begin{array}{l}
 x_{L_i} := u_i x_{R_i} \quad \rightsquigarrow \\
 \\
 x_{L_i} := b_i(x_{R_i}, x_{S_i}) \quad \rightsquigarrow
 \end{array}
 \left\{ \begin{array}{l}
 \bar{x} := \lambda[] \mathbf{begin} \quad \overline{x_{R_i}} += (\mathcal{D} u_i \overleftarrow{x_{R_i}}) \times \overline{x_{L_i}}; \\
 \quad \quad \quad \overline{x_{L_i}} := 0; \\
 \quad \quad \quad \bar{x} [] \mathbf{end} \\
 \\
 \overleftarrow{x_{L_i}} := u_i \overleftarrow{x_{R_i}} \\
 \\
 \bar{x} := \lambda[] \mathbf{begin} \quad \overline{x_{R_i}} += (\mathcal{D}_1 b_i(\overleftarrow{x_{R_i}}, \overleftarrow{x_{S_i}})) \times \overline{x_{L_i}}; \\
 \quad \quad \quad \overline{x_{S_i}} += (\mathcal{D}_2 b_i(\overleftarrow{x_{R_i}}, \overleftarrow{x_{S_i}})) \times \overline{x_{L_i}}; \\
 \quad \quad \quad \overline{x_{L_i}} := 0; \\
 \quad \quad \quad \bar{x} [] \mathbf{end} \\
 \\
 \overleftarrow{x_{L_i}} := b_i(\overleftarrow{x_{R_i}}, \overleftarrow{x_{S_i}})
 \end{array} \right.$$

$$\overleftarrow{x} \equiv x$$

A (Not So) Brief Tutorial on AD

$$\begin{array}{l}
 x_{L_i} := u_i x_{R_i} \\
 \\
 x_{L_i} := b_i(x_{R_i}, x_{S_i})
 \end{array}
 \rightsquigarrow
 \left\{ \begin{array}{l}
 \overleftarrow{x}_{L_i} := u_i \overleftarrow{x}_{R_i} \\
 \vdots \\
 \overleftarrow{x}_{R_i} + := (\mathcal{D} u_i \overleftarrow{x}_{R_i}) \times \overleftarrow{x}_{L_i}; \\
 \overleftarrow{x}_{L_i} := 0 \\
 \overleftarrow{x}_{L_i} := b_i(\overleftarrow{x}_{R_i}, \overleftarrow{x}_{S_i}) \\
 \vdots \\
 \overleftarrow{x}_{R_i} + := (\mathcal{D}_1 b_i(\overleftarrow{x}_{R_i}, \overleftarrow{x}_{S_i})) \times \overleftarrow{x}_{L_i}; \\
 \overleftarrow{x}_{S_i} + := (\mathcal{D}_2 b_i(\overleftarrow{x}_{R_i}, \overleftarrow{x}_{S_i})) \times \overleftarrow{x}_{L_i}; \\
 \overleftarrow{x}_{L_i} := 0
 \end{array} \right.$$

$$\overleftarrow{x} \equiv x$$

The Functional Reverse-Mode Transformation

$$x_{L_1} := u_1 \ x_{S_1}$$

$$\vdots$$

$$x_{L_n} := u_n \ x_{S_n}$$

The Functional Reverse-Mode Transformation

$$\left. \begin{array}{l} x_{L_1} := u_1 x_{S_1} \\ \vdots \\ x_{L_n} := u_n x_{S_n} \end{array} \right\} \rightsquigarrow \left\{ \begin{array}{l} x_{L_1} := u_1 x_{S_1} \\ \vdots \\ x_{L_n} := u_n x_{S_n} \\ \overline{x_1} := 0 \\ \vdots \\ \overline{x_m} := 0 \\ \overline{x_{S_n}} + := (\mathcal{D} u_n x_{S_n}) \times \overline{x_{L_n}} \\ \overline{x_{L_n}} := 0 \\ \vdots \\ \overline{x_{S_1}} + := (\mathcal{D} u_1 x_{S_1}) \times \overline{x_{L_1}} \\ \overline{x_{L_1}} := 0 \end{array} \right.$$

The Functional Reverse-Mode Transformation

$$\left. \begin{array}{l} x_{L_1} \quad := \quad u_1 \ x_{S_1} \\ \vdots \\ x_{L_n} \quad := \quad u_n \ x_{S_n} \end{array} \right\} \rightsquigarrow \left\{ \begin{array}{l} x_{L_1} \quad := \quad u_1 \ x_{S_1} \\ \vdots \\ x_{L_n} \quad := \quad u_n \ x_{S_n} \\ \overline{x_1} \quad := \quad 0 \\ \vdots \\ \overline{x_m} \quad := \quad 0 \\ \overline{x_{S_n}} \quad +:= \quad (\mathcal{D} \ u_n \ x_{S_n}) \times \overline{x_{L_n}} \\ \overline{x_{L_n}} \quad := \quad 0 \\ \vdots \\ \overline{x_{S_1}} \quad +:= \quad (\mathcal{D} \ u_1 \ x_{S_1}) \times \overline{x_{L_1}} \\ \overline{x_{L_1}} \quad := \quad = 0 \end{array} \right.$$

The Functional Reverse-Mode Transformation

$$\left. \begin{array}{l} x_{L_1} := u_1 x_{S_1} \\ \vdots \\ x_{L_n} := u_n x_{S_n} \end{array} \right\} \rightsquigarrow \left\{ \begin{array}{l} x_{L_1} := u_1 x_{S_1} \\ \vdots \\ x_{L_n} := u_n x_{S_n} \\ \overline{x_1} := 0 \\ \vdots \\ \overline{x_m} := 0 \\ \overline{x_{S_n}} +:= (\mathcal{D} u_n x_{S_n}) \times \overline{x_{L_n}} \\ \overline{x_{L_n}} := 0 \\ \vdots \\ \overline{x_{S_1}} +:= (\mathcal{D} u_1 x_{S_1}) \times \overline{x_{L_1}} \\ \overline{x_{L_1}} := 0 \end{array} \right.$$

The Functional Reverse-Mode Transformation

$$\left. \begin{array}{l} x_{L_1} := u_1 x_{S_1} \\ \vdots \\ x_{L_n} := u_n x_{S_n} \end{array} \right\} \rightsquigarrow \left\{ \begin{array}{l} x_{L_1} := u_1 x_{S_1} \\ \vdots \\ x_{L_n} := u_n x_{S_n} \\ \overline{x_1} := 0 \\ \vdots \\ \overline{x_m} := 0 \\ \overline{x_{S_n}} +: = (\mathcal{D} u_n x_{S_n}) \times \overline{x_{L_n}} \\ \overline{x_{L_n}} := 0 \\ \vdots \\ \overline{x_{S_1}} +: = (\mathcal{D} u_1 x_{S_1}) \times \overline{x_{L_1}} \\ \overline{x_{L_1}} := 0 \end{array} \right.$$

The Functional Reverse-Mode Transformation

$$\left. \begin{array}{l} x_1 = u_1 x_{S_1} \\ \vdots \\ x_n = u_n x_{S_n} \end{array} \right\} \rightsquigarrow \left\{ \begin{array}{l} x_1 = u_1 x_{S_1} \\ \vdots \\ x_n = u_n x_{S_n} \\ \overline{x_0} := 0 \\ \vdots \\ \overline{x_{n-1}} := 0 \\ \overline{x_{S_n}} +: = (\mathcal{D} u_n x_{S_n}) \times \overline{x_n} \\ \vdots \\ \overline{x_{S_1}} +: = (\mathcal{D} u_1 x_{S_1}) \times \overline{x_1} \end{array} \right.$$

The Functional Reverse-Mode Transformation

$$\left. \begin{array}{l} x_1 = u_1 x_{S_1} \\ \vdots \\ x_n = u_n x_{S_n} \end{array} \right\} \rightsquigarrow \left\{ \begin{array}{l} x_1 = u_1 x_{S_1} \\ \vdots \\ x_n = u_n x_{S_n} \\ \overline{x_0} := 0 \\ \vdots \\ \overline{x_{n-1}} := 0 \\ \overline{x_{S_n}} +:= (\mathcal{D} u_n x_{S_n}) \times \overline{x_n} \\ \vdots \\ \overline{x_{S_1}} +:= (\mathcal{D} u_1 x_{S_1}) \times \overline{x_1} \end{array} \right.$$

The Functional Reverse-Mode Transformation

$$\overline{u}_i \triangleq \lambda \overline{x} (\mathcal{D} u_i x_{S_i}) \times \overline{x}$$

$$\left. \begin{array}{l} x_1 = u_1 x_{S_1} \\ \vdots \\ x_n = u_n x_{S_n} \end{array} \right\} \rightsquigarrow \left\{ \begin{array}{l} x_1 = u_1 x_{S_1} \\ \vdots \\ x_n = u_n x_{S_n} \\ \overline{x_0} := 0 \\ \vdots \\ \overline{x_{n-1}} := 0 \\ \overline{x_{S_n}} + := \overline{u_n} \overline{x_n} \\ \vdots \\ \overline{x_{S_1}} + := \overline{u_1} \overline{x_1} \end{array} \right.$$

The Functional Reverse-Mode Transformation

$$\overleftarrow{u} \ x \triangleq ((u \ x), (\lambda \overline{x} (\mathcal{D} \ u \ x) \times \overline{x})))$$

$$\left. \begin{array}{l} x_1 = u_1 \ x_{S_1} \\ \vdots \\ x_n = u_n \ x_{S_n} \end{array} \right\} \rightsquigarrow \left\{ \begin{array}{l} (x_1, \overline{x}_1) = \overleftarrow{u}_1 \ x_{S_1} \\ \vdots \\ (x_n, \overline{x}_n) = \overleftarrow{u}_n \ x_{S_n} \\ \overline{x}_0 := 0 \\ \vdots \\ \overline{x}_{n-1} := 0 \\ \overline{x}_{S_n} + := \overline{x}_n \ \overline{x}_n \\ \vdots \\ \overline{x}_{S_1} + := \overline{x}_1 \ \overline{x}_1 \end{array} \right.$$

The Functional Reverse-Mode Transformation

$$\overleftarrow{u} \ x \triangleq ((u \ x), (\lambda \overleftarrow{x} (\mathcal{D} \ u \ x) \times \overleftarrow{x}))$$

$$\left. \begin{array}{l} x_1 = u_1 x_{S_1} \\ \vdots \\ x_n = u_n x_{S_n} \end{array} \right\} \rightsquigarrow \left\{ \begin{array}{l} (x_1, \overleftarrow{x}_1) = \overleftarrow{u}_1 x_{S_1} \\ \vdots \\ (x_n, \overleftarrow{x}_n) = \overleftarrow{u}_n x_{S_n} \\ \overleftarrow{x}_0 := 0 \\ \vdots \\ \overleftarrow{x}_{n-1} := 0 \\ \overleftarrow{x}_{S_n} + := \overleftarrow{x}_n \ \overleftarrow{x}_n \\ \vdots \\ \overleftarrow{x}_{S_1} + := \overleftarrow{x}_1 \ \overleftarrow{x}_1 \end{array} \right.$$

The Functional Reverse-Mode Transformation

$$\left. \begin{array}{l} x_1 = x_{R_1} x_{S_1} \\ \vdots \\ x_n = x_{R_n} x_{S_n} \end{array} \right\} \rightsquigarrow \left\{ \begin{array}{l} (\overleftarrow{x_1}, \overleftarrow{x_1}) = \overleftarrow{x_{R_1}} \overleftarrow{x_{S_1}} \\ \vdots \\ (\overleftarrow{x_n}, \overleftarrow{x_n}) = \overleftarrow{x_{R_n}} \overleftarrow{x_{S_n}} \\ \overleftarrow{x_0} := 0 \\ \vdots \\ \overleftarrow{x_{n-1}} := 0 \\ \overleftarrow{x_{S_n}} +: = \overleftarrow{x_n} \overleftarrow{x_n} \\ \vdots \\ \overleftarrow{x_{S_1}} +: = \overleftarrow{x_1} \overleftarrow{x_1} \end{array} \right.$$

The Functional Reverse-Mode Transformation

$$\left. \begin{array}{l} x_1 = x_{R_1} x_{S_1} \\ \vdots \\ x_n = x_{R_n} x_{S_n} \end{array} \right\} \rightsquigarrow \left\{ \begin{array}{l} (\overleftarrow{x_1}, \overline{x_1}) = \overleftarrow{x_{R_1}} \overleftarrow{x_{S_1}} \\ \vdots \\ (\overleftarrow{x_n}, \overline{x_n}) = \overleftarrow{x_{R_n}} \overleftarrow{x_{S_n}} \\ \overline{x_0} := 0 \\ \vdots \\ \overline{x_{n-1}} := 0 \\ \overline{x_{S_n}} +:= \overline{x_n} \overline{x_n} \\ \vdots \\ \overline{x_{S_1}} +:= \overline{x_1} \overline{x_1} \end{array} \right.$$

The Functional Reverse-Mode Transformation

$$\left. \begin{array}{l} x_1 = x_{R_1} x_{S_1} \\ \vdots \\ x_n = x_{R_n} x_{S_n} \end{array} \right\} \rightsquigarrow \left\{ \begin{array}{l} (\overleftarrow{x_1}, \overleftarrow{x_1}) = \overleftarrow{x_{R_1}} \overleftarrow{x_{S_1}} \\ \vdots \\ (\overleftarrow{x_n}, \overleftarrow{x_n}) = \overleftarrow{x_{R_n}} \overleftarrow{x_{S_n}} \\ \overleftarrow{x_0} := 0 \\ \vdots \\ \overleftarrow{x_{n-1}} := 0 \\ \overleftarrow{x_{S_n}} +:= \overleftarrow{x_n} \overleftarrow{x_n} \\ \vdots \\ \overleftarrow{x_{S_1}} +:= \overleftarrow{x_1} \overleftarrow{x_1} \end{array} \right.$$

The Functional Reverse-Mode Transformation

$$\left. \begin{array}{l} x_1 = x_{R_1} x_{S_1} \\ \vdots \\ x_n = x_{R_n} x_{S_n} \end{array} \right\} \rightsquigarrow \left\{ \begin{array}{l} (\overleftarrow{x_1}, \overleftarrow{x_1}) = \overleftarrow{x_{R_1}} \overleftarrow{x_{S_1}} \\ \vdots \\ (\overleftarrow{x_n}, \overleftarrow{x_n}) = \overleftarrow{x_{R_n}} \overleftarrow{x_{S_n}} \\ \overleftarrow{x_0} := 0 \\ \vdots \\ \overleftarrow{x_{n-1}} := 0 \\ \overleftarrow{x_{S_n}} \text{ } + := \overleftarrow{x_n} \overleftarrow{x_n} \\ \vdots \\ \overleftarrow{x_{S_1}} \text{ } + := \overleftarrow{x_1} \overleftarrow{x_1} \end{array} \right.$$

The Functional Reverse-Mode Transformation

$$\left. \begin{array}{l} x_1 = x_{R_1} x_{S_1} \\ \vdots \\ x_n = x_{R_n} x_{S_n} \end{array} \right\} \rightsquigarrow \left\{ \begin{array}{l} (\overleftarrow{x}_1, \overleftarrow{x}_1) = \overleftarrow{x}_{R_1} \overleftarrow{x}_{S_1} \\ \vdots \\ (\overleftarrow{x}_n, \overleftarrow{x}_n) = \overleftarrow{x}_{R_n} \overleftarrow{x}_{S_n} \\ \overleftarrow{x}_0 := \mathbf{0} (\overleftarrow{\mathcal{J}}^{-1} \overleftarrow{x}_0) \\ \vdots \\ \overleftarrow{x}_{n-1} := \mathbf{0} (\overleftarrow{\mathcal{J}}^{-1} \overleftarrow{x}_{n-1}) \\ \overleftarrow{x}_{S_n} \oplus := \overleftarrow{x}_n \overleftarrow{x}_n \\ \vdots \\ \overleftarrow{x}_{S_1} \oplus := \overleftarrow{x}_1 \overleftarrow{x}_1 \end{array} \right.$$

Modularity

 $\nabla f \mathbf{x}$

$$\triangleq \frac{\partial f(\mathbf{x})}{\partial x_1}, \dots, \frac{\partial f(\mathbf{x})}{\partial x_n}$$

Modularity

$$\nabla f \mathbf{x} \triangleq \frac{\partial f(\mathbf{x})}{\partial x_1}, \dots, \frac{\partial f(\mathbf{x})}{\partial x_n}$$

$$\text{GRADIENTDESCENT } f \mathbf{x}_0 \triangleq \dots \mathbf{x}_{i+1} := \dots \nabla f \mathbf{x}_i \dots$$

Modularity

| | | |
|----------------------------------|--------------|---|
| $\nabla f \mathbf{x}$ | \triangleq | $\frac{\partial f(\mathbf{x})}{\partial x_1}, \dots, \frac{\partial f(\mathbf{x})}{\partial x_n}$ |
| GRADIENTDESCENT $f \mathbf{x}_0$ | \triangleq | $\dots \mathbf{x}_{i+1} := \dots \nabla f \mathbf{x}_i \dots$ |
| argmin f | \triangleq | $\dots \text{GRADIENTDESCENT } f \mathbf{x}_0 \dots$ |

Modularity

$$\nabla f \mathbf{x} \triangleq \frac{\partial f(\mathbf{x})}{\partial x_1}, \dots, \frac{\partial f(\mathbf{x})}{\partial x_n}$$

$$\text{GRADIENTDESCENT } f \mathbf{x}_0 \triangleq \dots \mathbf{x}_{i+1} := \dots \nabla f \mathbf{x}_i \dots$$

$$\text{argmin } f \triangleq \dots \text{GRADIENTDESCENT } f \mathbf{x}_0 \dots$$

$$\text{NEUTRONFLUX } r \triangleq \boxed{\textit{classified}}$$

Modularity

| | | |
|----------------------------------|--------------|---|
| $\nabla f \mathbf{x}$ | \triangleq | $\frac{\partial f(\mathbf{x})}{\partial x_1}, \dots, \frac{\partial f(\mathbf{x})}{\partial x_n}$ |
| GRADIENTDESCENT $f \mathbf{x}_0$ | \triangleq | $\dots \mathbf{x}_{i+1} := \dots \nabla f \mathbf{x}_i \dots$ |
| argmin f | \triangleq | $\dots \text{GRADIENTDESCENT } f \mathbf{x}_0 \dots$ |
| NEUTRONFLUX r | \triangleq | <i>classified</i> |
| DEVIATION r | \triangleq | $((\text{NEUTRONFLUX } r) - \text{NEUTRONFLUX}_{\text{critical}})^2$ |

Modularity

| | | |
|----------------------------------|--------------|---|
| $\nabla f \mathbf{x}$ | \triangleq | $\frac{\partial f(\mathbf{x})}{\partial x_1}, \dots, \frac{\partial f(\mathbf{x})}{\partial x_n}$ |
| GRADIENTDESCENT $f \mathbf{x}_0$ | \triangleq | $\dots \mathbf{x}_{i+1} := \dots \nabla f \mathbf{x}_i \dots$ |
| argmin f | \triangleq | $\dots \text{GRADIENTDESCENT } f \mathbf{x}_0 \dots$ |
| NEUTRONFLUX r | \triangleq | classified |
| DEVIATION r | \triangleq | $((\text{NEUTRONFLUX } r) - \text{NEUTRONFLUX}_{\text{critical}})^2$ |
| r^* | \triangleq | argmin DEVIATION |

Fermi, E. (1946). *The Development of the first chain reaction pile*.
Proceedings of the American Philosophy Society, **90**:20–4.

Breaking Modularity

| | | |
|----------------------------------|--------------|---|
| $\nabla f \mathbf{x}$ | \triangleq | $(\vec{f} \mathbf{x} \triangleright \vec{e}_1), \dots, (\vec{f} \mathbf{x} \triangleright \vec{e}_n)$ |
| GRADIENTDESCENT $f \mathbf{x}_0$ | \triangleq | $\dots \mathbf{x}_{i+1} := \dots \nabla f \mathbf{x}_i \dots$ |
| argmin f | \triangleq | $\dots \text{GRADIENTDESCENT } f \mathbf{x}_0 \dots$ |
| NEUTRONFLUX r | \triangleq | classified |
| DEVIATION r | \triangleq | $((\text{NEUTRONFLUX } r) - \text{NEUTRONFLUX}_{\text{critical}})^2$ |
| r^* | \triangleq | argmin DEVIATION |

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Breaking Modularity

$$\nabla \vec{f} \mathbf{x} \triangleq (\vec{f} \mathbf{x} \triangleright \vec{e}_1), \dots, (\vec{f} \mathbf{x} \triangleright \vec{e}_n)$$

$$\text{GRADIENTDESCENT } f \mathbf{x}_0 \triangleq \dots \mathbf{x}_{i+1} := \dots \nabla f \mathbf{x}_i \dots$$

$$\text{argmin } f \triangleq \dots \text{GRADIENTDESCENT } f \mathbf{x}_0 \dots$$

$$\text{NEUTRONFLUX } r \triangleq \boxed{\textit{classified}}$$

$$\text{DEVIATION } r \triangleq ((\text{NEUTRONFLUX } r) - \text{NEUTRONFLUX}_{\text{critical}})^2$$

$$r^* \triangleq \text{argmin DEVIATION}$$

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Breaking Modularity

$$\nabla \vec{f} \mathbf{x} \triangleq (\vec{f} \mathbf{x} \triangleright \vec{e}_1), \dots, (\vec{f} \mathbf{x} \triangleright \vec{e}_n)$$

$$\text{GRADIENTDESCENT } f \mathbf{x}_0 \triangleq \dots \mathbf{x}_{i+1} := \dots \nabla \vec{f} \mathbf{x}_i \dots$$

$$\text{argmin } f \triangleq \dots \text{GRADIENTDESCENT } f \mathbf{x}_0 \dots$$

$$\text{NEUTRONFLUX } r \triangleq \boxed{\text{classified}}$$

$$\text{DEVIATION } r \triangleq ((\text{NEUTRONFLUX } r) - \text{NEUTRONFLUX}_{\text{critical}})^2$$

$$r^* \triangleq \text{argmin DEVIATION}$$

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Breaking Modularity

$$\nabla \vec{f} \mathbf{x} \triangleq (\vec{f} \mathbf{x} \triangleright \vec{e}_1'), \dots, (\vec{f} \mathbf{x} \triangleright \vec{e}_n')$$

$$\text{GRADIENTDESCENT } \vec{f} \mathbf{x}_0 \triangleq \dots \mathbf{x}_{i+1} := \dots \nabla \vec{f} \mathbf{x}_i \dots$$

$$\text{argmin } f \triangleq \dots \text{GRADIENTDESCENT } f \mathbf{x}_0 \dots$$

$$\text{NEUTRONFLUX } r \triangleq \boxed{\text{classified}}$$

$$\text{DEVIATION } r \triangleq ((\text{NEUTRONFLUX } r) - \text{NEUTRONFLUX}_{\text{critical}})^2$$

$$r^* \triangleq \text{argmin DEVIATION}$$

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Breaking Modularity

$$\nabla \vec{f} \mathbf{x} \triangleq (\vec{f} \mathbf{x} \triangleright \vec{e}_1^T), \dots, (\vec{f} \mathbf{x} \triangleright \vec{e}_n^T)$$

$$\text{GRADIENTDESCENT } \vec{f} \mathbf{x}_0 \triangleq \dots \mathbf{x}_{i+1} := \dots \nabla \vec{f} \mathbf{x}_i \dots$$

$$\text{argmin } f \triangleq \dots \text{GRADIENTDESCENT } \vec{f} \mathbf{x}_0 \dots$$

$$\text{NEUTRONFLUX } r \triangleq \boxed{\text{classified}}$$

$$\text{DEVIATION } r \triangleq ((\text{NEUTRONFLUX } r) - \text{NEUTRONFLUX}_{\text{critical}})^2$$

$$r^* \triangleq \text{argmin DEVIATION}$$

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Breaking Modularity

$$\nabla \vec{f} \mathbf{x} \triangleq (\vec{f} \mathbf{x} \triangleright \vec{e}_1), \dots, (\vec{f} \mathbf{x} \triangleright \vec{e}_n)$$

$$\text{GRADIENTDESCENT } \vec{f} \mathbf{x}_0 \triangleq \dots \mathbf{x}_{i+1} := \dots \nabla \vec{f} \mathbf{x}_i \dots$$

$$\text{argmin } \vec{f} \triangleq \dots \text{GRADIENTDESCENT } \vec{f} \mathbf{x}_0 \dots$$

$$\text{NEUTRONFLUX } r \triangleq \boxed{\text{classified}}$$

$$\text{DEVIATION } r \triangleq ((\text{NEUTRONFLUX } r) - \text{NEUTRONFLUX}_{\text{critical}})^2$$

$$r^* \triangleq \text{argmin DEVIATION}$$

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Breaking Modularity

$$\nabla \vec{f} \mathbf{x} \triangleq (\vec{f} \mathbf{x} \triangleright \vec{e}_1), \dots, (\vec{f} \mathbf{x} \triangleright \vec{e}_n)$$

$$\text{GRADIENTDESCENT } \vec{f} \mathbf{x}_0 \triangleq \dots \mathbf{x}_{i+1} := \dots \nabla \vec{f} \mathbf{x}_i \dots$$

$$\text{argmin } \vec{f} \triangleq \dots \text{GRADIENTDESCENT } \vec{f} \mathbf{x}_0 \dots$$

$$\text{NEUTRONFLUX } r \triangleq \boxed{\text{classified}}$$

$$\text{DEVIATION } r \triangleq ((\text{NEUTRONFLUX } r) - \text{NEUTRONFLUX}_{\text{critical}})^2$$

$$r^* \triangleq \text{argmin } \overrightarrow{\text{DEVIATION}}$$

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Breaking Modularity

$$\nabla \vec{f} \mathbf{x} \triangleq (\vec{f} \mathbf{x} \triangleright \vec{e}_1), \dots, (\vec{f} \mathbf{x} \triangleright \vec{e}_n)$$

$$\text{GRADIENTDESCENT } \vec{f} \mathbf{x}_0 \triangleq \dots \mathbf{x}_{i+1} := \dots \nabla \vec{f} \mathbf{x}_i \dots$$

$$\text{argmin } \vec{f} \triangleq \dots \text{GRADIENTDESCENT } \vec{f} \mathbf{x}_0 \dots$$

$$\text{NEUTRONFLUX } r \triangleq \boxed{\text{classified}}$$

$$\text{DEVIATION } r \triangleq ((\text{NEUTRONFLUX } r) - \text{NEUTRONFLUX}_{\text{critical}})^2$$

$$\text{DEVIATION} \xrightarrow{\text{ADIFOR}} \overrightarrow{\text{DEVIATION}}$$

$$r^* \triangleq \text{argmin } \overrightarrow{\text{DEVIATION}}$$

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Breaking Modularity

$$\nabla \vec{f} \mathbf{x} \triangleq (\vec{f} \mathbf{x} \triangleright \vec{e}_1^T), \dots, (\vec{f} \mathbf{x} \triangleright \vec{e}_n^T)$$

$$\text{GRADIENTDESCENT } \vec{f} \mathbf{x}_0 \triangleq \dots \mathbf{x}_{i+1} := \dots \nabla \vec{f} \mathbf{x}_i \dots$$

$$\text{argmin } \vec{f} \triangleq \dots \text{GRADIENTDESCENT } \vec{f} \mathbf{x}_0 \dots$$

$$\text{NEUTRONFLUX } r \triangleq \boxed{\text{classified}}$$

$$\text{NEUTRONFLUX} \xrightarrow[\rightsquigarrow]{\text{ADIFOR}} \overline{\text{NEUTRONFLUX}}$$

$$\text{DEVIATION } r \triangleq ((\text{NEUTRONFLUX } r) - \text{NEUTRONFLUX}_{\text{critical}})^2$$

$$\text{DEVIATION} \xrightarrow[\rightsquigarrow]{\text{ADIFOR}} \overline{\text{DEVIATION}}$$

$$r^* \triangleq \text{argmin } \overline{\text{DEVIATION}}$$

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| GRADIENTDESCENT $\vec{f} \mathbf{x}_0$ | \triangleq | $\dots \mathbf{x}_{i+1} := \dots \nabla \vec{f} \mathbf{x}_i \dots$ |
| argmin \vec{f} | \triangleq | $\dots \text{GRADIENTDESCENT } \vec{f} \mathbf{x}_0 \dots$ |
| NEUTRONFLUX \mathbf{r} | \triangleq | classified |
| NEUTRONFLUX | $\xrightarrow[\rightsquigarrow]{\text{ADIFOR}}$ | $\overrightarrow{\text{NEUTRONFLUX}}$ |
| DEVIATION \mathbf{r} | \triangleq | $((\text{NEUTRONFLUX } \mathbf{r}) - \text{NEUTRONFLUX}_{\text{critical}})^2$ |
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| argmin \overleftarrow{f} | \triangleq | $\dots \text{GRADIENTDESCENT } \overleftarrow{f} \mathbf{x}_0 \dots$ |
| NEUTRONFLUX \mathbf{r} | \triangleq | classified |
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| DEVIATION | $\overset{\text{TAPENADE}}{\rightsquigarrow}$ | $\overleftarrow{\text{DEVIATION}}$ |
| \mathbf{r}^* | \triangleq | argmin $\overleftarrow{\text{DEVIATION}}$ |

Fermi, E. (1946). *The Development of the first chain reaction pile*.
 Proceedings of the American Philosophy Society, **90**:20–4.

Breaking Modularity

| | | |
|---|---|---|
| $\nabla \overleftarrow{f} \mathbf{x}$ | \triangleq | $\dots \overleftarrow{f} \mathbf{x} \dots$ |
| $\mathcal{H} \overrightarrow{f} \mathbf{x}$ | \triangleq | $\dots \overrightarrow{f} \dots \mathbf{x} \dots$ |
| GRADIENTDESCENT $\overleftarrow{f} \mathbf{x}_0$ | \triangleq | $\dots \mathbf{x}_{i+1} := \dots \nabla \overleftarrow{f} \mathbf{x}_i \dots$ |
| NEWTONSMETHOD $\overleftarrow{f} \overrightarrow{f} \mathbf{x}_0$ | \triangleq | $\dots \mathbf{x}_{i+1} := \dots \nabla \overleftarrow{f} \mathbf{x}_i \dots \mathcal{H} \overrightarrow{f} \mathbf{x}_i \dots$ |
| argmin $\overleftarrow{f} \overrightarrow{f}$ | \triangleq | $\dots \text{NEWTONSMETHOD} \overleftarrow{f} \overrightarrow{f} \mathbf{x}_0 \dots$ |
| NEUTRONFLUX \mathbf{r} | \triangleq | classified |
| NEUTRONFLUX | $\overset{\text{TAPENADE}}{\rightsquigarrow}$ | $\overleftarrow{\text{NEUTRONFLUX}}$ |
| DEVIATION \mathbf{r} | \triangleq | $((\text{NEUTRONFLUX } \mathbf{r}) - \text{NEUTRONFLUX}_{\text{critical}})^2$ |
| DEVIATION | $\overset{\text{TAPENADE}}{\rightsquigarrow}$ | $\overleftarrow{\text{DEVIATION}}$ |
| \mathbf{r}^* | \triangleq | argmin $\overleftarrow{\text{DEVIATION}} \overrightarrow{\text{DEVIATION}}$ |

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Breaking Modularity

| | | |
|---|-----------------------------|---|
| $\nabla \overleftarrow{f} \mathbf{x}$ | \triangleq | $\dots \overleftarrow{f} \mathbf{x} \dots$ |
| $\mathcal{H} \overrightarrow{f} \mathbf{x}$ | \triangleq | $\dots \overrightarrow{f} \dots \mathbf{x} \dots$ |
| GRADIENTDESCENT $\overleftarrow{f} \mathbf{x}_0$ | \triangleq | $\dots \mathbf{x}_{i+1} := \dots \nabla \overleftarrow{f} \mathbf{x}_i \dots$ |
| NEWTONSMETHOD $\overleftarrow{f} \overrightarrow{f} \mathbf{x}_0$ | \triangleq | $\dots \mathbf{x}_{i+1} := \dots \nabla \overleftarrow{f} \mathbf{x}_i \dots \mathcal{H} \overrightarrow{f} \mathbf{x}_i \dots$ |
| argmin $\overleftarrow{f} \overrightarrow{f}$ | \triangleq | $\dots \text{NEWTONSMETHOD} \overleftarrow{f} \overrightarrow{f} \mathbf{x}_0 \dots$ |
| NEUTRONFLUX \mathbf{r} | \triangleq | classified |
| NEUTRONFLUX | TAPENADE \rightsquigarrow | $\overleftarrow{\text{NEUTRONFLUX}}$ |
| $\overleftarrow{\text{NEUTRONFLUX}}$ | TAPENADE \rightsquigarrow | $\overrightarrow{\overleftarrow{\text{NEUTRONFLUX}}}$ |
| DEVIATION \mathbf{r} | \triangleq | $((\text{NEUTRONFLUX } \mathbf{r}) - \text{NEUTRONFLUX}_{\text{critical}})^2$ |
| DEVIATION | TAPENADE \rightsquigarrow | $\overleftarrow{\text{DEVIATION}}$ |
| $\overleftarrow{\text{DEVIATION}}$ | TAPENADE \rightsquigarrow | $\overrightarrow{\overleftarrow{\text{DEVIATION}}}$ |
| \mathbf{r}^* | \triangleq | argmin $\overleftarrow{\text{DEVIATION}} \overrightarrow{\overleftarrow{\text{DEVIATION}}}$ |

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Restoring Modularity

| | | |
|----------------------------------|--------------|--|
| $\nabla f \mathbf{x}$ | \triangleq | |
| $\mathcal{H} f \mathbf{x}$ | \triangleq | |
| GRADIENTDESCENT $f \mathbf{x}_0$ | \triangleq | $\dots \mathbf{x}_{i+1} := \dots \nabla f \mathbf{x}_i \dots$ |
| NEWTONSMETHOD $f \mathbf{x}_0$ | \triangleq | $\dots \mathbf{x}_{i+1} := \dots \nabla f \mathbf{x}_i \dots \mathcal{H} f \mathbf{x}_i \dots$ |
| argmin f | \triangleq | $\dots \text{GRADIENTDESCENT } f \mathbf{x}_0 \dots$ |
| NEUTRONFLUX \mathbf{r} | \triangleq | <i>classified</i> |
| DEVIATION \mathbf{r} | \triangleq | $((\text{NEUTRONFLUX } \mathbf{r}) - \text{NEUTRONFLUX}_{\text{critical}})^2$ |
| \mathbf{r}^* | \triangleq | argmin DEVIATION |

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Restoring Modularity

| | | |
|----------------------------------|--------------|---|
| $\nabla f \mathbf{x}$ | \triangleq | $((\vec{\mathcal{J}} f) \mathbf{x} \triangleright \vec{\mathbf{e}}_1'), \dots, ((\vec{\mathcal{J}} f) \mathbf{x} \triangleright \vec{\mathbf{e}}_n')$ |
| $\mathcal{H} f \mathbf{x}$ | \triangleq | |
| GRADIENTDESCENT $f \mathbf{x}_0$ | \triangleq | $\dots \mathbf{x}_{i+1} := \dots \nabla f \mathbf{x}_i \dots$ |
| NEWTONSMETHOD $f \mathbf{x}_0$ | \triangleq | $\dots \mathbf{x}_{i+1} := \dots \nabla f \mathbf{x}_i \dots \mathcal{H} f \mathbf{x}_i \dots$ |
| argmin f | \triangleq | $\dots \text{GRADIENTDESCENT } f \mathbf{x}_0 \dots$ |
| NEUTRONFLUX \mathbf{r} | \triangleq | <i>classified</i> |
| DEVIATION \mathbf{r} | \triangleq | $((\text{NEUTRONFLUX } \mathbf{r}) - \text{NEUTRONFLUX}_{\text{critical}})^2$ |
| \mathbf{r}^* | \triangleq | argmin DEVIATION |

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Restoring Modularity

| | | |
|----------------------------------|--------------|--|
| $\nabla f \mathbf{x}$ | \triangleq | $\dots (\overleftarrow{\mathcal{J}} f) \mathbf{x} \dots$ |
| $\mathcal{H} f \mathbf{x}$ | \triangleq | |
| GRADIENTDESCENT $f \mathbf{x}_0$ | \triangleq | $\dots \mathbf{x}_{i+1} := \dots \nabla f \mathbf{x}_i \dots$ |
| NEWTONSMETHOD $f \mathbf{x}_0$ | \triangleq | $\dots \mathbf{x}_{i+1} := \dots \nabla f \mathbf{x}_i \dots \mathcal{H} f \mathbf{x}_i \dots$ |
| argmin f | \triangleq | $\dots \text{GRADIENTDESCENT } f \mathbf{x}_0 \dots$ |
| NEUTRONFLUX \mathbf{r} | \triangleq | classified |
| DEVIATION \mathbf{r} | \triangleq | $((\text{NEUTRONFLUX } \mathbf{r}) - \text{NEUTRONFLUX}_{\text{critical}})^2$ |
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Restoring Modularity

| | | |
|----------------------------------|--------------|--|
| $\nabla f \mathbf{x}$ | \triangleq | $\dots (\overleftarrow{\mathcal{J}} f) \mathbf{x} \dots$ |
| $\mathcal{H} f \mathbf{x}$ | \triangleq | $\dots (\overrightarrow{\mathcal{J}} (\overleftarrow{\mathcal{J}} f)) \dots \mathbf{x} \dots$ |
| GRADIENTDESCENT $f \mathbf{x}_0$ | \triangleq | $\dots \mathbf{x}_{i+1} := \dots \nabla f \mathbf{x}_i \dots$ |
| NEWTONSMETHOD $f \mathbf{x}_0$ | \triangleq | $\dots \mathbf{x}_{i+1} := \dots \nabla f \mathbf{x}_i \dots \mathcal{H} f \mathbf{x}_i \dots$ |
| argmin f | \triangleq | $\dots \text{NEWTONSMETHOD } f \mathbf{x}_0 \dots$ |
| NEUTRONFLUX \mathbf{r} | \triangleq | classified |
| DEVIATION \mathbf{r} | \triangleq | $((\text{NEUTRONFLUX } \mathbf{r}) - \text{NEUTRONFLUX}_{\text{critical}})^2$ |
| \mathbf{r}^* | \triangleq | argmin DEVIATION |

Fermi, E. (1946). *The Development of the first chain reaction pile*.
Proceedings of the American Philosophy Society, **90**:20–4.

Having your cake and eating it too

- Convenient

- Fast

Having your cake and eating it too

- Convenient
 - \mathcal{D} formulated as a higher-order function in the language
 - no arbitrary restrictions
 - applies to all data types and constructs in the language, including code produced by \mathcal{D} and even \mathcal{D} itself

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- \mathcal{D} formulated as a higher-order function in the language
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 - applies to all data types and constructs in the language, including code produced by \mathcal{D} and even \mathcal{D} itself
- higher-order derivatives
 - $(\mathcal{D} (\mathcal{D} f))$

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- higher-order derivatives
 - $(\mathcal{D} (\mathcal{D} f))$
- nesting
 - $(\mathcal{D} (\text{lambda } (...) \dots (\mathcal{D} (\text{lambda } (...) \dots)) \dots))$

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 - $(\mathcal{D} (\text{lambda } (...) \dots (\mathcal{D} (\text{lambda } (...) \dots)) \dots))$

- Fast

- \mathcal{D} implemented by reflective transformation of environments and code associated with closures

Having your cake and eating it too

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- higher-order derivatives
 - $(\mathcal{D} (\mathcal{D} f))$
- nesting
 - $(\mathcal{D} (\text{lambda } (...) \dots (\mathcal{D} (\text{lambda } (...) \dots)) \dots))$

- Fast

- \mathcal{D} implemented by reflective transformation of environments and code associated with closures
- compile away reflection with partial evaluation implemented by flow analysis

Monovariant Flow Analysis: 0-CFA

```
(define ( $\mathcal{D}$  f)  
...)
```

Monovariant Flow Analysis: 0-CFA

```
(define ( $\mathcal{D}$  f)  
  ...)
```

```
( $\mathcal{D}$  (lambda (x)  $2x^3$ ))
```

Monovariant Flow Analysis: 0-CFA

```
(define (D f:  $(\lambda x 2x^3)$ )  
  ...)
```

```
(D (lambda (x)  $2x^3$ ))
```

Monovariant Flow Analysis: 0-CFA

```
(define (D f:( $\lambda x 2x^3$ ))  
  ...:( $\lambda x 6x^2$ ))
```

```
(D (lambda (x)  $2x^3$ ))
```

Monovariant Flow Analysis: 0-CFA

```
(define (D f:( $\lambda x 2x^3$ ))  
  ...:( $\lambda x 6x^2$ ))
```

```
(D (lambda (x)  $2x^3$ )):( $\lambda x 6x^2$ )
```

Monovariant Flow Analysis: 0-CFA

```
(define (D f:( $\lambda x 2x^3$ ))  
  ...:( $\lambda x 6x^2$ ))
```

```
(D (lambda (x)  $2x^3$ )):( $\lambda x 6x^2$ )
```

```
(D (lambda (x)  $3x^4$ ))
```

Monovariant Flow Analysis: 0-CFA

```
(define (D f : (λx 2x3) ∪ (λx 3x4))  
  ... : (λx 6x2))
```

```
(D (lambda (x) 2x3)) : (λx 6x2)
```

```
(D (lambda (x) 3x4))
```

Monovariant Flow Analysis: 0-CFA

```
(define (D f : (λx 2x3) ∪ (λx 3x4))  
  ... : (λx 6x2) ∪ (λx 12x3))
```

```
(D (lambda (x) 2x3)) : (λx 6x2)
```

```
(D (lambda (x) 3x4))
```

Monovariant Flow Analysis: 0-CFA

```
(define (D f : ( $\lambda x$   $2x^3$ )  $\cup$  ( $\lambda x$   $3x^4$ ))  
  ... : ( $\lambda x$   $6x^2$ )  $\cup$  ( $\lambda x$   $12x^3$ ))
```

```
(D (lambda (x)  $2x^3$ )) : ( $\lambda x$   $6x^2$ )
```

```
(D (lambda (x)  $3x^4$ )) : ( $\lambda x$   $12x^3$ )
```

Monovariant Flow Analysis: 0-CFA

```
(define (D f : ( $\lambda x$   $2x^3$ )  $\cup$  ( $\lambda x$   $3x^4$ ))  
  ... : ( $\lambda x$   $6x^2$ )  $\cup$  ( $\lambda x$   $12x^3$ ))
```

```
(D (lambda (x)  $2x^3$ )) : ( $\lambda x$   $6x^2$ )  $\cup$  ( $\lambda x$   $12x^3$ )
```

```
(D (lambda (x)  $3x^4$ )) : ( $\lambda x$   $6x^2$ )  $\cup$  ( $\lambda x$   $12x^3$ )
```

Monovariant Flow Analysis: 0-CFA

```
(define (D f: ( $\lambda x$   $2x^3$ )  $\cup$  ( $\lambda x$   $3x^4$ ))  
  ...: ( $\lambda x$   $6x^2$ )  $\cup$  ( $\lambda x$   $12x^3$ ))
```

```
(D (lambda (x)  $2x^3$ )) : ( $\lambda x$   $6x^2$ )  $\cup$  ( $\lambda x$   $12x^3$ )
```

```
(D (lambda (x)  $3x^4$ )) : ( $\lambda x$   $6x^2$ )  $\cup$  ( $\lambda x$   $12x^3$ )
```

Monovariant Flow Analysis: 0-CFA

```
(define ( $\mathcal{D}$  f)  
  ...)
```

Monovariant Flow Analysis: 0-CFA

```
(define (D f)  
  ...)
```

```
(D (D (lambda (x) e2x)))
```

Monovariant Flow Analysis: 0-CFA

```
(define (D f:( $\lambda x e^{2x}$ ))  
  ...)
```

```
(D (D (lambda (x)  $e^{2x}$ )))
```

Monovariant Flow Analysis: 0-CFA

```
(define (D f:( $\lambda x e^{2x}$ )
...:( $\lambda x 2e^{2x}$ ))

(D (D (lambda (x)  $e^{2x}$ )))
```

Monovariant Flow Analysis: 0-CFA

```
(define (D f:( $\lambda x e^{2x}$ ))  
  ...:( $\lambda x 2e^{2x}$ ))
```

```
(D (D (lambda (x)  $e^{2x}$ ))):( $\lambda x 2e^{2x}$ ))
```

Monovariant Flow Analysis: 0-CFA

```
(define (D f:( $\lambda x e^{2x}$ )  $\cup$  ( $\lambda x 2e^{2x}$ ))  
  ...:( $\lambda x 2e^{2x}$ ))
```

```
(D (D (lambda (x)  $e^{2x}$ )):( $\lambda x 2e^{2x}$ ))
```

Monovariant Flow Analysis: 0-CFA

```
(define (D f:( $\lambda x e^{2x}$ )  $\cup$  ( $\lambda x 2e^{2x}$ ))  
  ...:( $\lambda x 2e^{2x}$ )  $\cup$  ( $\lambda x 4e^{2x}$ ))
```

```
(D (D (lambda (x)  $e^{2x}$ )):( $\lambda x 2e^{2x}$ ))
```

Monovariant Flow Analysis: 0-CFA

```
(define (D f:( $\lambda x e^{2x}$ )  $\cup$  ( $\lambda x 2e^{2x}$ ))  
...:( $\lambda x 2e^{2x}$ )  $\cup$  ( $\lambda x 4e^{2x}$ ))
```

```
(D (D (lambda (x)  $e^{2x}$ ))):(  $\lambda x 2e^{2x}$ )  $\cup$  ( $\lambda x 4e^{2x}$ ))
```

Monovariant Flow Analysis: 0-CFA

```
(define (D f:(λx e2x) ∪ (λx 2e2x) ∪ (λx 4e2x))  
  ...:(λx 2e2x) ∪ (λx 4e2x))
```

```
(D (D (lambda (x) e2x))):(λx 2e2x) ∪ (λx 4e2x))
```

Monovariant Flow Analysis: 0-CFA

```
(define (D f:( $\lambda x e^{2x}$ )  $\cup$  ( $\lambda x 2e^{2x}$ )  $\cup$  ( $\lambda x 4e^{2x}$ )  $\cup$  ...)
...:( $\lambda x 2e^{2x}$ )  $\cup$  ( $\lambda x 4e^{2x}$ )  $\cup$  ...)
```

```
(D (D (lambda (x)  $e^{2x}$ ))):(  $\lambda x 2e^{2x}$ )  $\cup$  ( $\lambda x 4e^{2x}$ )  $\cup$  ...)
```

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

```
(define ( $\mathcal{D}$  f) ...)
```

Shivers, III, O. G. (1991). *Control-Flow Analysis of Higher-Order Languages or Taming Lambda*. Ph.D. thesis, CMU.

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

```
(define ( $\mathcal{D}$  f) ...)
```

```
(define (g ...) ... ( $\mathcal{D}$  (lambda (x)  $2x^3$ )) ...)
```

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Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

```
(define ( $\mathcal{D}_g$  f) ...)
```

```
(define (g ...) ... ( $\mathcal{D}$  (lambda (x)  $2x^3$ )) ...)
```

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Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

```
(define ( $\mathcal{D}_g$  f:( $\lambda x$   $2x^3$ )) ...)
```

```
(define (g ...) ... ( $\mathcal{D}$  (lambda (x)  $2x^3$ )) ...)
```

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Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

```
(define (Dg f:(λx 2x3) ...:(λx 6x2))
```

```
(define (g ...) ... (D (lambda (x) 2x3)) ...)
```

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Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

```
(define (Dg f:(λx 2x3) ...:(λx 6x2))
```

```
(define (g ...) ... (D (lambda (x) 2x3)):(λx 6x2) ...)
```

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Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

```
(define (Dg f:(λx 2x3) ...:(λx 6x2))
```

```
(define (g ...) ... (D (lambda (x) 2x3)):(λx 6x2) ...)
```

```
(define (h ...) ... (D (lambda (x) 3x4)) ...)
```

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Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

```
(define (Dg f : (λx 2x3)) ... : (λx 6x2))
```

```
(define (Dh f) ...)
```

```
(define (g ...) ... (D (lambda (x) 2x3)) : (λx 6x2) ...)
```

```
(define (h ...) ... (D (lambda (x) 3x4)) ...)
```

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Polyvariant Flow Analysis: k -CFA

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```
(define (Dg f:(λx 2x3) ...:(λx 6x2))
```

```
(define (Dh f:(λx 3x4) ...)
```

```
(define (g ...) ... (D (lambda (x) 2x3)):(λx 6x2) ...)
```

```
(define (h ...) ... (D (lambda (x) 3x4)) ...)
```

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Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

```
(define (Dg f:(λx 2x3) ...:(λx 6x2))
```

```
(define (Dh f:(λx 3x4) ...:(λx 12x3))
```

```
(define (g ...) ... (D (lambda (x) 2x3)):(λx 6x2) ...)
```

```
(define (h ...) ... (D (lambda (x) 3x4)) ...)
```

Shivers, III, O. G. (1991). *Control-Flow Analysis of Higher-Order Languages or Taming Lambda*. Ph.D. thesis, CMU.

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

```
(define ( $\mathcal{D}_g$  f : ( $\lambda x$   $2x^3$ )) ... : ( $\lambda x$   $6x^2$ ))
```

```
(define ( $\mathcal{D}_h$  f : ( $\lambda x$   $3x^4$ )) ... : ( $\lambda x$   $12x^3$ ))
```

```
(define (g ...) ... ( $\mathcal{D}$  (lambda (x)  $2x^3$ )) : ( $\lambda x$   $6x^2$ ) ...)
```

```
(define (h ...) ... ( $\mathcal{D}$  (lambda (x)  $3x^4$ )) : ( $\lambda x$   $12x^3$ ) ...)
```

Shivers, III, O. G. (1991). *Control-Flow Analysis of Higher-Order Languages or Taming Lambda*. Ph.D. thesis, CMU.

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

```
(define ((compose n f) x)
  (if (zero? n) x ((compose (- n 1) f) (f x))))
```

Shivers, III, O. G. (1991). *Control-Flow Analysis of Higher-Order Languages or Taming Lambda*. Ph.D. thesis, CMU.

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

```
(define ((compose n f) x)
  (if (zero? n) x ((compose (- n 1) f) (f x))))

((compose k  $\mathcal{D}$ ) g)
```

Shivers, III, O. G. (1991). *Control-Flow Analysis of Higher-Order Languages or Taming Lambda*. Ph.D. thesis, CMU.

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

```
(define ((compose n f) x)
  (if (zero? n) x ((compose (- n 1) f) (f x))))
```

```
((compose k  $\mathcal{D}$ ) g)
```

```
(define ( $\mathcal{D}_{\text{compose}}$  f:g) ...)
```

Shivers, III, O. G. (1991). *Control-Flow Analysis of Higher-Order Languages or Taming Lambda*. Ph.D. thesis, CMU.

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

```
(define ((compose n f) x)
  (if (zero? n) x ((compose (- n 1) f) (f x))))
```

```
((compose k  $\mathcal{D}$ ) g)
```

```
(define ( $\mathcal{D}_{\text{compose}}$  f:g) ...)
```

```
(define ( $\mathcal{D}_{\text{compose:compose}}$  f:g') ...)
```

Shivers, III, O. G. (1991). *Control-Flow Analysis of Higher-Order Languages or Taming Lambda*. Ph.D. thesis, CMU.

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

```
(define ((compose n f) x)
  (if (zero? n) x ((compose (- n 1) f) (f x))))
```

```
((compose k  $\mathcal{D}$ ) g)
```

```
(define ( $\mathcal{D}_{\text{compose}}$  f:g) ...)
```

```
(define ( $\mathcal{D}_{\text{compose:compose}}$  f:g') ...)
```

```
(define ( $\mathcal{D}_{\text{compose:compose:compose}}$  f:g'') ...)
```

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Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

```
(define ((compose n f) x)
  (if (zero? n) x ((compose (- n 1) f) (f x))))
```

```
((compose k  $\mathcal{D}$ ) g)
```

```
(define ( $\mathcal{D}_{\text{compose}}$  f:g) ...)
```

```
(define ( $\mathcal{D}_{\text{compose:compose}}$  f:g') ...)
```

```
(define ( $\mathcal{D}_{\text{compose:compose:compose}}$  f:g'') ...)
```

⋮

Shivers, III, O. G. (1991). *Control-Flow Analysis of Higher-Order Languages or Taming Lambda*. Ph.D. thesis, CMU.

Polyvariant Flow Analysis

with Unbounded Context Sensitivity

$$\mathcal{E} : e \times \sigma \rightarrow v$$

Polyvariant Flow Analysis

with Unbounded Context Sensitivity

$$\mathcal{E} : e \times \sigma \rightarrow v$$

$$v ::= \#t \mid \#f \mid () \mid \mathbb{R} \mid (v_1, v_2) \mid \langle \sigma, e \rangle$$

$$\sigma ::= \{x_1 \mapsto v_1, \dots\}$$

Polyvariant Flow Analysis

with Unbounded Context Sensitivity

$$\mathcal{E} : e \times \sigma \rightarrow v$$

$$v ::= \#t \mid \#f \mid () \mid \mathbb{R} \mid (v_1, v_2) \mid \langle \sigma, e \rangle$$

$$\sigma ::= \{x_1 \mapsto v_1, \dots\}$$

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Polyvariant Flow Analysis

with Unbounded Context Sensitivity

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Memoize $\bar{\mathcal{E}}$ indexed (by suitable equivalence relations on) e and $\bar{\sigma}$.

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Side benefit: union-free

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No tags, tag checking, tag dispatching, indirect calls

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Side benefits: union-free, no cyclic abstract values

No tags, tag checking, tag dispatching, indirect calls

Allows complete unboxing: no allocation, reclamation, indirection

Game Theory

| | | | | | | |
|-----|----------|---------|---------------------------|----------|----------|-------|
| | | | B | | | |
| | | b_1 | \dots | b_j | \dots | b_n |
| | a_1 | | | | | |
| | \vdots | | \ddots | \vdots | | |
| A | a_i | \dots | $\text{PAYOFF}(a_i, b_j)$ | \dots | | |
| | \vdots | | \vdots | | \ddots | |
| | a_m | | | | | |

von Neumann, J. and Morgenstern, O. (1944). *Theory of Games and Economic Behavior*. Princeton University Press, Princeton, NJ.

Game Theory

| | | | | | | |
|-----|----------|-------|----------|----------------------|-----|----------|
| | | B | | | | |
| | | b_1 | ... | b_j | ... | b_n |
| | a_1 | | | | | |
| | \vdots | | \ddots | \vdots | | |
| A | a_i | ... | | PAYOFF(a_i, b_j) | ... | |
| | \vdots | | | \vdots | | \ddots |
| | a_m | | | | | |

$$\max_{a \in A} \min_{b \in B} \text{PAYOFF}(a, b)$$

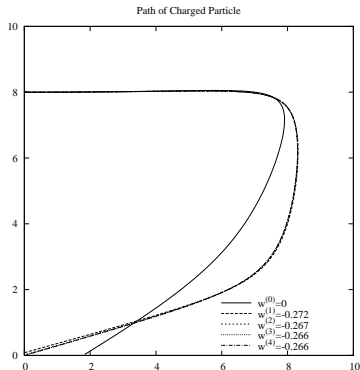
von Neumann, J. and Morgenstern, O. (1944). *Theory of Games and Economic Behavior*. Princeton University Press, Princeton, NJ.

| | | | |
|----------------|-----|----------------|-------------------------------|
| | | \mathbb{R}^n | |
| | | ... | b |
| | | ... | ... |
| \mathbb{R}^m | ... | ... | ... |
| a | | ... | PAYOFF(a , b) |
| | | ... | ... |
| | | ... | ... |

$$\max_{\mathbf{a} \in \mathbb{R}^m} \min_{\mathbf{b} \in \mathbb{R}^n} \text{PAYOFF}(\mathbf{a}, \mathbf{b})$$

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Cathode Ray Tubes



$$\text{potential: } p(\mathbf{x}; w) = \|\mathbf{x} - (10, 10 - w)\|^{-1} + \|\mathbf{x} - (10, 0)\|^{-1}$$

$$\ddot{\mathbf{x}}(t) = -\nabla_{\mathbf{x}} p(\mathbf{x})|_{\mathbf{x}=\mathbf{x}(t)}$$

$$\dot{\mathbf{x}}(t + \Delta t) = \dot{\mathbf{x}}(t) + \Delta t \ddot{\mathbf{x}}(t)$$

$$\mathbf{x}(t + \Delta t) = \mathbf{x}(t) + \Delta t \dot{\mathbf{x}}(t)$$

$$\text{When: } x_1(t + \Delta t) \leq 0$$

$$\text{let: } \Delta t_f = -x_1(t) / \dot{x}_1(t)$$

$$t_f = t + \Delta t_f$$

$$\mathbf{x}(t_f) = \mathbf{x}(t) + \Delta t_f \dot{\mathbf{x}}(t)$$

$$\text{Error: } E(w) = x_0(t_f)^2$$

$$\text{Find: } \underset{w}{\operatorname{argmin}} E(w)$$

Sprague, C. S. and George, R. H. (1939). *Cathode Ray Deflecting Electrode*. US Patent 2,161,437.

George, R. H. (1940). *Cathode Ray Tube*. US Patent 2,222,942.

Performance Comparison

| | | particle | | | | saddle | | | |
|---------|-----------------|----------|---------|--------|---------|---------|---------|-------|--------|
| | | FF | FR | RF | RR | FF | FR | RF | RR |
| VLAD | STALIN ∇ | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 | 1.00 |
| FORTRAN | ADIFOR | 1.52 | ■ | ■ | ■ | 2.07 | ■ | ■ | ■ |
| | TAPENADE | 3.40 | ■ | ■ | ■ | 2.56 | ■ | ■ | ■ |
| C++ | FADBAD++ | 65.69 | ■ | ■ | ■ | 22.44 | ■ | ■ | ■ |
| ML | MLTON | 53.89 | 88.88 | 16.08 | 28.06 | 40.39 | 51.21 | 1.86 | 2.67 |
| | OCAML | 160.50 | 340.35 | 147.91 | 263.66 | 107.71 | 156.33 | 6.75 | 13.51 |
| | SML/NJ | 106.21 | 182.45 | 105.04 | 185.15 | 84.38 | 106.01 | 3.55 | 6.31 |
| HASKELL | GHC | 165.22 | ■ | ■ | ■ | 121.18 | ■ | ■ | ■ |
| SCHEME | BIGLOO | 505.90 | 761.40 | 104.81 | 228.56 | 423.69 | 440.25 | 15.77 | 24.59 |
| | CHICKEN | 1120.37 | 2026.31 | 425.60 | 1872.85 | 889.58 | 1144.65 | 35.73 | 68.94 |
| | GAMBIT | 444.13 | 752.63 | 138.34 | 256.30 | 362.65 | 420.48 | 14.08 | 23.87 |
| | IKARUS | 192.07 | 312.28 | 61.79 | 114.87 | 158.88 | 205.97 | 6.75 | 11.40 |
| | LARCENY | 726.59 | 1108.18 | 144.55 | 270.14 | 571.81 | 613.65 | 19.14 | 29.77 |
| | MIT SCHEME | 1472.26 | 2500.00 | 309.66 | 591.36 | 1243.26 | 1428.57 | 51.36 | 79.10 |
| | MzC | 2073.26 | 3434.64 | 340.30 | 655.83 | 2436.26 | 1996.40 | 72.45 | 150.02 |
| | MzSCHEME | 2344.70 | 4076.16 | 409.95 | 843.68 | 2000.89 | 2332.43 | 80.78 | 134.00 |
| | SCHEME->C | 391.42 | 605.26 | 109.77 | 198.43 | 324.95 | 328.84 | 12.74 | 18.28 |
| | SCMUTILS | 3321.20 | ■ | ■ | ■ | 2800.71 | ■ | ■ | ■ |
| | STALIN | 208.10 | 366.08 | 51.84 | 91.86 | 166.96 | 212.93 | 7.68 | 11.40 |

- not implemented but could implement
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Gradient-Based Optimization

```
(define (e i n)
  (if (zero? n)
      '()
      (cons (if (zero? i) 1.0 0.0)
            (e (- i 1) (- n 1)))))
```

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```
(define (e i n)
  (if (zero? n)
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(define ((gradient f) x)
  (let ((n (length x)))
    (map (lambda (i) (tangent ((j* f) (bundle x (e i n))))
         (iota n)))))
```

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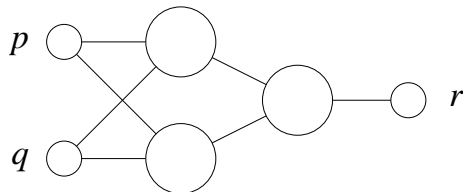
(define ((gradient f) x)
  (let ((n (length x)))
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         (iota n))))

(define (gradient-ascent f x0 n eta)
  (if (zero? n)
      (list x0 (f x0) ((gradient f) x0))
      (gradient-ascent f
                       (zip (lambda (xi gi) (+ xi (* eta gi)))
                            x0
                            ((gradient f) x0))
                       (- n 1)
                       eta)))
```

Gradient-Based Optimization

```
(define ((gradient f) x) (cdr ((cdr ((*j f) (*j x))) 1.0)))
```

```
(define (gradient-ascent f x0 n eta)
  (if (zero? n)
      (list x0 (f x0) ((gradient f) x0))
      (gradient-ascent f
                        (zip (lambda (xi gi) (+ xi (* eta gi)))
                            x0
                            ((gradient f) x0))
                        (- n 1)
                        eta)))
```



| p | q | r |
|-----|-----|-----|
| 0 | 0 | 0 |
| 0 | 1 | 1 |
| 1 | 0 | 1 |
| 1 | 1 | 0 |

Rumelhart, D. E., Hinton, G. E., and Williams, R. J. (1986). *Learning representations by back-propagating errors*. Nature, **323**:533–6.

Neural Networks in VLAD

```
(define ((sum-activities activities) bias ws)
  ((fold + bias) (zip * ws activities)))
```

Neural Networks in VLAD

```
(define ((sum-activities activities) bias ws)
  ((fold + bias) (zip * ws activities)))

(define (sum-layer activities ws-layer)
  (map (sum-activities activities) ws-layer))
```

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(define (sigmoid x) (/ 1 (+ (exp (- 0 x)) 1)))
```

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```
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(define (sum-layer activities ws-layer)
  (map (sum-activities activities) ws-layer))

(define (sigmoid x) (/ 1 (+ (exp (- 0 x)) 1)))

(define ((forward-pass ws-layers) in)
  (if (null? ws-layers)
      in
      ((forward-pass (cdr ws-layers))
       (map sigmoid (sum-layer in (car ws-layers))))))
```

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```
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      in
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(define ((error-on-dataset dataset) ws-layers)
  ((fold + 0)
   (map (lambda ((list in target))
          (* 0.5 (magnitude-squared (v- ((forward-pass ws-layers) in) target))))
        dataset)))
```

Neural Networks in VLAD

```
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         dataset)))

(gradient-descent (error-on-dataset '(((0 0) (0))
                                       ((0 1) (1))
                                       ((1 0) (1))
                                       ((1 1) (0))))
                  '(((0 -0.284227 1.16054) (0 0.617194 1.30467))
                    ((0 -0.084395 0.648461)))
                  1000.0
                  0.3)
```

Performance Comparison

| | | backprop | | |
|---------|-----------------|----------|-------|---------|
| | | Fs | Fv | R |
| VLAD | STALIN ∇ | 1.00 | ■ | 1.00 |
| FORTRAN | ADIFOR | 11.84 | 2.68 | ■ |
| | TAPENADE | 11.35 | 4.33 | 6.24 |
| C | ADIC | 16.33 | 3.93 | ■ |
| C++ | ADOL-C | 12.34 | 3.89 | 35.53 |
| | CPPAD | 42.15 | ■ | 23.69 |
| | FADBAD++ | 98.96 | 33.15 | 53.03 |
| ML | MLTON | 73.94 | ■ | 37.94 |
| | OCAML | 157.75 | ■ | 149.14 |
| | SML/NJ | 142.71 | ■ | 94.97 |
| HASKELL | GHC | ■ | ■ | ■ |
| SCHEME | BIGLOO | 577.45 | ■ | 306.60 |
| | CHICKEN | 1391.75 | ■ | 971.91 |
| | GAMBIT | 545.20 | ■ | 341.73 |
| | IKARUS | 216.42 | ■ | 147.49 |
| | LARCENY | 955.98 | ■ | 486.64 |
| | MIT SCHEME | 1900.04 | ■ | 1141.22 |
| | MzC | 2439.93 | ■ | 1571.52 |
| | MzSCHEME | 3477.86 | ■ | 1866.28 |
| | SCHEME->C | 484.24 | ■ | 233.75 |
| | SCMUTILS | 4544.48 | ■ | ■ |
| | STALIN | 832.68 | ■ | 367.84 |

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$P = \mathbf{if } x_0 \mathbf{ then } 0 \mathbf{ else if } x_1 \mathbf{ then } 1 \mathbf{ else } 2$

Koller, D., McAllester, D. , and Pfeffer, A. (1997). *Effective Bayesian Inference for Stochastic Programs*. Proceedings of the 14th National Conference on Artificial Intelligence (AAAI), pp. 740–7.

Probabilistic Lambda Calculus

$P = \text{if } x_0 \text{ then } 0 \text{ else if } x_1 \text{ then } 1 \text{ else } 2$

$$\Pr(x_0 \mapsto \mathbf{true}) = p_0$$

$$\Pr(x_0 \mapsto \mathbf{false}) = 1 - p_0$$

$$\Pr(x_1 \mapsto \mathbf{true}) = p_1$$

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$$\Pr(\mathcal{E}(P) = 0 | p_0, p_1) = p_0$$

$$\Pr(\mathcal{E}(P) = 1 | p_0, p_1) = (1 - p_0)p_1$$

$$\Pr(\mathcal{E}(P) = 2 | p_0, p_1) = (1 - p_0)(1 - p_1)$$

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$$\prod_{v \in \{0,1,2,2\}} \Pr(\mathcal{E}(P) = v | p_0, p_1) = p_0(1 - p_0)^3 p_1(1 - p_1)^2$$

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$$\prod_{v \in \{0,1,2,2\}} \Pr(\mathcal{E}(P) = v | p_0, p_1) = p_0(1 - p_0)^3 p_1(1 - p_1)^2$$

$$\operatorname{argmax}_{p_0, p_1} \prod_{v \in \{0,1,2,2\}} \Pr(\mathcal{E}(P) = v | p_0, p_1) = \left\langle \frac{1}{4}, \frac{1}{3} \right\rangle$$

Koller, D., McAllester, D., and Pfeffer, A. (1997). *Effective Bayesian Inference for Stochastic Programs*. Proceedings of the 14th National Conference on Artificial Intelligence (AAAI), pp. 740–7.

Probabilistic Prolog

$p(0).$

$p(X) :- q(X).$

$q(1).$

$q(2).$

Probabilistic Prolog

$$\Pr(\mathfrak{p}(0) \text{ .}) = p_0$$

$$\Pr(\mathfrak{p}(X) : \neg \mathfrak{q}(X) \text{ .}) = 1 - p_0$$

$$\Pr(\mathfrak{q}(1) \text{ .}) = p_1$$

$$\Pr(\mathfrak{q}(2) \text{ .}) = 1 - p_1$$

Probabilistic Prolog

$$\Pr(\text{p}(0) \text{ .}) = p_0$$

$$\Pr(\text{p}(X) : \text{-q}(X) \text{ .}) = 1 - p_0$$

$$\Pr(\text{q}(1) \text{ .}) = p_1$$

$$\Pr(\text{q}(2) \text{ .}) = 1 - p_1$$

$$\Pr(\text{?-p}(0) \text{ .}) = p_0$$

$$\Pr(\text{?-p}(1) \text{ .}) = (1 - p_0)p_1$$

$$\Pr(\text{?-p}(2) \text{ .}) = (1 - p_0)(1 - p_1)$$

Probabilistic Prolog

$$\Pr(\mathbf{p}(0) \text{ .}) = p_0$$

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$$\Pr(?-\mathbf{p}(0) \text{ .}) = p_0$$

$$\Pr(?-\mathbf{p}(1) \text{ .}) = (1 - p_0)p_1$$

$$\Pr(?-\mathbf{p}(2) \text{ .}) = (1 - p_0)(1 - p_1)$$

$$\prod_{q \in \{\mathbf{p}(0), \mathbf{p}(1), \mathbf{p}(2), \mathbf{p}(2)\}} \Pr(?-q \text{ .}) = p_0(1 - p_0)^3 p_1(1 - p_1)^2$$

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$$\operatorname{argmax}_{p_0, p_1} \prod_{q \in \{\mathbf{p}(0), \mathbf{p}(1), \mathbf{p}(2), \mathbf{p}(2)\}} \Pr(?-q \text{ .}) = \left\langle \frac{1}{4}, \frac{1}{3} \right\rangle$$

Probabilistic Lambda Calculus

```
(define (evaluate expression environment)
  (cond
    ((constant-expression? expression)
     (singleton-tagged-distribution
      (constant-expression-value expression)))
    ((variable-access-expression? expression)
     (lookup-value
      (variable-access-expression-variable expression) environment))
    ((lambda-expression? expression)
     (singleton-tagged-distribution
      (lambda (tagged-distribution)
        (evaluate
         (lambda-expression-body expression)
         (cons (make-binding (lambda-expression-variable expression)
                             tagged-distribution)
               environment))))))
    (else (let ((tagged-distribution
                 (evaluate (application-argument expression)
                           environment)))
                (map-tagged-distribution
                 (lambda (value) (value tagged-distribution))
                 (evaluate (application-callee expression) environment))))))
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Probabilistic Lambda Calculus

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Probabilistic Lambda Calculus

```
(gradient-ascent
 (lambda (p)
  (let ((tagged-distribution
        (evaluate if  $x_0$  then 0 else if  $x_1$  then 1 else 2
                (list  $\Pr(x_0 \mapsto \mathbf{true}) = p_0$   $\Pr(x_0 \mapsto \mathbf{false}) = 1 - p_0$ 
                       $\Pr(x_1 \mapsto \mathbf{true}) = p_1$   $\Pr(x_1 \mapsto \mathbf{false}) = 1 - p_1$ 
                      ...)))
    (map-reduce
     *
     1.0
     (lambda (value)
      (likelihood value tagged-distribution))
     '(0 1 2 2))))
 '(0.5 0.5)
 1000.0
 0.1)
```

Probabilistic Lambda Calculus

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        (evaluate if  $x_0$  then 0 else if  $x_1$  then 1 else 2
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Probabilistic Lambda Calculus

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Probabilistic Lambda Calculus

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Probabilistic Lambda Calculus

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      (likelihood value tagged-distribution))
     '(0 1 2 2))))
 '(0.5 0.5)
1000.0
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```

Probabilistic Prolog

```
(define (proof-distribution term clauses)
  (let ((offset ...))
    (map-reduce
      append
      '()
      (lambda (clause)
        (let ((clause (alpha-rename clause offset)))
          (let loop ((p (clause-p clause))
                     (substitution (unify term (clause-term clause)))
                     (terms (clause-terms clause)))
            (if (boolean? substitution)
                '()
                (if (null? terms)
                    (list (make-double p substitution))
                    (map-reduce
                      append
                      '()
                      (lambda (double)
                        (loop (* p (double-p double))
                              (append substitution (double-substitution double))
                              (rest terms))))
                    (proof-distribution
                     (apply-substitution substitution (first terms)) clauses)))))))
    clauses)))
```

Probabilistic Prolog

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(define (proof-distribution term clauses)
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                     (substitution (unify term (clause-term clause)))
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            (if (boolean? substitution)
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                      '()
                      (lambda (double)
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                    (proof-distribution
                     (apply-substitution substitution (first terms)) clauses)))))))
    clauses)))
```

Probabilistic Prolog

```
(define (proof-distribution term clauses)
  (let ((offset ...))
    (map-reduce
      append
      '()
      (lambda (clause)
        (let ((clause (alpha-rewrite clause offset)))
          (let loop ((p (clause-p clause))
                     (substitution (unify term (clause-term clause)))
                     (terms (clause-terms clause)))
            (if (boolean? substitution)
                '()
                (if (null? terms)
                    (list (make-double p substitution))
                    (map-reduce
                      append
                      '()
                      (lambda (double)
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Probabilistic Prolog

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  (let ((clauses (list Pr(p(0).) = p0
                       Pr(p(X):-q(X).) = 1 - p0
                       Pr(q(1).) = p1
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     (lambda (query)
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     '(p(0) p(1) p(2) p(2))))
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Generated Code

```
static void f2679(double a_f2679_0,double a_f2679_1,double a_f2679_2,double a_f2679_3){
    int t272381=((a_f2679_2==0.)?0:1);
    double t272406;
    double t272405;
    double t272404;
    double t272403;
    double t272402;
    if((t272381==0)){
        double t272480=(1.-a_f2679_0);
        double t272572=(1.-a_f2679_1);
        double t273043=(a_f2679_0+0.);
        double t274185=(t272480*a_f2679_1);
        double t274426=(t274185+0.);
        double t275653=(t272480*t272572);
        double t275894=(t275653+0.);
        double t277121=(t272480*t272572);
        double t277362=(t277121+0.);
        double t277431=(t277362*1.);
        double t277436=(t275894*t277431);
        double t277441=(t274426*t277436);
        double t277446=(t273043*t277441);
        ...
        double t1777107=(t1774696+t1715394);
        double t1777194=(0.-t1745420);
        double t1778533=(t1777194+t1419700);
        t272406=a_f2679_0;
        t272405=a_f2679_1;
        t272404=t277446;
        t272403=t1778533;
        t272402=t1777107;}
    else {...}
    r_f2679_0=t272406;
    r_f2679_1=t272405;
    r_f2679_2=t272404;
    r_f2679_3=t272403;
    r_f2679_4=t272402;}
```

Performance Comparison

| | | probabilistic-lambda-calculus | | probabilistic-prolog | |
|---------|-----------------|-------------------------------|---------|----------------------|----------|
| | | F | R | F | R |
| VLAD | STALIN ∇ | 1.00 | 1.00 | 1.00 | 1.00 |
| ML | MLTON | 106.45 | 124.95 | 789.41 | 483.47 |
| | OCAML | 215.73 | 538.68 | 1207.13 | 1534.61 |
| | SML/NJ | 197.75 | 272.45 | 2448.02 | 1471.94 |
| HASKELL | GHC | ■ | ■ | ■ | ■ |
| SCHEME | BIGLOO | 832.92 | 1048.11 | 14422.16 | 8286.06 |
| | CHICKEN | 2305.98 | 3283.00 | 66948.70 | 37792.84 |
| | GAMBIT | 879.88 | 1153.86 | 24316.03 | 13649.81 |
| | IKARUS | 437.46 | 531.10 | 8242.92 | 4845.86 |
| | LARCENY | 1651.01 | 1673.22 | 25589.62 | 14833.53 |
| | MIT SCHEME | 3491.10 | 4130.19 | 85819.57 | 48335.38 |
| | MzC | 5289.17 | 5929.14 | 154206.95 | 83480.27 |
| | MzSCHEME | 6235.78 | 7134.71 | 166129.12 | 91630.70 |
| | SCHEME->C | 682.15 | 794.31 | 10530.66 | 5980.27 |
| | SCMUTILS | 6456.99 | ■ | 80100.23 | ■ |
| | STALIN | 1240.73 | 1137.41 | 22511.79 | 10986.43 |

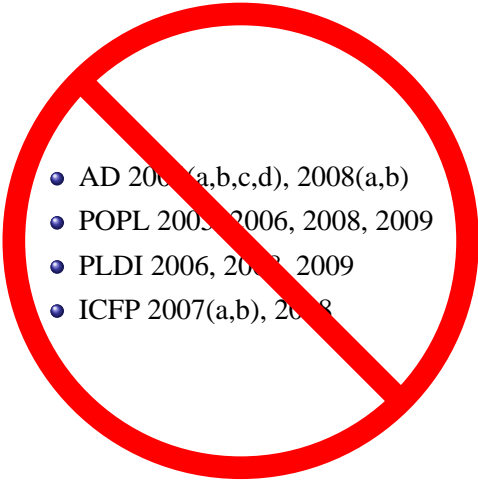
- not implemented but could implement, including FORTRAN, C, and C++
- not implemented in existing tool
- can't implement

Where You Can Read About This Work

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- AD 2004(a,b,c,d), 2008(a,b)
- POPL 2005, 2006, 2008, 2009
- PLDI 2006, 2008, 2009
- ICFP 2007(a,b), 2008

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Evaluation: C

It is not clear to me how this approach is better than automatic differentiation frameworks using C++ templates and the C++ type inference algorithm.

The paper seems correct, and I agree that it is hard to implement automatic differentiation both conveniently and efficiently without the proper programming language support (i.e., efficiently implemented generics with a minimal amount of type inference).

It's not clear to me that an efficient C++ templates implementation doesn't give the benefits listed here: it is "callee derives" (mostly) and efficiently implemented (using inlining). (The extent that it's not "callee derives" is that you need to put "template<class X>" in front of every function and declare all your variables as being of type "X" instead of "double".)

I was somewhat shocked that the only C++ implementation you compared to was apparently one that uses virtual function dispatch (FADBAD++). I Googled "automatic differentiation C++ template" and found two papers that do what I expected you to compare to:

Dan Piponi: Automatic Differentiation, C++ Templates and Photogrammetry, "to be published in the Journal of Graphics Tools", Sep. 2004.

and

M.E. Jerrell: Function Minimization and Automatic Differentiation Using C++, OOPSLA, 1989.

Evaluation: C. Weak paper, but it will not be an embarrassment to have it in POPL.
Confidence: Z. I am an informed outsider and tried my best to understand the paper.

==== Summary ====

Shows how to optimize a functional language with a built-in automatic differentiation operator.

==== Detailed Comments =====

The results look useful, but I wonder whether POPL is the right place to present them. Yes, the development involves functional programming. But it also involves a lot of concepts from scientific computing that may be unfamiliar to many and that are explained only minimally or not at all.

It is, of course, not excluded that the range of arguments or range of values of a function should consist wholly or partly of functions. The derivative, as this notion appears in the elementary differential calculus, is a familiar mathematical example of a function for which both ranges consist of functions.

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