

Automatic Differentiation of Functional Programs or Lambda the Ultimate Calculus

Jeffrey Mark Siskind
qobi@purdue.edu

School of Electrical and Computer Engineering
Purdue University

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

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

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

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

$$\langle x_1, x'_1 \rangle + \langle x_2, x'_2 \rangle = \langle (x_1 + x_2), (x'_1 + x'_2) \rangle$$
$$\langle x_1, x'_1 \rangle \times \langle x_2, x'_2 \rangle = \langle (x_1 \times x_2), (x_1 \times x'_2 + x_2 \times x'_1) \rangle$$

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

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))
                   (+ (* (primal x1) (tangent x2))
                      (* (tangent x1) (primal x2))))))

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

Dynamic Overloading: SCMUTILS

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

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

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(D f)
```

Dynamic Overloading: SCMUTILS

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

Dynamic Overloading: SCMUTILS

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(D (D f))
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Convenient

Dynamic Overloading: SCMUTILS

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(D (D f))
(D (lambda (x) ... (D (lambda (y) ...) ...) ...) ...)
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Convenient but **slow**

Dynamic Overloading: SCMUTILS

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                   (+ (* (primal x1) (tangent x2))
                      (* (tangent x1) (primal x2))))))

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

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(D (D f))
(D (lambda (x) ... (D (lambda (y) ...) ...) ...)) ...)
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  (let ((+ +) (* *))
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      (make-bundle (* (primal x1) (primal x2))
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                      (* (tangent x1) (primal x2))))))

(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))
(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))
(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

```
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
```

Fast

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
```

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
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
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
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
end
```

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

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
end
```

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

```
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**

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
end
```

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

```
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**

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
end
```

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

```
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
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Fast but **inconvenient**

Preprocessor: ADIFOR and TAPENADE

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

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function ggf(x, 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
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ggresult = 6.0d0*x*x*ggx+12.0d0*x*gx*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
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) ...
```

Static Overloading: FADBAD++

```
double f(double x) {return 2*x*x*x;}  
double x;  
... f(x) ...
```

```
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++

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

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double x;  
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F<F<double> > f(F<F<double> > x) {return 2*x*x*x;}  
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```

Slow and **inconvenient**

Static Overloading: FADBAD++

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

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```

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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);  
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... f(x).d(0).d(0) ...
```

Slow and **inconvenient**

Static Overloading: FADBAD++

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double f(double x) {return 2*x*x*x;}  
double x;  
... f(x) ...
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F<double> f(F<double> x) {return 2*x*x*x;}  
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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++

```
double f(double x) {return 2*x*x*x;}
double x;
... f(x) ...
```

```
F<double> f(F<double> x) {return 2*x*x*x;}
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Slow and **inconvenient**

Static Overloading: FADBAD++

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

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

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

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

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

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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

$$\begin{aligned}\text{bundle} & : \tau \times \overline{\tau} \rightarrow (\tau \triangleright \overline{\tau}) \\ \text{primal} & : (\tau \triangleright \overline{\tau}) \rightarrow \tau \\ \text{tangent} & : (\tau \triangleright \overline{\tau}) \rightarrow \overline{\tau} \\ \text{j*} & : (\tau_1 \rightarrow \tau_2) \rightarrow ((\tau_1 \triangleright \overline{\tau}_1) \rightarrow (\tau_2 \triangleright \overline{\tau}_2))\end{aligned}$$

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

Generalize to arbitrary types

Our API for Functional Forward AD

$$\begin{aligned}\text{bundle} & : \tau \times \overline{\tau} \rightarrow (\tau \triangleright \overline{\tau}) \\ \text{primal} & : (\tau \triangleright \overline{\tau}) \rightarrow \tau \\ \text{tangent} & : (\tau \triangleright \overline{\tau}) \rightarrow \overline{\tau} \\ \text{j*} & : (\tau_1 \rightarrow \tau_2) \rightarrow ((\tau_1 \triangleright \overline{\tau}_1) \rightarrow (\tau_2 \triangleright \overline{\tau}_2))\end{aligned}$$

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

j* maps a function to its *push forward*

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

j* maps a function to its *push forward*

Generalize to arbitrary types

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

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

Our API for Functional Forward AD

$$\begin{aligned}\text{bundle} &: \tau \times \overline{\tau} \rightarrow \overrightarrow{\tau} \\ \text{primal} &: \overrightarrow{\tau} \rightarrow \tau \\ \text{tangent} &: \overrightarrow{\tau} \rightarrow \overline{\tau} \\ \text{j*} &: (\tau_1 \rightarrow \tau_2) \rightarrow (\overrightarrow{\tau_1} \rightarrow \overrightarrow{\tau_2})\end{aligned}$$

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

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Our API for Functional Forward AD

$$\begin{aligned}\text{bundle} &: \tau \times \overline{\tau} \rightarrow \overline{\tau} \\ \text{primal} &: \overline{\tau} \rightarrow \tau \\ \text{tangent} &: \overline{\tau} \rightarrow \overline{\tau} \\ \overrightarrow{\mathcal{J}} &: (\tau_1 \rightarrow \tau_2) \rightarrow (\overline{\tau}_1 \rightarrow \overline{\tau}_2)\end{aligned}$$

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

j_* maps a function to its *push forward*

Generalize to arbitrary types

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

Can abbreviate $\tau \triangleright \overline{\tau}$ as $\overrightarrow{\tau}$

Sometimes write j_* as $\overrightarrow{\mathcal{J}}$

Our API for Functional Forward AD

$$\begin{aligned}\text{bundle} &: \tau \times \overrightarrow{\tau} \rightarrow \overrightarrow{\tau} \\ \text{primal} &: \overrightarrow{\tau} \rightarrow \tau \\ \text{tangent} &: \overrightarrow{\tau} \rightarrow \overrightarrow{\tau} \\ \text{j*} &: (\tau_1 \rightarrow \tau_2) \rightarrow (\overrightarrow{\tau_1} \rightarrow \overrightarrow{\tau_2})\end{aligned}$$

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

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

j* maps a function to its *push forward*

Generalize to arbitrary types

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

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Sometimes write j* as \mathcal{J}

Convenient

Our API for Functional Forward AD

```
bundle :  $\tau \times \overrightarrow{\tau} \rightarrow \overrightarrow{\tau}$   
primal :  $\overrightarrow{\tau} \rightarrow \tau$   
tangent :  $\overrightarrow{\tau} \rightarrow \overrightarrow{\tau}$   
j* :  $(\tau_1 \rightarrow \tau_2) \rightarrow (\overrightarrow{\tau}_1 \rightarrow \overrightarrow{\tau}_2)$ 
```

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

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

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```

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

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

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

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

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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?

Can abbreviate $\tau \triangleright \overline{\tau}$ as $\overrightarrow{\tau}$

Sometimes write j^* as \mathcal{J}

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

Convenient

Our API for Functional Forward AD

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

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Generalize to arbitrary types

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Sometimes write j^* as \mathcal{J}

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

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

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

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

A property

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

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

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

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

A property

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

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

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

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

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

A property

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

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

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

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

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

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

A property

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

A property

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

A property

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

A property

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

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

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

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

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

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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

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

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

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

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

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

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

0/1 matrix, every row has exactly one 1

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A property

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

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

0/1 matrix, every row has exactly one 1

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when f is a rearrangement function

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

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

What is the tangent of #t?

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

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when f is a rearrangement function

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f is a rearrangement function

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

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

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

A (Not So) Brief Tutorial on AD

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

$$y = f(x)$$

$$z = g(y)$$

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

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

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

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

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

$$\vdots$$

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

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

A (Not So) Brief Tutorial on AD

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

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

$$\vdots$$

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

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

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

A (Not So) Brief Tutorial on AD

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

A (Not So) Brief Tutorial on AD

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

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

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

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

A (Not So) Brief Tutorial on AD

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

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

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

⋮

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

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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

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

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

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

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

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

A (Not So) Brief Tutorial on AD

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

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

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

A (Not So) Brief Tutorial on AD

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

A (Not So) Brief Tutorial on AD

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

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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

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

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

⋮

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

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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

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

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

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

A (Not So) Brief Tutorial on AD

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

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

A (Not So) Brief Tutorial on AD

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

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

A (Not So) Brief Tutorial on AD

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

A (Not So) Brief Tutorial on AD

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

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

A (Not So) Brief Tutorial on AD

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

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

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

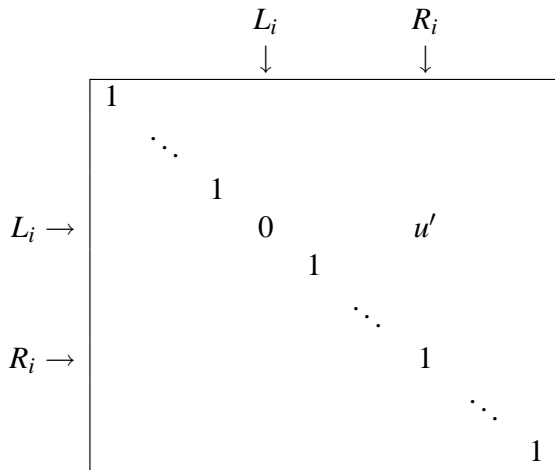
$$\mathbf{x}[L_i] := b(\mathbf{x}[R_i], \mathbf{x}[S_i])$$

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

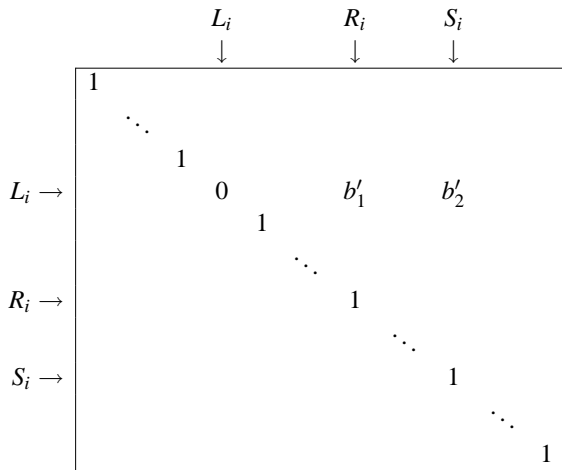
$$\mathbf{x}[L_i] := b(\mathbf{x}[R_i], \mathbf{x}[S_i])$$

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

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

A (Not So) Brief Tutorial on AD

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

A (Not So) Brief Tutorial on AD

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

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

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

A (Not So) Brief Tutorial on AD

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

A (Not So) Brief Tutorial on AD

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

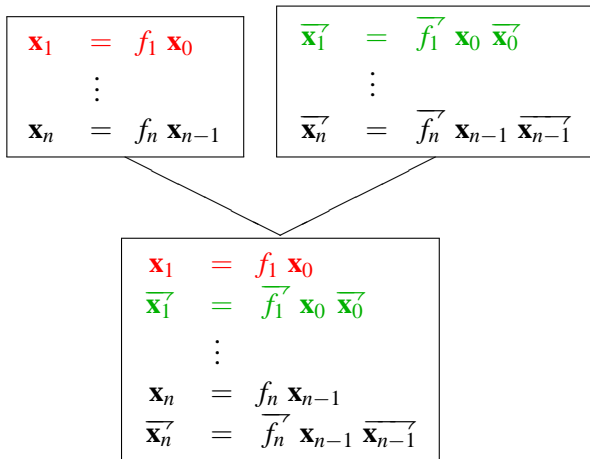
A (Not So) Brief Tutorial on AD

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

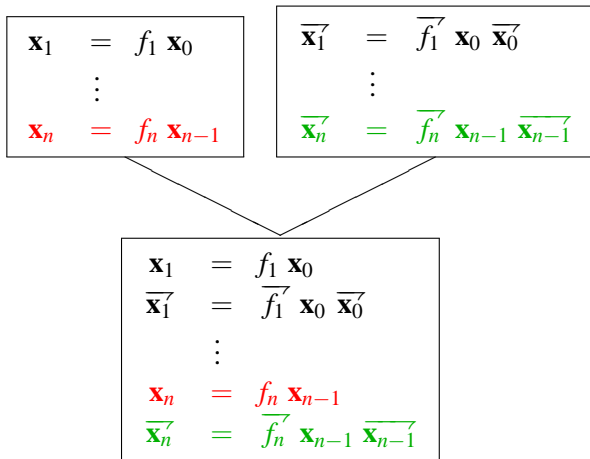
$$\begin{aligned}\overline{\mathbf{x}}_1' &= \overline{f_1}' \mathbf{x}_0 \overline{\mathbf{x}}_0' \\ &\vdots \\ \overline{\mathbf{x}}_n' &= \overline{f_n}' \mathbf{x}_{n-1} \overline{\mathbf{x}}_{n-1}'\end{aligned}$$

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

A (Not So) Brief Tutorial on AD



A (Not So) Brief Tutorial on AD



A (Not So) Brief Tutorial on AD

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

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

A (Not So) Brief Tutorial on AD

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

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

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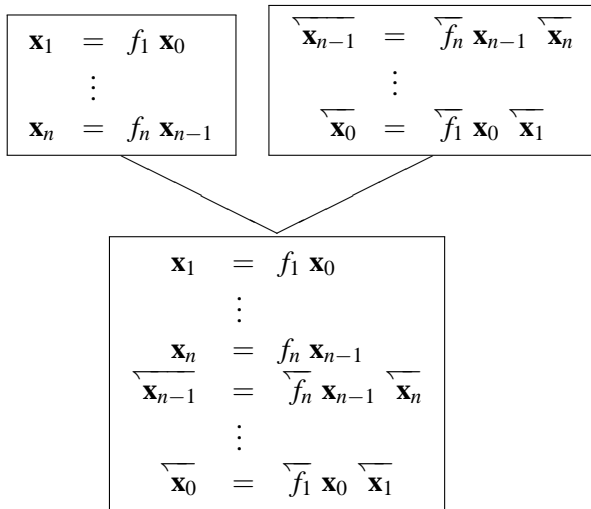
$$\begin{aligned}x_{L_i} &:= u_i x_{R_i} &\rightsquigarrow& \overrightarrow{x_{L_i}} := \overrightarrow{u_i} \overrightarrow{x_{R_i}} \\x_{L_i} &:= b_i (x_{R_i}, x_{S_i}) &\rightsquigarrow& \overrightarrow{x_{L_i}} := \overrightarrow{b_i} (\overrightarrow{x_{R_i}}, \overrightarrow{x_{S_i}})\end{aligned}$$

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

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

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

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A (Not So) Brief Tutorial on AD

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

$$\vdots$$

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

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

$$\vdots$$

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

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

$$\vdots$$

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

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

$$\vdots$$

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

A (Not So) Brief Tutorial on AD

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

$$\vdots$$

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

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

$$\vdots$$

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

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

$$\vdots$$

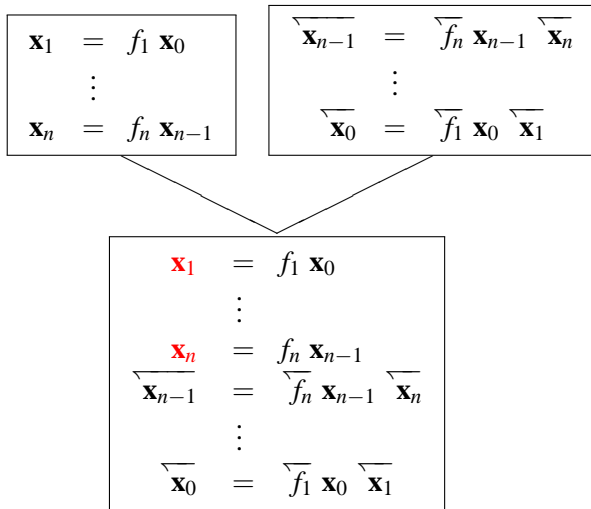
$$\mathbf{x}_n = f_n \mathbf{x}_{n-1}$$

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

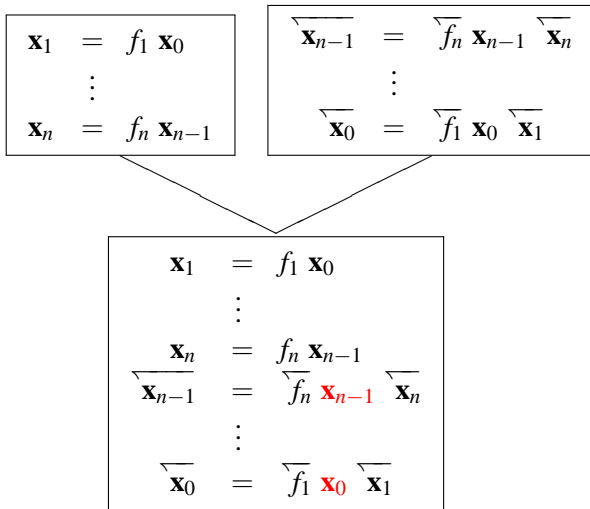
$$\vdots$$

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

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A (Not So) Brief Tutorial on AD



A (Not So) Brief Tutorial on AD

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

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

⋮

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

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

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

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

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

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

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

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

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

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

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

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

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

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

The Functional Reverse-Mode Transformation

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

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

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

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

The Functional Reverse-Mode Transformation

$$\left. \begin{array}{l}
 \lambda x_0 \mathbf{let} \ x_1 \triangleq x_{R_1} \ x_{S_1}; \\
 \quad \vdots \\
 \quad x_n \triangleq x_{R_n} \ x_{S_n} \\
 \mathbf{in} \ x_n \ \mathbf{end}
 \end{array} \right\} \rightsquigarrow \left\{ \begin{array}{l}
 \lambda \overleftarrow{x_0} \ \mathbf{let} \ (\overleftarrow{x_1}, \overline{x_1}) \triangleq \overleftarrow{x_{R_1}} \ \overleftarrow{x_{S_1}}; \\
 \quad \vdots \\
 \quad (\overleftarrow{x_n}, \overline{x_n}) \triangleq \overleftarrow{x_{R_n}} \ \overleftarrow{x_{S_n}} \\
 \mathbf{in} \ (\overleftarrow{x_n}, (\lambda \overleftarrow{x_n} \ \mathbf{let} \ \overleftarrow{x_0} \triangleq \mathbf{0} \ (\overleftarrow{\mathcal{J}}^{-1} \ \overleftarrow{x_0})); \\
 \quad \vdots \\
 \quad \overleftarrow{x_{n-1}} \triangleq \mathbf{0} \ (\overleftarrow{\mathcal{J}}^{-1} \ \overleftarrow{x_{n-1}}); \\
 \quad \overleftarrow{x_{S_n}} \oplus \triangleq \overline{x_n} \ \overleftarrow{x_n}; \\
 \quad \vdots \\
 \quad \overleftarrow{x_{S_1}} \oplus \triangleq \overline{x_1} \ \overleftarrow{x_1} \\
 \mathbf{in} \ \overleftarrow{x_0} \ \mathbf{end}) \ \mathbf{end}
 \end{array} \right.$$

The above is a white lie. The truth is a lot more complicated.

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

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

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

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

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

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

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

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

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

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

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

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Breaking Modularity

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

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

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

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

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

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

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Breaking Modularity

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

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

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

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

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

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

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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

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

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

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

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

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

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

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Proceedings of the American Philosophy Society, **90**:20–4.

Breaking Modularity

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

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

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

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

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

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

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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

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

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

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

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

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

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

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

Fermi, E. (1946). *The Development of the first chain reaction pile.*

Proceedings of the American Philosophy Society, **90**:20–4.

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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Breaking Modularity

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

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

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

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

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

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

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

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

Fermi, E. (1946). *The Development of the first chain reaction pile.*

Proceedings of the American Philosophy Society, **90**:20–4.

Breaking Modularity

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

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

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

$$\text{NEUTRONFLUX} \quad \xrightarrow{\text{ADIFOR}} \quad \overline{\text{NEUTRONFLUX}}$$

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

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

$$\text{NEUTRONFLUX} \quad \overset{\text{TAPENADE}}{\rightsquigarrow} \quad \overleftarrow{\text{NEUTRONFLUX}}$$

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

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$\nabla \overleftarrow{f} \mathbf{x}$	\triangleq	$\dots \overleftarrow{f} \mathbf{x} \dots$
GRADIENTDESCENT $\overleftarrow{f} \mathbf{x}_0$	\triangleq	$\dots \mathbf{x}_{i+1} := \dots \nabla \overleftarrow{f} \mathbf{x}_i \dots$
NEWTONSMETHOD $\overleftarrow{f} \mathbf{x}_0$	\triangleq	$\dots \mathbf{x}_{i+1} := \dots \nabla \overleftarrow{f} \mathbf{x}_i \dots \mathcal{H}f \mathbf{x}_i \dots$
argmin \overleftarrow{f}	\triangleq	$\dots \text{GRADIENTDESCENT } \overleftarrow{f} \mathbf{x}_0 \dots$
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	$\overset{\text{TAPENADE}}{\rightsquigarrow}$	$\overleftarrow{\text{DEVIATION}}$
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$$\text{argmin} \overleftarrow{f} \quad \triangleq \quad \dots \text{NEWTONSMETHOD} \overleftarrow{f} \mathbf{x}_0 \dots$$

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Breaking Modularity

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

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

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	$((\text{NEUTRONFLUX } \mathbf{r}) - \text{NEUTRONFLUX}_{\text{critical}})^2$
\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	$\dots (\overrightarrow{\mathcal{J}} (\overleftarrow{\mathcal{J}} f)) \dots \mathbf{x} \dots$
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	<div style="border: 1px solid black; padding: 2px; display: inline-block;">classified</div>
DEVIATION \mathbf{r}	\triangleq	$((\text{NEUTRONFLUX } \mathbf{r}) - \text{NEUTRONFLUX}_{\text{critical}})^2$
\mathbf{r}^*	\triangleq	argmin DEVIATION

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

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- 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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 - $(\mathcal{D} (\mathcal{D} f))$
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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
- 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:(λx 2x3))  
  ...:(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))
```

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

```
(D (lambda (x) 2x3)) : (λx 6x2) ∪ (λx 12x3)
```

```
(D (lambda (x) 3x4)) : (λx 6x2) ∪ (λx 12x3)
```

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

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

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

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

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

```
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```

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

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```
(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.

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```
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```

```
(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

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(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

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

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

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

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

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

$$\bar{v} ::= \#t \mid \#f \mid () \mid \mathbb{R} \mid (\bar{v}_1, \bar{v}_2) \mid \langle \bar{\sigma}, e \rangle \mid \overline{\mathbb{R}}$$

$$\bar{\sigma} ::= \{x_1 \mapsto \bar{v}_1, \dots\}$$

Polyvariant Flow Analysis

with Unbounded Context Sensitivity

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

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

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

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

$$\bar{v} ::= \#t \mid \#f \mid () \mid \mathbb{R} \mid (\bar{v}_1, \bar{v}_2) \mid \langle \bar{\sigma}, e \rangle \mid \bar{\mathbb{R}}$$

$$\bar{\sigma} ::= \{x_1 \mapsto \bar{v}_1, \dots\}$$

Polyvariant Flow Analysis

with Unbounded Context Sensitivity

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

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

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

$$\bar{v} ::= \#t \mid \#f \mid () \mid \mathbb{R} \mid (\bar{v}_1, \bar{v}_2) \mid \langle \bar{\sigma}, e \rangle \mid \bar{\mathbb{R}}$$

$$\bar{\sigma} ::= \{x_1 \mapsto \bar{v}_1, \dots\}$$

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

Polyvariant Flow Analysis

with Unbounded Context Sensitivity

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

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

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

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

$$\bar{v} ::= \#t \mid \#f \mid () \mid \mathbb{R} \mid (\bar{v}_1, \bar{v}_2) \mid \langle \bar{\sigma}, e \rangle \mid \bar{\mathbb{R}}$$

$$\bar{\sigma} ::= \{x_1 \mapsto \bar{v}_1, \dots\}$$

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

Not suitable for arbitrary (i.e., typical SCHEME, ML, HASKELL, etc.) programs.

Polyvariant Flow Analysis

with Unbounded Context Sensitivity

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

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

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

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

$$\bar{v} ::= \#t \mid \#f \mid () \mid \mathbb{R} \mid (\bar{v}_1, \bar{v}_2) \mid \langle \bar{\sigma}, e \rangle \mid \bar{\mathbb{R}}$$

$$\bar{\sigma} ::= \{x_1 \mapsto \bar{v}_1, \dots\}$$

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

Not suitable for arbitrary (i.e., typical SCHEME, ML, HASKELL, etc.) programs.

Is suitable for FORTRAN-like programs.

Polyvariant Flow Analysis

with Unbounded Context Sensitivity

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

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

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

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

$$\bar{v} ::= \#t \mid \#f \mid () \mid \mathbb{R} \mid (\bar{v}_1, \bar{v}_2) \mid \langle \bar{\sigma}, e \rangle \mid \bar{\mathbb{R}}$$

$$\bar{\sigma} ::= \{x_1 \mapsto \bar{v}_1, \dots\}$$

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

Not suitable for arbitrary (i.e., typical SCHEME, ML, HASKELL, etc.) programs.

Is suitable for FORTRAN-like programs.

Necessary for migrating reflective source-code transformation to compile time.

Polyvariant Flow Analysis

with Unbounded Context Sensitivity

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No tags, tag checking, tag dispatching, indirect calls

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Necessary for migrating reflective source-code transformation to compile time.

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.

			B		
	b_1	\dots	b_j	\dots	b_n
a_1					
\vdots		\ddots	\vdots		
A	a_i	\dots	PAYOFF(a_i, b_j)	\dots	
\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.

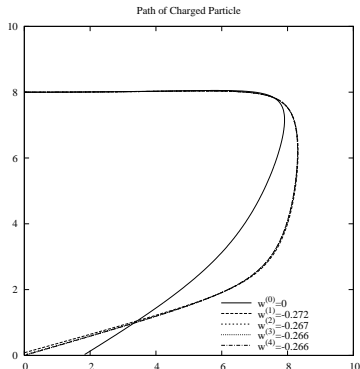
Game Theory

		\mathbb{R}^n		
		...	b	...
		<hr/>		
	\vdots	\ddots	\vdots	
\mathbb{R}^m	a	...	PAYOFF(a, b)	...
	\vdots		\vdots	\ddots

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

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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
STALIN ∇	1.00	1.00	1.00	1.00	1.00	1.00	1.00	1.00
ADIFOR	1.52	■	■	■	2.07	■	■	■
TAPENADE	3.40	■	■	■	2.56	■	■	■
FADBAD++	65.69	■	■	■	22.44	■	■	■
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
GHC	165.22	■	■	■	121.18	■	■	■
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
- not implemented in existing tool
- can't implement

Gradient-Based Optimization

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

Gradient-Based Optimization

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

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

Gradient-Based Optimization

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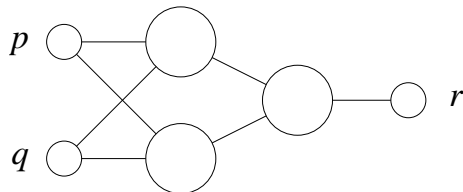
(define ((gradient f) x)
  (let ((n (length x)))
    (map (lambda (i) (tangent ((j* f) (bundle x (e i n))))
         (iota n))))

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

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

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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))))))
```

Neural Networks in VLAD

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

(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)))
```

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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)))

(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
STALIN ∇	1.00	■	1.00
ADIFOR	11.84	2.68	■
TAPENADE	11.35	4.33	6.24
ADIC	16.33	3.93	■
ADOL-C	12.34	3.89	35.53
CPPAD	42.15	■	23.69
FADBAD++	98.96	33.15	53.03
MLTON	73.94	■	37.94
OCAML	157.75	■	149.14
SML/NJ	142.71	■	94.97
GHC	■	■	■
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

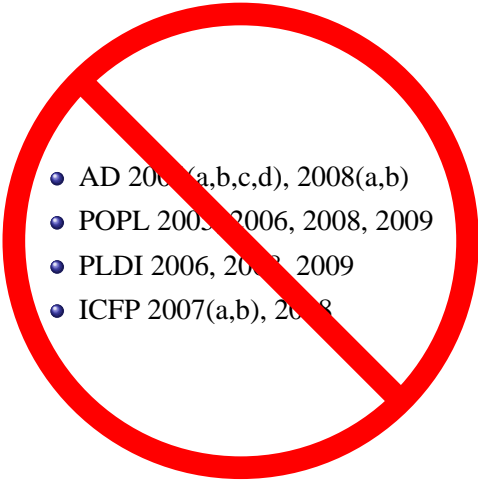
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Where You Can Read About This Work

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- AD 2004(a,b,c,d), 2008(a,b)
- POPL 2005, 2006, 2008, 2009
- PLDI 2006, 2008, 2009
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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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(¶4)

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Gottfried Leibniz
|
Jacob Bernoulli
|
Johann Bernoulli
|
Leonhard Euler
|
Joseph Louis Lagrange
|
Simeon Poisson
|
Michel Chasles
|
Hubert Anson Newton
|
Eliakim Hastings Moore
|
Oswald Veblen
|
Alonzo Church