

Automatic Differentiation for Probabilistic Programming

Jeffrey Mark Siskind
qobi@purdue.edu

School of Electrical and Computer Engineering
Purdue University

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

The Essence

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(define (f x) 2x3)
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`(define (f x) 2x3)` $\rightsquigarrow$ `(define (f' x) 6x2)`

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need **polyvariant** flow analysis

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

Arithmetic on Complex Numbers

$$a + bi$$

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.

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Arithmetic on Dual Numbers

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$$\varepsilon^2 = 0, \text{ but } \varepsilon \neq 0$$

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

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

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

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

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

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))
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(D f)
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Convenient

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Convenient but **slow**

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

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( $\mathcal{D}$  ( $\mathcal{D}$  f))
( $\mathcal{D}$  (lambda (x) ... ( $\mathcal{D}$  (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
```

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Fast

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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
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```

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```
function f(x)
double precision x, f
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end
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```
AD_TOP = f
AD_IVARS = x
AD_DVARS = f
```

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

Fast but **inconvenient**

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

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

Static Overloading: FADBAD++

```
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;}  
F<double> x;  
x.diff(0, 1);  
... f(x).d(0) ...
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Slow and **inconvenient**

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F<F<double> > x;  
x.diff(0, 1);  
x.diff(0, 1).diff(0,1);  
... f(x).d(0).d(0) ...
```

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

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

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Slow and **inconvenient**

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Slow and inconvenient

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Slow and inconvenient

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```

Slow and **inconvenient**

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) ...
```

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

```
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) ...
```

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

```
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) ...
```

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

Slow and **inconvenient**

Our API for Functional Forward AD

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

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 $\overline{\mathbb{R}^h}$

Our API for Functional Forward AD

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

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

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

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

Our API for Functional Forward AD

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

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

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

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

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

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

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

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

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

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

Our API for Functional Forward AD

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

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

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

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

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

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

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

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

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

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

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?

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

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

$\mathbf{j}^*$ 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 $\mathbf{j}^*$ as $\overrightarrow{\mathcal{J}}$

Our API for Functional Forward AD

$\text{bundle} : \tau \times \overrightarrow{\tau} \rightarrow \overrightarrow{\tau}$
 $\text{primal} : \overrightarrow{\tau} \rightarrow \tau$
 $\text{tangent} : \overrightarrow{\tau} \rightarrow \overrightarrow{\tau}$
 $\text{j}^* : (\tau_1 \rightarrow \tau_2) \rightarrow (\overrightarrow{\tau_1} \rightarrow \overrightarrow{\tau_2})$

```
(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?

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

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*` : $(\tau_1 \rightarrow \tau_2) \rightarrow (\overrightarrow{\tau_1} \rightarrow \overrightarrow{\tau_2})$

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(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 $\overrightarrow{\mathbb{R}^n}$

`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 $\tau \triangleright \overrightarrow{\tau}$ as $\overrightarrow{\tau}$

Sometimes write `j*` as $\mathcal{J}$

Convenient

Our API for Functional Forward AD

`bundle` : $\tau \times \overrightarrow{\tau} \rightarrow \overrightarrow{\tau}$

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

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?

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

Sometimes write `j*` as $\overrightarrow{\mathcal{J}}$

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

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

`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 $\tau \triangleright \overrightarrow{\tau}$ as $\overrightarrow{\tau}$

Sometimes write `j*` as `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)  
(D (D f))  
(D (lambda (x) ... (D (lambda (y) ...) ...) ...))
```

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

`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 $\tau \triangleright \overrightarrow{\tau}$ as $\overrightarrow{\tau}$

Sometimes write `j*` as $\overrightarrow{\mathcal{J}}$

What is `(j* 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*` : $(\tau_1 \rightarrow \tau_2) \rightarrow (\overrightarrow{\tau_1} \rightarrow \overrightarrow{\tau_2})$

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

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

Sometimes write `j*` as $\overrightarrow{\mathcal{J}}$

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

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

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

`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 $\tau \triangleright \overrightarrow{\tau}$ as $\overrightarrow{\tau}$

Sometimes write `j*` as $\overrightarrow{\mathcal{J}}$

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

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

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

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

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

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

A property

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

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

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

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

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

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

A property

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

A property

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

A property

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

A property

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

A property

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

A property

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

A property

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

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

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

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

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

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

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

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

0/1 matrix, every row has exactly one 1

A property

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

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

0/1 matrix, every row has exactly one 1

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

A property

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

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

0/1 matrix, every row has exactly one 1

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

A property

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

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

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when f is a rearrangement function

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

A property

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 x &: \mathbb{R}^n & \overline{x} &: \mathbb{R}^n & f &: \mathbb{R}^n \rightarrow \mathbb{R}^m \\
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 ((j^* f) x) &= (\text{bundle } (f (\text{primal } x)) (((\mathcal{J} f) (\text{primal } x)) (\text{tangent } x))) \\
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 \end{aligned}$$

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A property

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

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

0/1 matrix, every row has exactly one 1

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when f is a rearrangement function

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

What is the tangent of $\#t$?

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

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

when f is a rearrangement function

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

$f : (\#t \ x \ y) \mapsto (\#t \ x \ y) \quad \text{but} \quad f : (\#f \ x \ y) \mapsto (\#f \ y \ x)$

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f is a rearrangement function

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

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f is a rearrangement function

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

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

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f is a rearrangement function

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

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

What is the tangent of $\#t$?

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

when f is a rearrangement function

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

$f : (\#t \ x \ y) \mapsto (\#t \ x \ y) \quad \text{but} \quad f : (\#f \ x \ y) \mapsto (\#f \ y \ x)$

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 $(j^* \circ j^*)$?

when f is a rearrangement function

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

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

when f is a rearrangement function

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

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

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

when f is a rearrangement function

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

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

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

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

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

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

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

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

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

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

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

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

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

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

$$\vdots$$

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

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

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

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

$$\vdots$$

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

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

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

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

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

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

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

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

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

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

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

$$\vdots$$

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

A (Not So) Brief Tutorial on AD

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

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

A (Not So) Brief Tutorial on AD

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

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

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

A (Not So) Brief Tutorial on AD

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

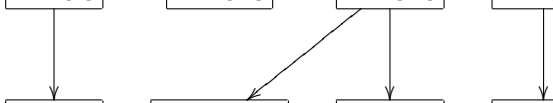
$$(\mathcal{J} f_n \mathbf{x}_{n-1}) \times \cdots \times (\mathcal{J} f_1 \mathbf{x}_0) = (\overline{f_1}' \mathbf{x}_0) \circ \cdots \circ (\overline{f_n}' \mathbf{x}_{n-1})$$

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

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

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

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

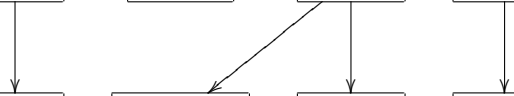
$$\mathbf{x}[L_i] := u_i \mathbf{x}[R_i]$$

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

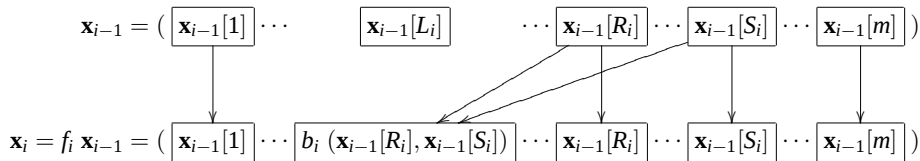
$$\mathbf{x}[L_i] := u_i \mathbf{x}[R_i]$$

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

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

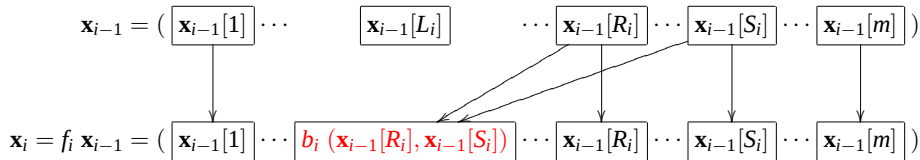
$$\mathbf{x}[L_i] := u_i \mathbf{x}[R_i]$$

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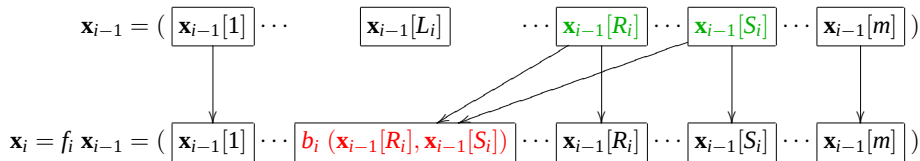
$$\mathbf{x}[L_i] := b(\mathbf{x}[R_i], \mathbf{x}[S_i])$$

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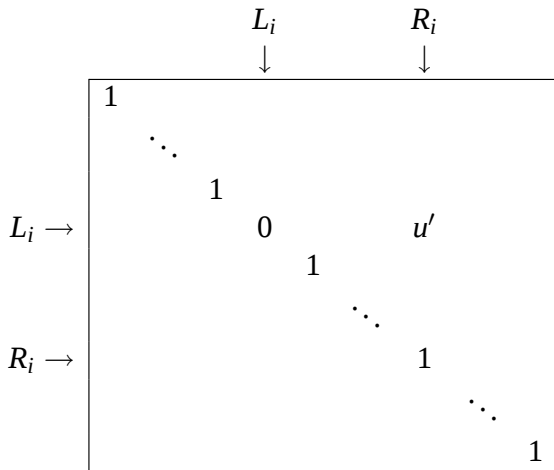
$$\mathbf{x}[L_i] := b(\mathbf{x}[R_i], \mathbf{x}[S_i])$$

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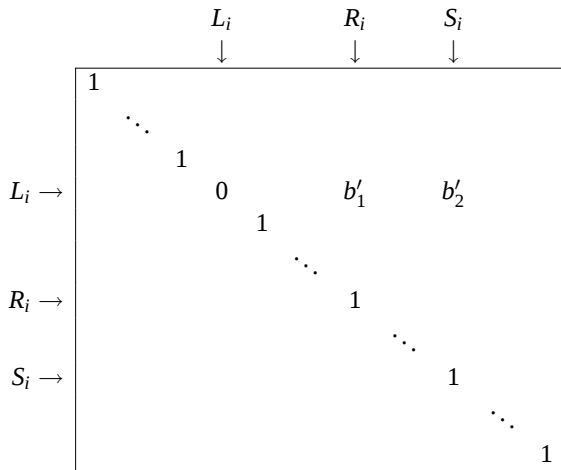
$$\mathbf{x}[L_i] := b(\mathbf{x}[R_i], \mathbf{x}[S_i])$$

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

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

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

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

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

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

$$\begin{pmatrix} \overline{\mathbf{x}}'_{i-1}[1] \\ \vdots \\ \overline{\mathbf{x}}'_{i-1}[L_i - 1] \\ u' \times \overline{\mathbf{x}}'_{i-1}[R_i] \\ \overline{\mathbf{x}}'_{i-1}[L_i + 1] \\ \vdots \\ \overline{\mathbf{x}}'_{i-1}[R_i] \\ \vdots \\ \overline{\mathbf{x}}'_{i-1}[m] \end{pmatrix} = \begin{pmatrix} 1 & & & & & & & \\ & \ddots & & & & & & \\ & & 1 & & & & & \\ & & & 0 & & u' & & \\ & & & & 1 & & & \\ & & & & & \ddots & & \\ & & & & & & 1 & \\ & & & & & & & \ddots \\ & & & & & & & & 1 \end{pmatrix} \begin{pmatrix} \overline{\mathbf{x}}'_{i-1}[1] \\ \vdots \\ \overline{\mathbf{x}}'_{i-1}[L_i - 1] \\ \overline{\mathbf{x}}'_{i-1}[L_i] \\ \overline{\mathbf{x}}'_{i-1}[L_i + 1] \\ \vdots \\ \overline{\mathbf{x}}'_{i-1}[R_i] \\ \vdots \\ \overline{\mathbf{x}}'_{i-1}[m] \end{pmatrix}$$

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

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

$$\begin{pmatrix} \overline{\mathbf{x}}'_{i-1}[1] \\ \vdots \\ \overline{\mathbf{x}}'_{i-1}[L_i - 1] \\ \textcolor{red}{u' \times \overline{\mathbf{x}}'_{i-1}[R_i]} \\ \overline{\mathbf{x}}'_{i-1}[L_i + 1] \\ \vdots \\ \overline{\mathbf{x}}'_{i-1}[R_i] \\ \vdots \\ \overline{\mathbf{x}}'_{i-1}[m] \end{pmatrix} = \begin{pmatrix} 1 & & & & & & & \\ & \ddots & & & & & & \\ & & 1 & & & & & \\ & & & 0 & & u' & & \\ & & & & 1 & & & \\ & & & & & \ddots & & \\ & & & & & & 1 & \\ & & & & & & & \ddots \\ & & & & & & & & 1 \end{pmatrix} \begin{pmatrix} \overline{\mathbf{x}}'_{i-1}[1] \\ \vdots \\ \overline{\mathbf{x}}'_{i-1}[L_i - 1] \\ \overline{\mathbf{x}}'_{i-1}[L_i] \\ \overline{\mathbf{x}}'_{i-1}[L_i + 1] \\ \vdots \\ \overline{\mathbf{x}}'_{i-1}[R_i] \\ \vdots \\ \overline{\mathbf{x}}'_{i-1}[m] \end{pmatrix}$$

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

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

$$\begin{pmatrix} \overline{\mathbf{x}}'_{i-1}[1] \\ \vdots \\ \overline{\mathbf{x}}'_{i-1}[L_i - 1] \\ \textcolor{red}{u' \times \overline{\mathbf{x}}'_{i-1}[R_i]} \\ \overline{\mathbf{x}}'_{i-1}[L_i + 1] \\ \vdots \\ \overline{\mathbf{x}}'_{i-1}[R_i] \\ \vdots \\ \overline{\mathbf{x}}'_{i-1}[m] \end{pmatrix} = \begin{pmatrix} 1 & & & & & & & \\ & \ddots & & & & & & \\ & & 1 & & & & & \\ & & & 0 & & u' & & \\ & & & & 1 & & & \\ & & & & & \ddots & & \\ & & & & & & 1 & \\ & & & & & & & \ddots \\ & & & & & & & & 1 \end{pmatrix} \begin{pmatrix} \overline{\mathbf{x}}'_{i-1}[1] \\ \vdots \\ \overline{\mathbf{x}}'_{i-1}[L_i - 1] \\ \overline{\mathbf{x}}'_{i-1}[L_i] \\ \overline{\mathbf{x}}'_{i-1}[L_i + 1] \\ \vdots \\ \textcolor{green}{\overline{\mathbf{x}}'_{i-1}[R_i]} \\ \vdots \\ \overline{\mathbf{x}}'_{i-1}[m] \end{pmatrix}$$

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

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

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

A (Not So) Brief Tutorial on AD

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

$$\begin{pmatrix} \overline{\mathbf{x}_{i-1}}[1] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[L_i - 1] \\ b'_1 \times \overline{\mathbf{x}_{i-1}}[R_i] + b'_2 \times \overline{\mathbf{x}_{i-1}}[S_i] \\ \overline{\mathbf{x}_{i-1}}[L_i + 1] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[R_i] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[S_i] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[m] \end{pmatrix} = \begin{pmatrix} 1 & & & & & \\ & \ddots & & & & \\ & & 1 & & & \\ & & & 0 & b'_1 & b'_2 \\ & & & & 1 & \\ & & & & & \ddots \\ & & & & 1 & \\ & & & & & \ddots \\ & & & & & & 1 & \\ & & & & & & & \ddots \\ & & & & & & & & 1 \end{pmatrix} \begin{pmatrix} \overline{\mathbf{x}_{i-1}}[1] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[L_i - 1] \\ \overline{\mathbf{x}_{i-1}}[L_i] \\ \overline{\mathbf{x}_{i-1}}[L_i + 1] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[R_i] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[S_i] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[m] \end{pmatrix}$$

$$\begin{aligned}b'_1 &= \mathcal{D}_1 b_i (\mathbf{x}_{i-1}[R_i], \mathbf{x}_{i-1}[S_i]) \\ b'_2 &= \mathcal{D}_2 b_i (\mathbf{x}_{i-1}[R_i], \mathbf{x}_{i-1}[S_i])\end{aligned}$$

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

$$\begin{pmatrix} \overline{\mathbf{x}_{i-1}}[1] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[L_i - 1] \\ \textcolor{red}{b'_1 \times \overline{\mathbf{x}_{i-1}}[R_i] + b'_2 \times \overline{\mathbf{x}_{i-1}}[S_i]} \\ \overline{\mathbf{x}_{i-1}}[L_i + 1] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[R_i] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[S_i] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[m] \end{pmatrix} = \begin{pmatrix} 1 & & & & & \\ & \ddots & & & & \\ & & 1 & & & \\ & & & 0 & b'_1 & b'_2 \\ & & & & 1 & \\ & & & & & \ddots \\ & & & & & & 1 \\ & & & & & & & \ddots \\ & & & & & & & & 1 \\ & & & & & & & & & \ddots \\ & & & & & & & & & & 1 \end{pmatrix} \begin{pmatrix} \overline{\mathbf{x}_{i-1}}[1] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[L_i - 1] \\ \overline{\mathbf{x}_{i-1}}[L_i] \\ \overline{\mathbf{x}_{i-1}}[L_i + 1] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[R_i] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[S_i] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[m] \end{pmatrix}$$

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

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

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

$$\begin{pmatrix} \overline{\mathbf{x}_{i-1}}[1] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[L_i - 1] \\ \textcolor{red}{b'_1 \times \overline{\mathbf{x}_{i-1}}[R_i] + b'_2 \times \overline{\mathbf{x}_{i-1}}[S_i]} \\ \overline{\mathbf{x}_{i-1}}[L_i + 1] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[R_i] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[S_i] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[m] \end{pmatrix} = \begin{pmatrix} 1 & & & & & \\ & \ddots & & & & \\ & & 1 & & & \\ & & & 0 & b'_1 & b'_2 \\ & & & & 1 & \\ & & & & & \ddots \\ & & & & & & 1 \\ & & & & & & & \ddots \\ & & & & & & & & 1 \\ & & & & & & & & & \ddots \\ & & & & & & & & & & 1 \end{pmatrix} \begin{pmatrix} \overline{\mathbf{x}_{i-1}}[1] \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[L_i - 1] \\ \overline{\mathbf{x}_{i-1}}[L_i] \\ \overline{\mathbf{x}_{i-1}}[L_i + 1] \\ \vdots \\ \textcolor{green}{\overline{\mathbf{x}_{i-1}}[R_i]} \\ \vdots \\ \textcolor{green}{\overline{\mathbf{x}_{i-1}}[S_i]} \\ \vdots \\ \overline{\mathbf{x}_{i-1}}[m] \end{pmatrix}$$

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

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

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

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

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

$$\begin{pmatrix} \overline{\mathbf{x}_i}[1] \\ \vdots \\ \overline{\mathbf{x}_i}[L_i - 1] \\ 0 \\ \overline{\mathbf{x}_i}[L_i + 1] \\ \vdots \\ u' \times \overline{\mathbf{x}_i}[L_i] + \overline{\mathbf{x}_i}[R_i] \\ \vdots \\ \overline{\mathbf{x}_i}[m] \end{pmatrix} = \begin{pmatrix} 1 & & & & & & & \\ & \ddots & & & & & & \\ & & 1 & & & & & \\ & & & 0 & & & & \\ & & & & 1 & & & \\ & & & & & \ddots & & \\ & & & u' & & & 1 & \\ & & & & & & & \ddots \\ & & & & & & & & 1 \end{pmatrix} \begin{pmatrix} \overline{\mathbf{x}_i}[1] \\ \vdots \\ \overline{\mathbf{x}_i}[L_i - 1] \\ \overline{\mathbf{x}_i}[L_i] \\ \overline{\mathbf{x}_i}[L_i + 1] \\ \vdots \\ \overline{\mathbf{x}_i}[R_i] \\ \vdots \\ \overline{\mathbf{x}_i}[m] \end{pmatrix}$$

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

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

$$\begin{pmatrix} \overline{\mathbf{x}_i}[1] \\ \vdots \\ \overline{\mathbf{x}_i}[L_i - 1] \\ \textcolor{red}{0} \\ \overline{\mathbf{x}_i}[L_i + 1] \\ \vdots \\ \textcolor{red}{u' \times \overline{\mathbf{x}_i}[L_i] + \overline{\mathbf{x}_i}[R_i]} \\ \vdots \\ \overline{\mathbf{x}_i}[m] \end{pmatrix} = \begin{pmatrix} 1 & & & & & & & \\ & \ddots & & & & & & \\ & & 1 & & & & & \\ & & & 0 & & & & \\ & & & & 1 & & & \\ & & & & & \ddots & & \\ & & & & u' & & 1 & \\ & & & & & & & \ddots \\ & & & & & & & & 1 \end{pmatrix} \begin{pmatrix} \overline{\mathbf{x}_i}[1] \\ \vdots \\ \overline{\mathbf{x}_i}[L_i - 1] \\ \overline{\mathbf{x}_i}[L_i] \\ \overline{\mathbf{x}_i}[L_i + 1] \\ \vdots \\ \overline{\mathbf{x}_i}[R_i] \\ \vdots \\ \overline{\mathbf{x}_i}[m] \end{pmatrix}$$

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

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

$$\begin{pmatrix} \overline{\mathbf{x}_i}[1] \\ \vdots \\ \overline{\mathbf{x}_i}[L_i - 1] \\ \textcolor{red}{0} \\ \overline{\mathbf{x}_i}[L_i + 1] \\ \vdots \\ \textcolor{red}{u' \times \overline{\mathbf{x}_i}[L_i] + \overline{\mathbf{x}_i}[R_i]} \\ \vdots \\ \overline{\mathbf{x}_i}[m] \end{pmatrix} = \begin{pmatrix} 1 & & & & & & & \\ & \ddots & & & & & & \\ & & 1 & & & & & \\ & & & 0 & & & & \\ & & & & 1 & & & \\ & & & & & \ddots & & \\ & & & & u' & & 1 & \\ & & & & & & & \ddots \\ & & & & & & & & 1 \end{pmatrix} \begin{pmatrix} \overline{\mathbf{x}_i}[1] \\ \vdots \\ \overline{\mathbf{x}_i}[L_i - 1] \\ \textcolor{green}{\overline{\mathbf{x}_i}[L_i]} \\ \overline{\mathbf{x}_i}[L_i + 1] \\ \vdots \\ \textcolor{green}{\overline{\mathbf{x}_i}[R_i]} \\ \vdots \\ \overline{\mathbf{x}_i}[m] \end{pmatrix}$$

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

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

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

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

$$\begin{pmatrix} \overline{\mathbf{x}_i}[1] \\ \vdots \\ \overline{\mathbf{x}_i}[L_i - 1] \\ 0 \\ \overline{\mathbf{x}_i}[L_i + 1] \\ \vdots \\ b'_1 \times \overline{\mathbf{x}_i}[L_i] + \overline{\mathbf{x}_i}[R_i] \\ \vdots \\ b'_2 \times \overline{\mathbf{x}_i}[L_i] + \overline{\mathbf{x}_i}[S_i] \\ \vdots \\ \overline{\mathbf{x}_i}[m] \end{pmatrix} = \begin{pmatrix} 1 & & & & & & & & \\ & \ddots & & & & & & & \\ & & 1 & & & & & & \\ & & & 0 & & & & & \\ & & & & 1 & & & & \\ & & & & & \ddots & & & \\ & & b'_1 & & & & 1 & & \\ & & & & & & & \ddots & \\ & & b'_2 & & & & & & 1 \\ & & & & & & & & \ddots & \\ & & & & & & & & & 1 \end{pmatrix} \begin{pmatrix} \overline{\mathbf{x}_i}[1] \\ \vdots \\ \overline{\mathbf{x}_i}[L_i - 1] \\ \overline{\mathbf{x}_i}[L_i] \\ \overline{\mathbf{x}_i}[L_i + 1] \\ \vdots \\ \overline{\mathbf{x}_i}[R_i] \\ \vdots \\ \overline{\mathbf{x}_i}[S_i] \\ \vdots \\ \overline{\mathbf{x}_i}[m] \end{pmatrix}$$

$$\begin{aligned}b'_1 &= \mathcal{D}_1 b_i (\mathbf{x}_{i-1}[R_i], \mathbf{x}_{i-1}[S_i]) \\ b'_2 &= \mathcal{D}_2 b_i (\mathbf{x}_{i-1}[R_i], \mathbf{x}_{i-1}[S_i])\end{aligned}$$

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

$$\begin{pmatrix} \overline{\mathbf{x}_i}[1] \\ \vdots \\ \overline{\mathbf{x}_i}[L_i - 1] \\ 0 \\ \overline{\mathbf{x}_i}[L_i + 1] \\ \vdots \\ b'_1 \times \overline{\mathbf{x}_i}[L_i] + \overline{\mathbf{x}_i}[R_i] \\ \vdots \\ b'_2 \times \overline{\mathbf{x}_i}[L_i] + \overline{\mathbf{x}_i}[S_i] \\ \vdots \\ \overline{\mathbf{x}_i}[m] \end{pmatrix} = \begin{pmatrix} 1 & & & & & & & & \\ & \ddots & & & & & & & \\ & & 1 & & & & & & \\ & & & 0 & & & & & \\ & & & & 1 & & & & \\ & & & & & \ddots & & & \\ & & b'_1 & & & & 1 & & \\ & & & & & & & \ddots & \\ & & b'_2 & & & & & & 1 \\ & & & & & & & & \ddots & \\ & & & & & & & & & 1 \end{pmatrix} \begin{pmatrix} \overline{\mathbf{x}_i}[1] \\ \vdots \\ \overline{\mathbf{x}_i}[L_i - 1] \\ \overline{\mathbf{x}_i}[L_i] \\ \overline{\mathbf{x}_i}[L_i + 1] \\ \vdots \\ \overline{\mathbf{x}_i}[R_i] \\ \vdots \\ \overline{\mathbf{x}_i}[S_i] \\ \vdots \\ \overline{\mathbf{x}_i}[m] \end{pmatrix}$$

$$\begin{aligned}b'_1 &= \mathcal{D}_1 b_i (\mathbf{x}_{i-1}[R_i], \mathbf{x}_{i-1}[S_i]) \\ b'_2 &= \mathcal{D}_2 b_i (\mathbf{x}_{i-1}[R_i], \mathbf{x}_{i-1}[S_i])\end{aligned}$$

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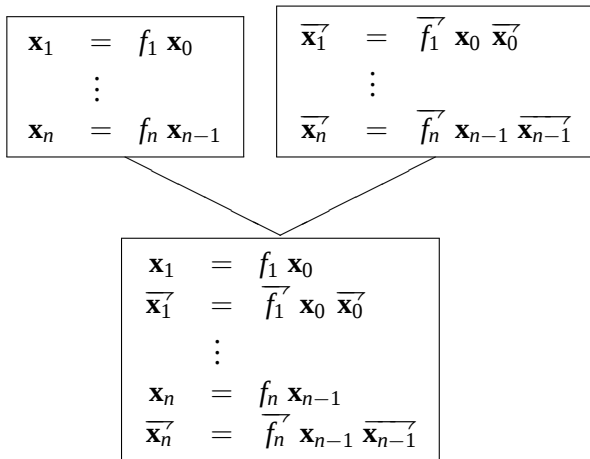
$$\begin{aligned}\overline{\mathbf{x}_{i-1}} &= \overline{f_i} \mathbf{x}_{i-1} \overline{\mathbf{x}_i} \\ &= (\mathcal{J} f_i \mathbf{x}_{i-1})^\top \times \overline{\mathbf{x}_i}\end{aligned}$$

$$\begin{pmatrix} \overline{\mathbf{x}_i}[1] \\ \vdots \\ \overline{\mathbf{x}_i}[L_i - 1] \\ 0 \\ \overline{\mathbf{x}_i}[L_i + 1] \\ \vdots \\ b'_1 \times \overline{\mathbf{x}_i}[L_i] + \overline{\mathbf{x}_i}[R_i] \\ \vdots \\ b'_2 \times \overline{\mathbf{x}_i}[L_i] + \overline{\mathbf{x}_i}[S_i] \\ \vdots \\ \overline{\mathbf{x}_i}[m] \end{pmatrix} = \begin{pmatrix} 1 & & & & & & & & \\ & \ddots & & & & & & & \\ & & 1 & & & & & & \\ & & & 0 & & & & & \\ & & & & 1 & & & & \\ & & & & & \ddots & & & \\ & & b'_1 & & & & 1 & & \\ & & & & & & & \ddots & \\ & b'_2 & & & & & & & 1 \\ & & & & & & & \ddots & \\ & & & & & & & & 1 \end{pmatrix} \begin{pmatrix} \overline{\mathbf{x}_i}[1] \\ \vdots \\ \overline{\mathbf{x}_i}[L_i - 1] \\ \overline{\mathbf{x}_i}[L_i] \\ \overline{\mathbf{x}_i}[L_i + 1] \\ \vdots \\ \overline{\mathbf{x}_i}[R_i] \\ \vdots \\ \overline{\mathbf{x}_i}[S_i] \\ \vdots \\ \overline{\mathbf{x}_i}[m] \end{pmatrix}$$

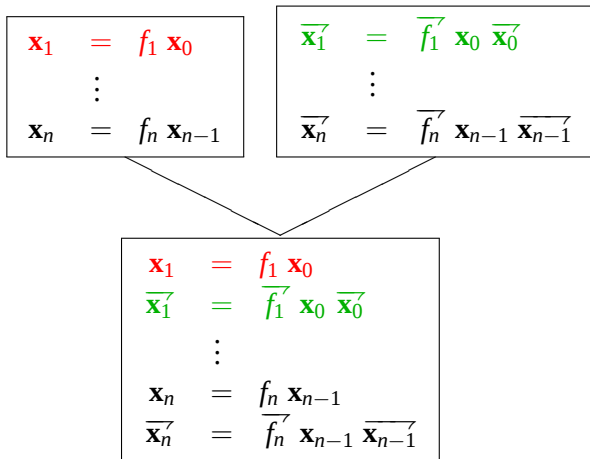
$$b'_1 = \mathcal{D}_1 b_i (\mathbf{x}_{i-1}[R_i], \mathbf{x}_{i-1}[S_i])$$

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

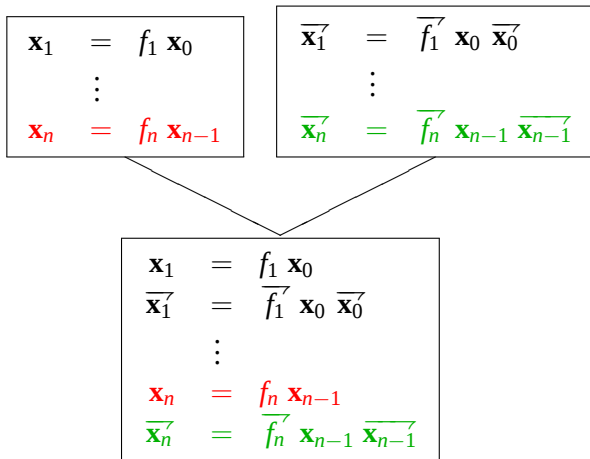
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$$\mathbf{x}_1 = f_1 \mathbf{x}_0$$

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

$$\vdots$$

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

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

$$(\mathbf{x}_1, \overline{\mathbf{x}}_1') = ((f_1 \mathbf{x}_0), (\overline{f_1'} \mathbf{x}_0 \overline{\mathbf{x}}_0'))$$

$$\vdots$$

$$(\mathbf{x}_n, \overline{\mathbf{x}}_n') = ((f_n \mathbf{x}_{n-1}), (\overline{f_n'} \mathbf{x}_{n-1} \overline{\mathbf{x}}_{n-1}'))$$

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

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

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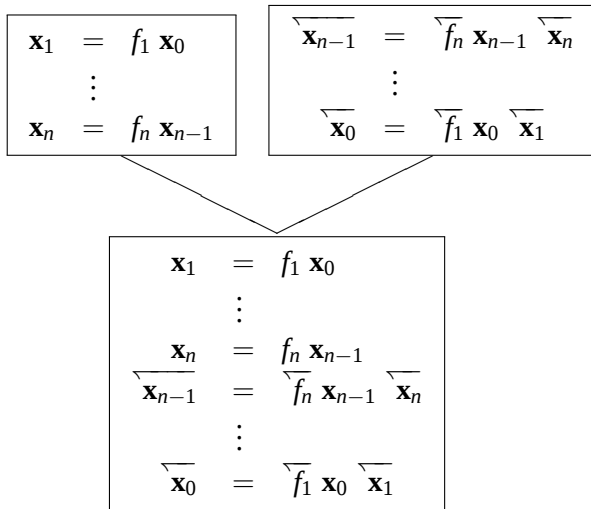
$$\begin{aligned}x_{L_i} &:= u_i \ x_{R_i} & \rightsquigarrow & \quad \overrightarrow{x_{L_i}} := \overrightarrow{u_i} \ \overrightarrow{x_{R_i}} \\x_{L_i} &:= b_i (x_{R_i}, x_{S_i}) & \rightsquigarrow & \quad \overrightarrow{x_{L_i}} := \overrightarrow{b_i} (\overrightarrow{x_{R_i}}, \overrightarrow{x_{S_i}})\end{aligned}$$

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

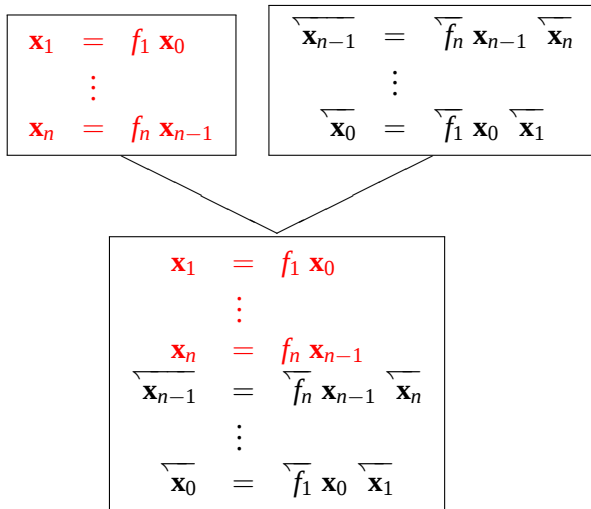
$$\overrightarrow{u} \ \overrightarrow{x} \equiv ((u \ x), ((\mathcal{D} \ u \ x) \times \overline{x}))$$

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

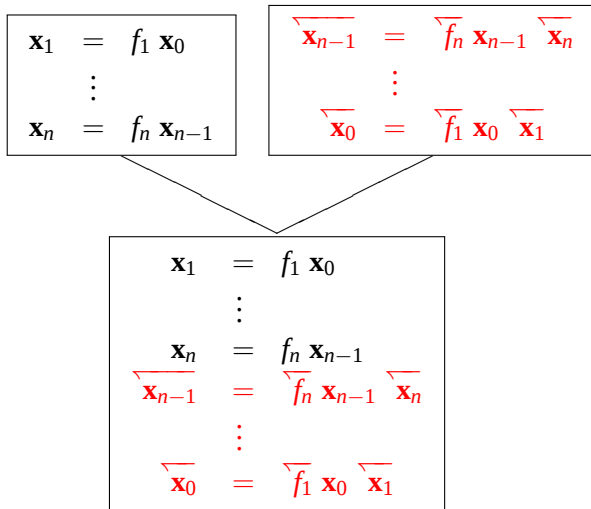
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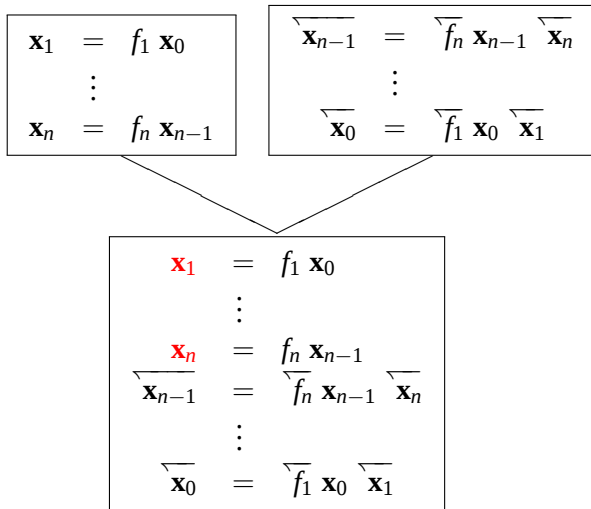
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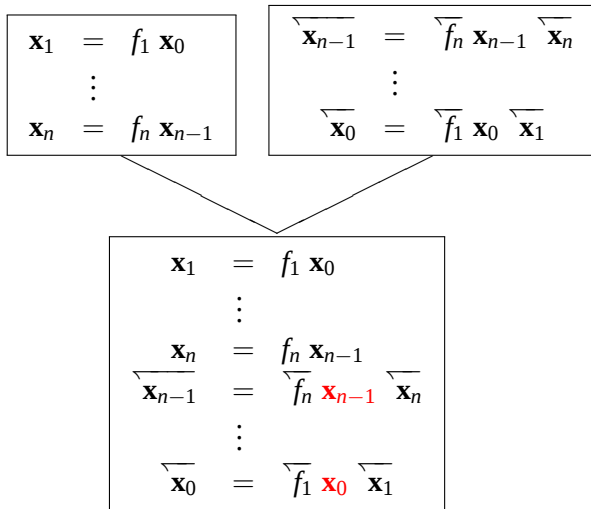
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$$\mathbf{x}_1 = f_1 \mathbf{x}_0$$

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

$$\vdots$$

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

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

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

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

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

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

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

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

$$\overline{x} \equiv x$$

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

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

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

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

The Functional Reverse-Mode Transformation

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

The Functional Reverse-Mode Transformation

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

The Functional Reverse-Mode Transformation

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

The Functional Reverse-Mode Transformation

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

The Functional Reverse-Mode Transformation

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

The Functional Reverse-Mode Transformation

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

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

The Functional Reverse-Mode Transformation

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

$$\left. \begin{array}{lcl} x_1 & = & u_1 \ x_{S_1} \\ \vdots & & \\ x_n & = & u_n \ x_{S_n} \end{array} \right\} \rightsquigarrow \left\{ \begin{array}{lcl} (x_1, \overleftarrow{x_1}) & = & \overleftarrow{u_1} \ x_{S_1} \\ \vdots & & \\ (x_n, \overleftarrow{x_n}) & = & \overleftarrow{u_n} \ x_{S_n} \\ \overleftarrow{x_0} & := & 0 \\ \vdots & & \\ \overleftarrow{x_{n-1}} & := & 0 \\ \overleftarrow{x_{S_n}} & +: = & \overleftarrow{x_n} \ \overleftarrow{x_n} \\ \vdots & & \\ \overleftarrow{x_{S_1}} & +: = & \overleftarrow{x_1} \ \overleftarrow{x_1} \end{array} \right.$$

The Functional Reverse-Mode Transformation

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

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

The Functional Reverse-Mode Transformation

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

The Functional Reverse-Mode Transformation

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

The Functional Reverse-Mode Transformation

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

The Functional Reverse-Mode Transformation

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

The Functional Reverse-Mode Transformation

$$\left. \begin{array}{lcl} x_1 & = & x_{R_1} x_{S_1} \\ \vdots & & \\ x_n & = & x_{R_n} x_{S_n} \end{array} \right\} \rightsquigarrow \left\{ \begin{array}{ll} (\overleftarrow{x_1}, \overline{x_1}) & = \quad \overleftarrow{x_{R_1}} \overleftarrow{x_{S_1}} \\ \vdots & \\ (\overleftarrow{x_n}, \overline{x_n}) & = \quad \overleftarrow{x_{R_n}} \overleftarrow{x_{S_n}} \\ \overline{x_0} & := \quad \mathbf{0} \, (\overleftarrow{\mathcal{J}}^{-1} \overleftarrow{x_0}) \\ \vdots & \\ \overline{x_{n-1}} & := \quad \mathbf{0} \, (\overleftarrow{\mathcal{J}}^{-1} \overleftarrow{x_{n-1}}) \\ \overline{x_{S_n}} & \oplus := \quad \overline{x_n} \overline{x_n} \\ \vdots & \\ \overline{x_{S_1}} & \oplus := \quad \overline{x_1} \overline{x_1} \end{array} \right.$$

The Functional Reverse-Mode Transformation

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

The Functional Reverse-Mode Transformation

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

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

$$\begin{aligned}\nabla f \mathbf{x} &\triangleq \frac{\partial f(\mathbf{x})}{\partial x_1}, \dots, \frac{\partial f(\mathbf{x})}{\partial x_n} \\ \text{GRADIENTDESCENT } f \mathbf{x}_0 &\triangleq \dots \mathbf{x}_{i+1} := \dots \nabla f \mathbf{x}_i \dots\end{aligned}$$

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

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

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

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

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

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

Modularity

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

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

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

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

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

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

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

Breaking Modularity

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

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

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

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

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

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Breaking Modularity

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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

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Breaking Modularity

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

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

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

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

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

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

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Breaking Modularity

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

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

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

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

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

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

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

Breaking Modularity

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

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

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

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

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

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

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

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

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Breaking Modularity

$$\nabla \overrightarrow{f} \mathbf{x} \triangleq (\overrightarrow{f} \mathbf{x} \triangleright \overrightarrow{e_1}), \dots, (\overrightarrow{f} \mathbf{x} \triangleright \overrightarrow{e_n})$$

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

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

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

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

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

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

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

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Breaking Modularity

$$\nabla \overrightarrow{f} \mathbf{x} \triangleq \dots \overleftarrow{f} \mathbf{x} \dots$$

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

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

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

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

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

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

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

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

Restoring Modularity

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

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

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

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

Restoring Modularity

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

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Restoring Modularity

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

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

Having your cake and eating it too

- Convenient

- Fast

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 - $\mathcal{D}$ formulated as a higher-order function in the language
- 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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 - $(\mathcal{D} (\mathcal{D} f))$

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

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- Fast

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

Having your cake and eating it too

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

- Fast

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

Monovariant Flow Analysis: 0-CFA

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

Monovariant Flow Analysis: 0-CFA

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

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

Monovariant Flow Analysis: 0-CFA

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

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

Monovariant Flow Analysis: 0-CFA

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

Monovariant Flow Analysis: 0-CFA

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

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

Monovariant Flow Analysis: 0-CFA

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

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

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

Monovariant Flow Analysis: 0-CFA

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

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

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

Monovariant Flow Analysis: 0-CFA

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

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

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

Monovariant Flow Analysis: 0-CFA

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

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

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

Monovariant Flow Analysis: 0-CFA

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

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

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

Monovariant Flow Analysis: 0-CFA

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

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

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

Monovariant Flow Analysis: 0-CFA

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

Monovariant Flow Analysis: 0-CFA

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

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

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

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

Monovariant Flow Analysis: 0-CFA

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

```
(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)  $2x^3$ )) ...)
```

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

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

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

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

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

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

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

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

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

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

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

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

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

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

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

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

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

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

```
(define (h ...) ... (D (lambda (x)  $3x^4$ )) ...)
```

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

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

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

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

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

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

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

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

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

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

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

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

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

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

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

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

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

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

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

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

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

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

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

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

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

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

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

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

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

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

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

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

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

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

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

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

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

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

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

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

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

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

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

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

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

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

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

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

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

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

Polyvariant Flow Analysis: k -CFA

with Bounded Context Sensitivity

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

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

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

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

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

⋮

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

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

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

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

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

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

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

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

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

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

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

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

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

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Side benefit: union-free

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No tags, tag checking, tag dispatching, indirect calls

Polyvariant Flow Analysis

with Unbounded Context Sensitivity

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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

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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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	...	b_j	... b_n
A	a_1				
	$\vdots$		$\ddots$	$\vdots$	
	a_i		...	PAYOFF(a_i, b_j)	...
	$\vdots$			$\vdots$	$\ddots$
	a_m				

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

Game Theory

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

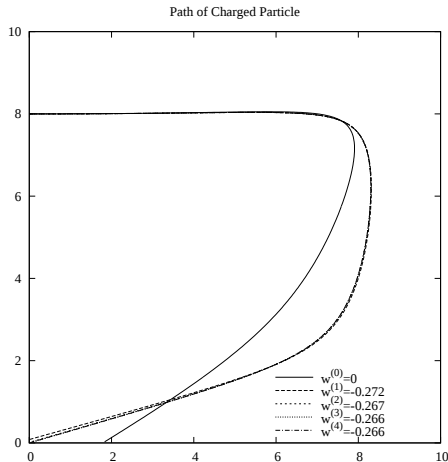
Game Theory

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

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

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

Cathode Ray Tubes



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

	saddle				particle			
	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
IKARUS	150.12	194.49	22.35	33.17	197.32	331.95	243.78	329.86
STALIN	162.20	202.23	26.15	36.54	222.87	380.89	212.39	269.96
SCHEME->C	298.53	312.39	41.52	54.33	394.65	583.13	424.48	537.91
CHICKEN	847.28	1088.64	124.56	203.88	1117.53	2030.66	1565.15	5291.80
BIGLOO	385.67	394.14	55.89	69.72	541.27	764.74	431.03	607.71
GAMBIT	349.43	404.47	49.70	71.16	435.53	748.82	532.35	712.63
LARCENY	537.53	543.88	66.47	85.35	696.14	1163.91	564.50	747.98
MzC	2234.68	2456.97	319.20	428.20	2765.03	3406.98	1311.92	1843.73
MzSCHEME	1874.02	2229.77	273.29	397.18	2314.17	4059.10	1651.69	2416.09
SCMUTILS	1141.06	1351.11	156.71	236.22	1428.85	2354.06	1163.29	1660.56
MLTON	42.42				69.48			
SML/NJ	63.20				84.59			
OCAML	79.65				121.60			
GHC	117.74				157.47			
FADBAD++	21.64				74.01			
ADIFOR	1.96				1.46			
TAPENADE	2.47				3.27			

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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      (cons (if (zero? i) 1.0 0.0)
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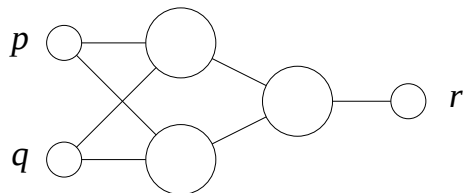
(define (gradient-ascent f x0 n eta)
  (if (zero? n)
      (list x0 (f x0) ((gradient f) x0))
      (gradient-ascent f
                        (zip (lambda (xi gi) (+ xi (* eta gi)))
                            x0
                            ((gradient f) x0))
                        (- n 1)
                        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)))
```

Neural Networks



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))
```

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))

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

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))

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

(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))))))

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

Neural Networks in VLAD

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

(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))
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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

	forward scalar	forward vector	reverse
STALIN ∇	1.39		1.00
FADBAD++	103.65	35.38	52.40
ADOL-C	12.80	4.07	32.55
CppAD	44.10		22.20
ADIC	17.50	4.13	
ADIFOR	12.38	2.79	
TAPENADE	11.86	4.54	5.80

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

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

Probabilistic Lambda Calculus

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

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

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

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

$$\Pr(x_1 \mapsto \mathbf{false}) = 1 - p_1$$

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

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

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

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

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

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

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

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Probabilistic Prolog

$p(\theta).$

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

$q(1).$

$q(2).$

Probabilistic Prolog

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

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

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

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

Probabilistic Prolog

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

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

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

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

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

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

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

Probabilistic Prolog

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

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

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

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

Probabilistic Lambda Calculus

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

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```

Probabilistic Lambda Calculus

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

Probabilistic Lambda Calculus

```
(define (evaluate expression environment)
  (cond
    ((constant-expression? expression)
     (singleton-tagged-distribution
      (constant-expression-value expression)))
    ((variable-access-expression? expression)
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Probabilistic Lambda Calculus

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

Probabilistic Lambda Calculus

(gradient-ascent

(lambda (p)

(let ((tagged-distribution

(evaluate **if** x_0 **then** 0 **else if** x_1 **then** 1 **else** 2

(list $\Pr(x_0 \mapsto \text{true}) = p_0$ $\Pr(x_0 \mapsto \text{false}) = 1 - p_0$

$\Pr(x_1 \mapsto \text{true}) = p_1$ $\Pr(x_1 \mapsto \text{false}) = 1 - p_1$

...))))

(map-reduce

*

1.0

(lambda (value)

(likelihood value tagged-distribution))

'(0 1 2 2))))

'(0.5 0.5)

1000.0

0.1)

Probabilistic Lambda Calculus

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```

Probabilistic Prolog

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

Probabilistic Prolog

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```

Probabilistic Prolog

```
(gradient-ascent
 (lambda (p)
  (let ((clauses (list Pr(p(0).) = p0
                        Pr(p(X):-q(X).) = 1 - p0
                        Pr(q(1).) = p1
                        Pr(q(2).) = 1 - p1))))
    (map-reduce
     *
     1.0
     (lambda (query)
      (likelihood (proof-distribution query clauses)))
      '(p(0) p(1) p(2) p(2)))))
 '(0.5 0.5)
1000.0
0.1)
```

Probabilistic Prolog

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(gradient-ascent
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    (map-reduce
     *
     1.0
     (lambda (query)
      (likelihood (proof-distribution query clauses)))
      '(p(0) p(1) p(2) p(2)))))
'(0.5 0.5)
1000.0
0.1)
```

Generated Code

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

Performance Comparison

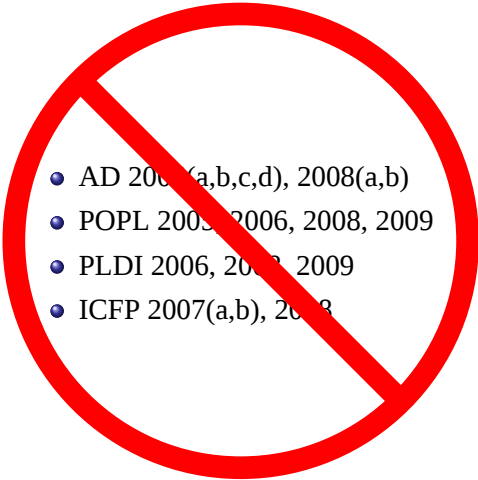
	probabilistic- lambda-calculus		probabilistic- prolog	
	forward	reverse	forward	reverse
STALIN ∇	1.00	1.00	1.00	1.00
IKARUS	499.13	419.37	10,384.61	4,347.82
SCHEME- $\rightarrow$ C	934.34	660.69	11,394.23	5,838.50
BIGLOO	1,367.10	967.60	14,531.25	7,701.86
GAMBIT	1,155.64	1,035.55	24,831.73	12,931.67
LARCENY	2,294.07	1,412.75	24,471.15	12,906.83
STALIN	1,313.00	1,925.84	22,524.03	14,633.54
CHICKEN	2,168.67	2,659.16	53,320.31	28,434.78
SCMUTILS	6,448.24	3,449.89	85,697.11	43,416.14
MzC	5,227.44	5,325.36	144,765.62	118,192.54
MzSCHEME	8,667.32	6,370.29	157,965.74	124,975.15

Where You Can Read About This Work

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

Where You Can Read About This Work

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

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