

Know what you are differentiating

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Know what you are differentiating

AD tools, like compilers, do not *know* your code.

⇒ therefore they apply general methods

The AD tool may fail on a code that hides its maths
(or on non-smooth math)

Knowing your code, the tool may apply a better method, ... sometimes the only one that works.

The **A** in **AD** could stand for “Assisted”

⇒ *You* assist the tool!

Codes that hide their maths

Implementation/discretization may hide mathematical meaning e.g.

► solving $f(x, y) = 0$ by bisection

► discrete special cases: `(a==1 ? b : a*b)`

Code not “chainrule-differentiable” needs rewriting.

Code can also be non-smooth:

```
rainfall = MAX(rainfall, 0)
```

Useful AD on this is an active research field, with little tool support.

When assistance is welcome

An AD tool cannot detect in general,
but can do a good job when indicated.

Examples:

- ▶ Black boxes
- ▶ Linear solvers
- ▶ Independent iterations
- ▶ Fixed-point iterations
- ▶ ...

Black boxes

Compiled libraries, non-smooth procedures. . .

→ no source, or source not differentiable

- ▶ replace black box with a dummy function
- ▶ differentiate code around black-box
- ▶ differentiate black-box mathematically
- ▶ replace AD-differentiated dummy with hand-written diff black-box

Loops with Independent Iterations

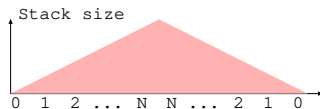
```
for (i=0 ; i<=N ; ++i) {  
    iteration i  
}
```

Suppose no loop-carried dependency
(except possible sum-reduction)

Loops with Independent Iterations

Plain reverse AD builds:

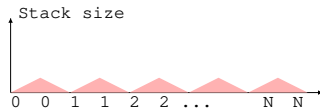
```
for (i=0 ; i<=N ; ++i) {  
    forward sweep of iteration i  
}  
for (i=N ; i>=0 ; --i) {  
    backward sweep of iteration i  
}
```



Loop #2 also has independent iterations

Reverse iterations, **fuse** with loop #1:

```
for (i=0 ; i<=N ; ++i) {  
    forward sweep of iteration i  
    backward sweep of iteration i  
}
```



⇒ **Reduces peak** stack size dramatically!

Linear Solvers (forward mode)

A perfect black box example:
either no source or poorly differentiable.

Go back to math instead:

$$Ax = b$$

Tangent AD:

$$\dot{A}x + A\dot{x} = \dot{b}$$

$$A\dot{x} = \dot{b} - \dot{A}x$$

Linear Solvers (forward mode)

Hand-written SOLVE_D:

```
SOLVE_D(A,Ad,x,xd,b,bd) {  
    SOLVE(A,x,b)  
    bdcopy = bd  
    DGEMV(-1,Ad,x,1,bdcopy)  
    SOLVE(A,xd,bdcopy)  
}
```

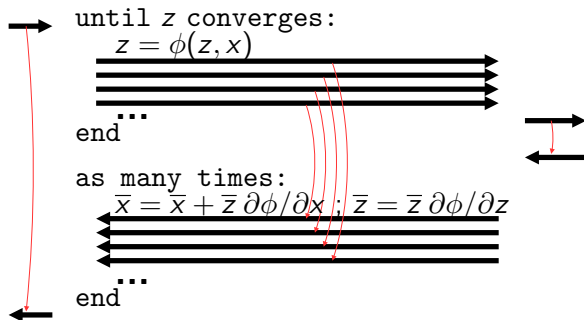
Linear Solvers (reverse mode)

(Similarly, with a little more effort...)

Hand-written SOLVE_B:

```
SOLVE_B(A, Ab, x, xb, b, bb) {  
    At = TRANSPOSE(A)  
    SOLVE(At, tmp, xb)  
    bb[:] = bb[:] + tmp[:]  
    SOLVE(A, x, b)  
    for each i and each j {  
        Ab[i,j] = Ab[i,j] - x[j]*tmp[i]  
    }  
    xb[:] = 0.0  
}
```

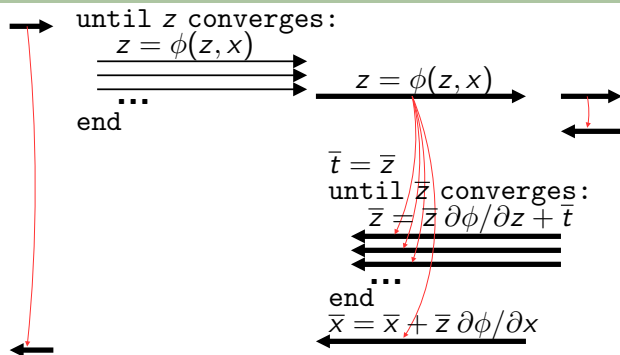
Fixed point iterations (reverse mode)



You should **not** do that!

- ▶ **all values** from intermediate iterations are **stored**
- ▶ **poor convergence** guarantees of the adjoint sweep

Fixed point Two-Phases Adjoint



- ▶ Only the **converged** primal iteration is stored, then is **used several times**.
- ▶ The adjoint iteration has its **own convergence control**
- ▶ Converges in **one step** if primal has quadratic convergence

Thank you