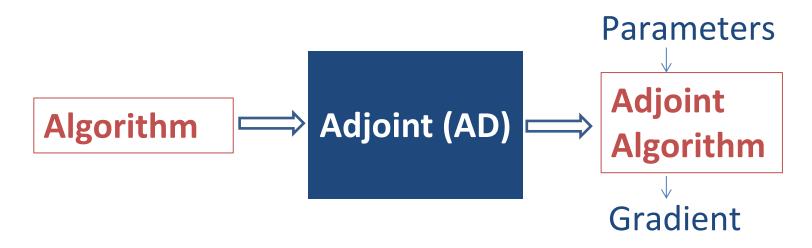
14th European Workshop on AD

ADJOINTS OF FIXED-POINT ITERATIONS Ala Taftaf

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Adjoint Algorithms



- Adjoint algorithms : the most efficient way to obtain gradient of a numerical simulation
- Computing the gradients has a cost (time, memory)
- One way around : take advantage of the structures of the given program (parallel loops, fixed point methods,...)

• Many implicit functions F(z, x) = 0 are computed with convergent iterations:

Initial guess $z_0(x)$

Iterate
$$z_{k+1}(x) = \varphi(z_k(x), x)$$

To fixed point $z_*(x) = arphi(z_*(x)$, x)

• Followed by an objective function $y = f(z_*, x)$

Adjoint of Fixed Point Iterations

• First iterations of a fixed-point search :

$$z_0(x)$$

$$z_1(x) = \varphi(z_0(x), x)$$

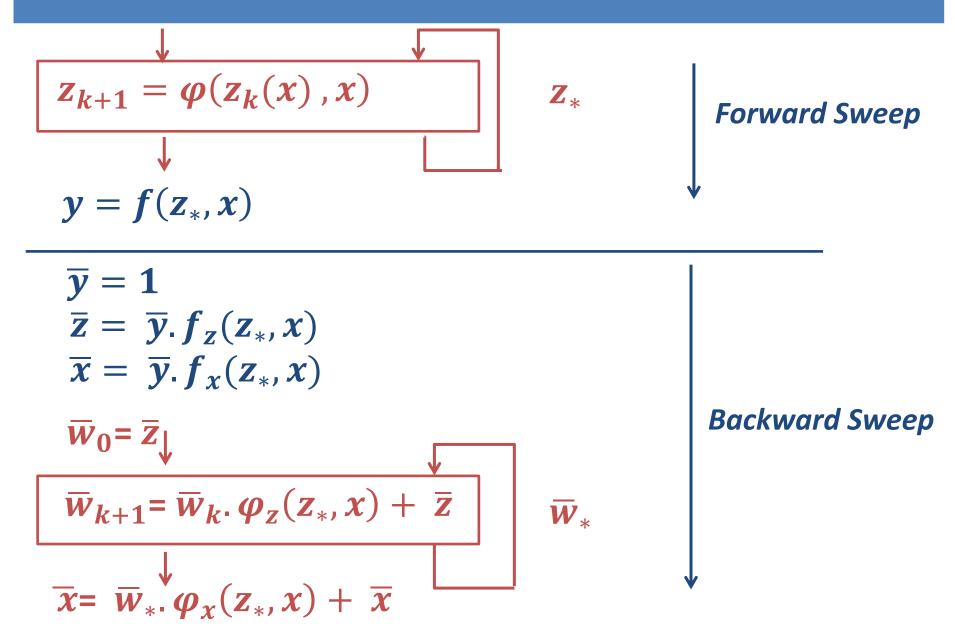
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operate on a meaningless state vector =>No adjoint
needed

- Adjoin only the (few) last iterations
- Save trajectory storage

At least two authors have studied mathematically fixed point iterations with the goal of defining an efficient adjoint : Christianson and Griewank

Christianson's strategy (BC)



Griewank's strategy (AG) : Piggyback

$$w_{k} = F(z_{k}, x)$$

$$z_{k+1} = z_{k} \cdot P_{k} \cdot w_{k}$$

$$x_{*}$$

$$y = f(z_{*}, x)$$

 $\mathbf{\Lambda}$

Original Program

$$\psi$$

$$w_{k} = F(z_{k}, x)$$

$$\overline{z}_{k} = \overline{w}_{k} \cdot F_{z}(z_{k}, x) + \overline{y} \cdot f_{z}(z_{k}, x)$$

$$z_{k+1} = z_{k} \cdot P_{k} \cdot w_{k}$$

$$\overline{w}_{k+1} = \overline{w}_{k} \cdot \overline{z}_{k} \cdot P_{k}$$

$$Z_{*}, \ \overline{w}_{*}$$

 $\overline{x}_* = \overline{w}_* \cdot F_x(z_*, x) + \overline{y} \cdot f_x(z_*, x)$

Adjoint Program

- Both manage to avoid naïve inversion of the original sequence of iterations => Save trajectory storage
- Stopping criterion of the adjoint fixed point is distinct from the original test
- Convergence rate is similar for the derivative computation and the original computation

AG and BC: Differences

	AG	BC
Shape of iteration step	Additional Assumptions $z_{k+1} = z_k - P_k \cdot w_k$	General $z_{k+1} = \varphi(z_k, x)$
Start of adjoining	From the Beginning (more iterations)	After a total convergence
Sequel of F.P. loop	Adjoin repeatedly	Adjoin once

Mostly for implementation reasons, we want :

- No assumption for iteration shape
- no multiple differentiation of the sequel
- differentiate only the (few) last iterations
- to preserve the 2 sweeps structure of the adjoint code

In the computation of the adjoint ($\overline{w}_{k+1} = \overline{w}_k$. $\varphi_z(z_*, x) + \overline{z}$) we need to determine (φ, z, x) in the original program => Adding new directives

> $z \quad \chi$ \$AD Begin FP loop /// **REPEAT UNTIL (***z* stationary) **\$AD Begin FP iteration** $\mathbf{z} = \boldsymbol{\varphi}(\mathbf{z}, \mathbf{x})$ **\$AD END FP iteration END REPEAT \$AD END FP loop** $\mathbf{y} = \mathbf{f}(\mathbf{z}, \mathbf{x})$



Is a differentiation strategy for the iterative computations really needed in practice, as yet more focused strategies exist ?

Have you samples of code where this strategy could be effective ?



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Thank you for your attention!