



# MANE 6960:

## Adjoint for Scientists and Engineers

Lecture 9

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# Computing Derivatives:

## Finite-Difference Approximations

# Forward-Difference Approximation

The class of **finite-difference approximations** use Taylor's theorem to construct difference formulae that approximate the derivative.

- The **forward-difference approximation** is the simplest and most commonly used finite-difference method in optimization

# Forward-Difference Approximation (cont.)

The forward-difference approximation is easy to derive.

- suppose we want the  $j^{\text{th}}$  partial derivative of  $f$ , that is  $\partial f / \partial x_j$
- apply Taylor's theorem at an arbitrary point  $x \in \mathbb{R}^n$  in the direction  $e_j$ .
- $e_j$  is the (unit) basis vector for the  $j^{\text{th}}$  variable, which picks out the partial derivative we want; it is given by

$$e_j \equiv [0 \quad 0 \quad \cdots \quad 0 \quad 1 \quad 0 \quad \cdots \quad 0]^T$$

# Forward-Difference Approximation (cont.)

We get

$$f(x + he_j) = f(x) + h \frac{\partial f}{\partial x_j} + \frac{h^2}{2} \frac{\partial^2 f}{\partial x_j^2}(x + \bar{h}e_j)$$

- $h > 0$  denotes the step size we have control of
- $\bar{h} \in (0, h)$  is required by Taylor's theorem

Rearranging the above expression we obtain the (first-order) forward-difference approximation.

# Forward-Difference Approximation (cont.)

## Definition: Forward-Difference Approximation

The forward-difference approximation of the partial derivative  $\partial f / \partial x_j$  of the function  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  is given by

$$\frac{f(x + he_j) - f(x)}{h} = \frac{\partial f}{\partial x_j} + \underbrace{L_{\text{FD}}h}_{\text{error}}$$

where  $h > 0$  is the step size,  $e_j$  is the  $j^{\text{th}}$  Cartesian basis vector, and  $L_{\text{FD}}$  is a constant that does not depend on  $h$ .

- Note that the error is proportional to  $h$ .

# Exercise

Compute the partial derivative  $\partial f / \partial x_1$  of

$$f(x) = \frac{3}{2}x_1^2 + x_2^2 - 2x_1x_2 + 2x_1^3 + \frac{1}{2}x_1^4$$

at  $x = (-1, -1)^T$  using the forward-difference approximation.

- Try step sizes  $h = 10^{-2}$ ,  $h = 10^{-7}$ , and  $h = 10^{-20}$
- Compare your estimates to the true derivative

# The Effect of Round-off Errors

The error in the forward-difference approximation,  $L_{\text{FD}}h$ , suggests that we make  $h$  small, but this failed in the previous exercise.

- This failure is not unique
- The problem is round-off error

Computers can only represent a finite number of digits of  $f(x)$  accurately.

- For IEEE double precision arithmetic, the relative error in approximating  $f(x)$  is  $\epsilon_{\text{mach}} \approx 10^{-16}$ .



# The Effect of Round-off Errors (cont.)

Let  $\tilde{f}(x)$  denote the computed value of  $f(x)$  (i.e. the exact value) and  $\tilde{f}(x + he_j)$  the computed value of  $f(x + he_j)$ . Then<sup>1</sup>

$$|\tilde{f}(x) - f(x)| \leq \epsilon_{\text{mach}} L_f$$

$$\text{and} \quad |\tilde{f}(x + he_j) - f(x + he_j)| \leq \epsilon_{\text{mach}} L_f$$

where  $|f(x)| \leq L_f$  for all  $x$  in the region of interest.

If we substitute these bounds into the forward-difference formula, we get the following error estimate:

$$\text{Error}_{\text{FD}} \approx L_{\text{FD}} h + \frac{2\epsilon_{\text{mach}} L_f}{h}$$

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<sup>1</sup>This analysis is based on [NW06, pg. 196]

# The Effect of Round-off Errors (cont.)

We want the value of  $h$  that minimizes this error; we know how to do that!

Taking the derivative of our estimate of  $\text{Error}_{\text{FD}}$  with respect to  $h$  and setting the result to zero, we find

$$h^* = \sqrt{\frac{2L_f \epsilon_{\text{mach}}}{L_{\text{FD}}}} \approx \sqrt{\epsilon_{\text{mach}}}$$

- For double precision floating point numbers,  $h^* \approx 10^{-8}$ .

# The Effect of Round-off Errors (cont.)

Thus, the error in the finite-difference approximation cannot be made zero.

This can be partially ameliorated by using a more accurate finite-difference approximation. . .

# Central-Difference Approximation

## Definition: Central-Difference Approximation

The central-difference approximation of the partial derivative  $\partial f / \partial x_j$  of the function  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  is given by

$$\frac{f(x + he_j) - f(x - he_j)}{2h} = \frac{\partial f}{\partial x_j} + \underbrace{L_{\text{CD}} h^2}_{\text{error}}$$

where  $h > 0$  is the step size,  $e_j$  is the  $j^{\text{th}}$  Cartesian basis vector, and  $L_{\text{CD}}$  is a constant that does not depend on  $h$ .

- The optimal step size in this case is  $h^* \approx \sqrt[3]{\epsilon_{\text{mach}}}$ .

# Computational Cost

In addition to accuracy issues, finite-difference methods have a significant computational cost: to evaluate  $\nabla f$  when  $x \in \mathbb{R}^n$ ,

- the forward-difference approximation requires  $n$  function evaluations, not including the unperturbed value  $f(x)$ ;
- the central-difference approximation requires  $2n$  function evaluations.

# Pros & Cons of Finite-Difference Approximations

- ✓ easy to implement
- ✓ can be applied to almost any “black-box” function
- ✗ accuracy can be an issue, especially for badly scaled problems; also, choosing  $h^*$  may be difficult for multiple inputs
- ✗ computational cost scales with the  $\#$  of inputs

# Computing Derivatives: Complex-Step Approximation

# Refresher: Complex/Imaginary Numbers

Before we start, here's a (very) brief refresher on complex numbers

- $i = \sqrt{-1}$ .
- A general complex number can be written as  $z = x + iy$ , where  $x, y \in \mathbb{R}$ .
- A complex number can be thought of as a point in the plane.



# The Complex-Step Method

If  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  is a function of real variables, and it is complex differentiable — which implies the function is complex analytic — on the domain of interest, then

$$\begin{aligned} f(x + ihe_j) &= f(x) + ih \frac{\partial f}{\partial x_j} + \frac{i^2 h^2}{2} \frac{\partial^2 f}{\partial x_j^2} + \frac{i^3 h^3}{6} \frac{\partial^3 f}{\partial x_j^3} + \dots \\ &= f(x) + ih \frac{\partial f}{\partial x_j} - \frac{h^2}{2} \frac{\partial^2 f}{\partial x_j^2} - \frac{ih^3}{6} \frac{\partial^3 f}{\partial x_j^3} + \dots \end{aligned}$$

Taking the imaginary part of this equation, and rearranging gives...

# The Complex-Step Method (cont.)

## Definition: Complex-Step Approximation [ST98]

The complex-step approximation of the partial derivative  $\partial f / \partial x_j$  of the function  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  is given by

$$\frac{\Im[f(x + ihe_j)]}{h} = \frac{\partial f}{\partial x_j} + \underbrace{L_{\text{CS}}h^2}_{\text{error}}$$

where  $h > 0$  is the step size,  $e_j$  is the  $j^{\text{th}}$  Cartesian basis vector, and  $L_{\text{CS}}$  is a constant that does not depend on  $h$ .

- $\Im$  means “get the coefficient multiplying  $i$ ”

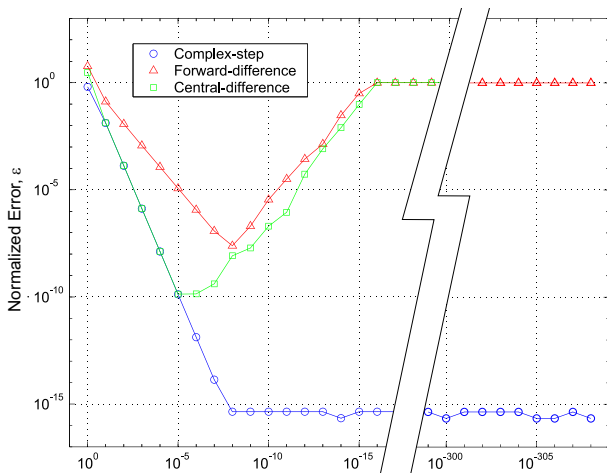
# The Complex-Step Method (cont.)

This is cool for the following reason:

- the complex-step formula does not involve differences of function values;
- therefore, there is no subtractive cancellation due to round-off;
- therefore, **we can make  $h$  as small as we want and make the error disappear!**

OK, there is a limit to how small we can make  $h$ , because there is a smallest exponent allowed on computers, namely  $h > 2.22 \times 10^{-308}$ .

# The Complex-Step Method (cont.)



Relative error in the derivative vs. decreasing step size

# Complex-Step Pitfalls

While the complex-step method is straightforward to apply, there are a couple things to look out for [MSA03]:

- need to redefine `min`, `max`, and `abs`
- some trig and inverse trig functions may need to be redefined
- in Matlab and Julia, use the `transpose` function or the `.'` operator for transposing vectors and matrices, otherwise you get the conjugate transpose.

Let's look at the `abs` function as an example.

## Complex-Step Pitfalls (cont.)

In many programming languages, `abs` is defined as the modulus of the complex number, i.e.  $|z| = \sqrt{x^2 + y^2}$ .

- This is not what we want.

Instead, we can define a new function:

```
1  function cabs(z)
2      % complexified version of the absolute value
3      if real(z) >= 0
4          return z
5      else
6          return -z
7      end
8  end
```

# Pros & Cons of the Complex-Step Approximation

- ✓ easy to implement
- ✓ can be applied to almost any “black-box” function that accepts complex variables
- ✓ many open-source codes can be easily adapted to use complex-step.
- ✗ computational cost still scales with the # of input variables

# Exercise

Compute the partial derivative  $\partial f / \partial x_1$  of

$$f(x) = \frac{3}{2}x_1^2 + x_2^2 - 2x_1x_2 + 2x_1^3 + \frac{1}{2}x_1^4$$

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- Try step sizes  $h = 10^{-2}$ ,  $h = 10^{-7}$ , and  $h = 10^{-20}$
- Compare your estimates to the true derivative



# References

- [MSA03] Joaquim R. R. A. Martins, Peter Sturdza, and Juan J. Alonso, *The complex-step derivative approximation*, ACM Transactions on Mathematical Software **29** (2003), no. 3, 245–262.
- [NW06] J. Nocedal and S. J. Wright, *Numerical Optimization*, second ed., Springer–Verlag, Berlin, Germany, 2006.
- [ST98] William Squire and George Trapp, *Using complex variables to estimate derivatives of real functions*, SIAM Review **40** (1998), no. 1, 110–112.