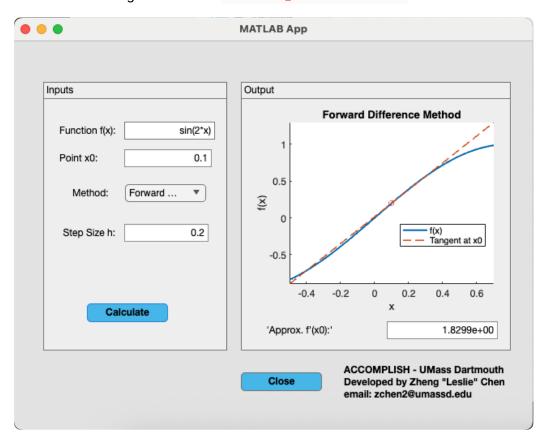
## **ACCOMPLISH - Numerical Differentiation**

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### Goal

In this module, students will finish building a MATLAB App Designer application that approximates derivatives of arbitrary functions. Starting from the provided template ( Numerical\_Differentiation\_App.mlapp ), they will implement all finite-difference algorithms in the numerical\_differentiation.m file.



### Overview of the GUI

#### 1. User inputs

- Function: Enter the formula, for example  $f(x) = \sin(x)$
- Evaluation Point: Specify the value of  $x_0$
- Method: Select a finite-difference formula from the drop-down menu

#### 2. App framework

- Starter File: Numerical\_Differentiation\_App.mlapp
- Algorithm Library: All differentiation routines are implemented as separate functions in numerical\_differentiation.m
- Instructor Version: A complete app (with working code) is provided to the advisor only

#### 3. Student assignment

- Review the mathematical background for each finite-difference formula
- Fill in the missing code blocks in numerical\_differentiation.m
- Detailed guidance and derivations will be provided in the sections that follow

#### **MATLAB GUI Resources**

- App Designer YouTube tutorial (recommended) https://youtu.be/nb0jHVXKY2w
- App Designer documentation

https://www.mathworks.com/help/matlab/app-designer.html

- Create and Run a Simple App Using App Designer tutorial
   https://www.mathworks.com/help/matlab/creating\_guis/create-a-simple-app-or-gui-using-app-designer.html
- MATLAB App Designer product page
   https://www.mathworks.com/products/matlab/app-designer.html

### **Included Differentiation Methods**

Students will implement these finite-difference formulas:

- Two-point formulas
  - Forward-difference
  - Backward-difference
- Three-point formulas
- Five-point formulas
- Second-derivative midpoint formula (optional)

Next, we will provide mathematical derivations and coding hints for each method so you can seamlessly integrate them into the GUI.

# **Table of Contents**

To approximate  $f'(x_0)$ :

- Two-point formulas
- Three-point formulas
- Five-point formulas

To approximate  $f''(x_0)$  (optional):

· Second-derivative midpoint formula

# Two-point\_formulas

#### **Derivation**

• Forward-difference: From the Taylor expansion of  $f(x_0 + h)$ , we have

$$f(x_0+h)=f(x_0)+h\,f'(x_0)+O(h^2)$$
.

Rearranging gives

$$f'(x_0) \; pprox \; rac{f(x_0+h)-f(x_0)}{h} + O(h) \, ,$$

a first-order accurate formula.

• Backward-difference: Similarly, expanding  $f(x_0 - h)$  yields

$$f(x_0 - h) = f(x_0) - h f'(x_0) + O(h^2),$$

so

$$f'(x_0) \; pprox \; rac{f(x_0) - f(x_0 - h)}{h} + O(h) \, ,$$

also first-order accurate.

## **Coding hints**

- Inside your Calculate\_pushbutton\_Callback, branch on handles.Method to distinguish forward vs. backward.
- For forward, you will need f(x0 + h) and f(x0).
- For backward, you will need f(x0) and f(x0 h).
- Compute the difference of those two values and then divide by h.

## Three-point\_formulas

#### **Derivation**

• Central-difference: Using Taylor series for  $f(x_0 \pm h)$  and subtracting gives

$$f'(x_0) \ pprox \ rac{f(x_0+h)-f(x_0-h)}{2h} + O(h^2) \, ,$$

a second-order accurate midpoint formula.

• Endpoint (forward/backward): At boundaries, the three-point endpoint formula is

$$f'(x_0) \, pprox \, rac{-3f(x_0) + 4f(x_0 + h) - f(x_0 + 2h)}{2h} + O(h^2)$$

(and its backward analogue).

# **Coding hints**

- Branch on the three sub-methods (central, forward-endpoint, backward-endpoint).
- Central uses f(x0 + h) and f(x0 h).
- Endpoint uses three sample points: at  $\times 0$ , at one step away ( $\pm h$ ), and at two steps away ( $\pm 2 \times h$ ).
- Combine those three function values with the weights from the derivation, then divide by 2\*h.

## Five-point\_formulas

#### **Derivation**

The fourth-order accurate central five-point formula comes from combining Taylor expansions at  $x_0 \pm h$  and  $x_0 \pm h$  to cancel lower-order error terms:

$$f'(x_0) \, pprox \, rac{f(x_0-2h) \, - \, 8 \, f(x_0-h) \, + \, 8 \, f(x_0+h) \, - \, f(x_0+2h)}{12 \, h} \, + \, O(h^4).$$

## **Coding hints**

- This central formula requires four off-grid evaluations:  $x0 \pm h$  and  $x0 \pm 2*h$ , plus x0 itself (if you verify consistency).
- Look back at the derivation to pick the four coefficients  $\{1, -8, 8, -1\}$ .
- Form the weighted sum of those four f(...) calls and divide by 12\*h.

# Second-derivative\_midpoint\_formula

### **Derivation**

• From the Taylor expansions at  $x_0 \pm h$ , the second-derivative midpoint formula is

$$f''(x_0) \ pprox \ rac{f(x_0+h) \ - \ 2 \, f(x_0) \ + \ f(x_0-h)}{h^2} \ + \ O(h^2).$$

## **Coding hints**

- Evaluate the function at  $x_0 + h$ ,  $x_0$ , and  $x_0 h$  to obtain the three sample values.
- Multiply these values by the weights  $\{1, -2, 1\}$  to form the numerator of the approximation.
- Divide the weighted sum by  $h^2$  to compute the second-derivative estimate.
- Store the result in its own variable (for example, secondDerivative) to keep it separate from any firstderivative outputs.