

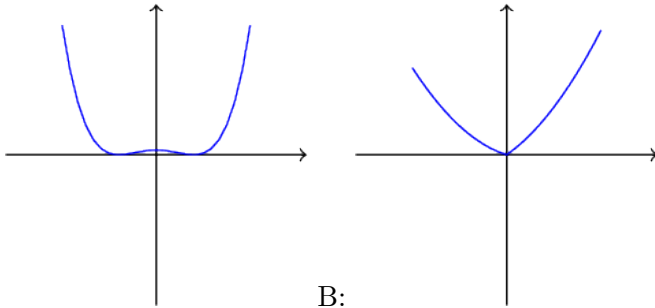
Worksheet

Part 1. Objectives

- Be able to find a minimum of a function in 1D using Golden Section Search and Newton
- Be able to find a minimum of a function in n dimensions using Steepest Descent and Newton
- Know the convergence rates of these methods and their behavior on different types of functions
- Be able to apply Lagrange multipliers for constrained optimization

Part 2. Unimodality

Which of these functions is unimodal?



- (A) Function A
(B) Function B
(C) Function A and B
(D) Neither

Part 3. Entries of the Hessian matrix

Consider the function

$$f(x, y, z) = x^2 + 3xy + 5z^2.$$

What is the entry $H_{2,1}$ in the second row and first column of the Hessian matrix of f ?

Part 4. Golden Section Search

Complete the code doing Golden Section Search for function minimization below. Stop the iteration when your brackets are less than 10^{-5} wide.

The supplied plotting code shows the evolution of your brackets. **Observe** how one of the bracket midpoints stays the same from one iteration to the next.

INPUT:

- f , a function to minimize. *Note:* f is not unimodal.
- a and b , the left and right ends of the starting bracket.

OUTPUT:

- a , b , the final ends of your bracket.

```
import numpy as np
brackets = []

while ....
    gs = (np.sqrt(5)-1)/2
    m2 = a + gs*(b-a)
    m1 = a + (1-gs)*(b-a)

    brackets.append([a, m1, m2, b])

    ...

# plotting code below, no need to modify
import matplotlib.pyplot as plt
x = np.linspace(-10, 10)
plt.plot(x, f(x))

brackets = np.array(brackets)
for i in range(4):
    plt.plot(brackets[:, i], 3*np.arange(len(brackets)), "o-")
```