## Labwork 1: Gradient Descent

## April 29, 2025

### 1 Introduction

Gradient descent: definition

Gradient descent is an optimization algorithm used to minimize a function by iteratively moving in the direction of the steepest descent, as defined by the negative of the gradient.

# 2 Implementation

The implementation of gradient descent in this lab minimizes the function f(x). The algorithm starts with an initial value of x, a learning rate r, and a convergence threshold  $\epsilon$ . It iteratively updates x using the formula  $x = x - r \cdot f'(x)$ , where f'(x) is the gradient of f(x). The process continues until f(x) becomes less than or equal to  $\epsilon$ , at which point the algorithm returns the final values of x and f(x).

## 3 Evaluation

Test with  $y = x^4$ Output of my program:

Enter initial x: 2

Enter learning rate r: 0.05

Enter convergence threshold epsilon: 0.005

The x and f(x) steps: Current x: 0.3999999999999, f(x): 0.02559999999999977

```
Current x: 0.3302717824378793, f(x): 0.011898326472229414

Current x: 0.3230666095439659, f(x): 0.010893521523512159

Current x: 0.3163227856997814, f(x): 0.010012024563155282

Current x: 0.30999252748521344, f(x): 0.009234319577443924

Current x: 0.3040347583400302, f(x): 0.008544623793691794

Current x: 0.2984139379836053, f(x): 0.007930058940763027

Current x: 0.2930991333237783, f(x): 0.00738003016633253

Current x: 0.28806327389789593, f(x): 0.0068857550438280605

Current x: 0.2832825498919133, f(x): 0.006439902494252792

Current x: 0.27873592144853765, f(x): 0.006036313180470917

Current x: 0.27440471566128116, f(x): 0.005669780969569371

Current x: 0.27027229326037167, f(x): 0.005335880645204224

Current x: 0.2663237711378589, f(x): 0.005303830985947597

Current x: 0.2625457899501838, f(x): 0.0047513851304683925

Final x: 0.2625457899501838, Final f(x): 0.0047513851304683925
```

How ever, for the learning rate of 0.2, the program gives: OverflowError: (34, 'Numerical result out of range')

The reason for this is that the learning rate is too high, causing the algorithm to overshoot the minimum and produce an overflow error.

### 4 Conclusion

In this labwork, I implemented the gradient descent algorithm to minimize the function  $f(x) = x^4$ . The result is that the algorithm converged to a minimum value of f(x) at  $x \approx 0.4$  with a final function value of approximately 0.0256.

An interesting finding is that the choice of learning rate significantly affects the convergence speed and stability of the algorithm. A smaller learning rate leads to slower convergence, while a larger one may cause divergence.