Sihong HE, #ID: 89483859

I use my own optimizer from HW1 on HW2. There are 9 functions on my HW2:

- 1. optimizer1D() which from my HW1 aim to do linear search.
- 2. conjugate gradients() which is base on CG algorithm.
- 3.  $\frac{\text{func1}}{\text{func1}} (1 x_0)^2 + 100(x_0 x_1^2)^2$  and its gradient function funcg1.
- 4.  $\frac{\text{func2}}{(x_0 1)^2 + (x_1 2x_0)^2 + (x_2 x_1)^2 + (x_3 x_2)^2 + (2x_4 x_3)^2 + (x_5 x_4)^2 + (x_6 x_6)^2 + (x_7 2x_6)^2 + (x_8 x_7)^2 + (2x_9 x_8)^2}{\text{funcg2}}$  func1 funcg1 func2 funcg2 are used to test conjugate\_gradients().
- 5. alternate\_method() which base on steepest descent algorithm.
- 6. eval() and eval\_alt() to print the average error of the output argument vector, function value produced by the algorithm, total number of CG iterations along with statistical error bars on those quantities

My start points are sampled from uniform distribution U (-1,1). The true mathematical optimums for two test functions are [1,1] and [1,2,2,2,1,1,1,2,2,1] respectively. For the stopping criterion, I choose x tolerance = 1e–6 and function tolerance = 1e–8,  $\epsilon_{\rm G}=1e-9$ ,  $\epsilon_{\rm A}=1e-9$ ,  $\epsilon_{\rm R}=1e-9$ . The machine precision is 2.22e–16. I set a seed(123).

## Problem 1

Function	Average error	Function value	# of CG iterations
Func1	3.54e-3+-4.56e-3	6.73e-6+-1.96e-5	33.1+-15.13
Func2	1.22+-1.33	0.106+-0.136	45.55+-27.68

## Problem 2

Function	Average error	Function value	# of iterations
Func1	2.22e-2+-2.659e-2	2.47e-4+-5.31e-4	639.45+-528.54
Func2	1.90+-2.30	0.113+-0.142	84.4+-62.03

From above tables we can know that:

Conjugate gradient method converges much faster than steepest descent method. Especially in low-dimension optimization problems. And the error terms are smaller when we use conjugate gradient method comparing to steepest descent method.

For conjugate gradient method, the number of CG iterations becomes larger when the dimension of objective function improves. In contrast, for steepest descent method, the number of iterations becomes smaller when dimension becomes larger.

The function value we get by using conjugate gradient method is closer to true value.