

Project 2 –

FMNN25: Advanced Numerical Algorithms in Python Claus Führer, Centre for Mathematical Sciences

This describes the second bigger programming project in the course, devoted to some classical optimization algorithms.

# Task 1

Design an optimization problem class in Python. Its constructor should take an objective function as input, optionally also its gradient can be provided.

# Task 2

Design a general optimization method class. It should be generic for special kinds of Quasi-Newton methods. Derive from this method class classes for special methods in the next tasks.

#### Task 3

Implement classical Newton's method for finding a minimum of the objective function. Approximate the Hessian by finite differences and a symmetrizing step:  $G := \frac{1}{2}(\bar{G} + \bar{G}^{T})$ , if necessary.

## Task 4

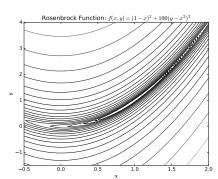
Provide Newton's method with exact line search method.

#### Task 5

Test the performance of this method on the Rosenbrock function

$$f(x) = 100(x_2 - x_1^2)^2 + (1 - x_1)^2.$$

Demonstrate its behavior also in a way, comparable to the right picture in Fig. 1.



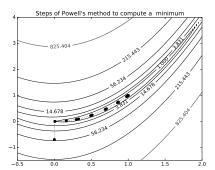


Figure 1: A contour plot of Rosenbrock's function and steps towards its minimum.

## Task 6

Write an inexact line search method based on the Goldstein/Wolfe conditions (Algorithm: Fletcher pp.34ff in the additional course material)

#### Task 7

Test this seperately from an optimization method on Rosenbrock's function and use the parameters given on p.37 in the book mentioned above.

## Task 8

Provide Newton's method with the inexact line search as an alternative to the exact one.

#### Task 9

Derive from Newton's method five classes of Quasi Newton methods:

- Simple Broyden rank-1 update of G and applying Sherman-Morisson's formula ("good Broyden")
- Simple Broyden rank-1 update of  $H = G^{-1}$  ("bad Broyden")
- Symmetric Broyden update
- DFP rank-2 update
- BFGS rank-2 update

## Task 10

Download from the course's webpage the testexample chebyquad\_problem.py. It contains the objective function chebyquad and its gradient.

The objective function in mathematical terms reads:

$$f(x) = \sum_{i=0}^{n} \left( \frac{1}{n} \sum_{j=0}^{n} T_i(2x_j - 1) - \int_0^1 T_i(2\xi - 1) d\xi \right)^2 \quad x = [x_0, \dots, x_n] \quad x_j \in [0, 1]$$
(1)

where  $T_i(t)$  is the *i*-th degree Chebyshev polynomial and  $t \in [-1, 1]$ .

# Task 11

Minimizing chebyquad corresponds to finding a set of optimal points  $x_j$  such that the mean value in (1) approximates best the corresponding integrals. Use your code to compute these points for n = 4, 8, 11. Compare your results with those obtained from scipy.optimize.fmin\_bfgs.

# Task 12

The matrix  $H^{(k)}$  of the BFGS method should approximate  $G(x^{(k)})^{-1}$ , where G is the Hessian of the problem. Study the quality of the approximation with growing k.

Lycka till!