# Exercises Lecture 4 (Chapter 6)

Make sure to import Numpy and the optimize module from SciPy.

```
import numpy as np
import scipy.optimize as optimize
```

## Question 1

Execute fsolve() with initial guesses in [-10, -9.5, -9, ..., 9, 9.5, 10] to find roots of the function  $f(x) = x \cdot \cos(x)$ . You may use a for-loop to iterate over the initial guesses. Store the found roots in a list called roots.

The output list roots should look like this. Recall that fsolve() returns an array with one element (which you can access by indexing it at position 0)

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[-10.995574287564276, 1.5707963267948954, -7.853981633974491, -7.853981633974483, -7.853981633974491, -7.8539816399140, -7.8539816399140, -7.8539816399140, -7.853981639140, -7.8539816399140, -7.85398160, -7.85598160, -7.85598160, -7.85598160, -7.85598160, -7.85598160, -7.85598160, -7.85598160, -7.85598160, -7.85598160, -7.85598160, -7.85598100, -7.85598100, -7.85598100, -7.85598100, -7.855981000000000000000000000

#### Question 2

We will write a function that can compute a root of a polynomial function.

a) Write a function that takes as input a number x and an array of coefficients  $a = [a_0, \dots, a_{n-1}]$ . It should return the function  $f(x) = \sum_{i=0}^{n-1} a_i x^i$ .

The derivative of f is given by  $f'(x) = \sum_{i=1}^{n-1} i \cdot a_i x^{i-1}$ .

- b) Write a function  $f_{deriv}$  that takes as input a number x and an array of coefficients  $a = [a_0, \dots, a_{n-1}]$ . It should return the derivative f'(x) of f in the point x.
- c) Write a function  $\mathtt{root\_polynomial}$  that takes as input an array of coefficients  $a = [a_0, \ldots, a_{n-1}]$  and an initial guess r. It should output a root of the function f if one is found, and a message saying no root was found otherwise. Use  $\mathtt{root\_scalar}()$  with Newton's method and initial guess r. The output property  $\mathtt{converged}$  of  $\mathtt{root\_scalar}()$  tells you whether or not Newton's method has found a root. Also, you will need to use the  $\mathtt{args}$  keyword  $\mathtt{argument}$ .

Your functions should give the following output on the input below.

```
# First polynomial
a = np.array([1.5, 0, -2.5, 0.3, 1])
r = 1.5

result = root_polynomial(a,r)
print(result)

# Second polynomial (this function has no root)
a2 = np.array([1.5, 0, -2.5, 0.3, 3])
r2 = 1.5

result2 = root_polynomial(a2,r2)
print(result2)
```

The found root is x = -1.5180670079327394No root was found; it might be that no root exists.

#### Question 3

In this problem, we try to solve the 2-by-2 nonlinear system of equations f(x) = 0 introduced by Bloggs where

$$f(x) = \begin{bmatrix} f_0(x) \\ f_1(x) \end{bmatrix} = \begin{bmatrix} x_1^2 - x_2 + 1 \\ x_1 - \cos(\frac{\pi}{2}x_2) \end{bmatrix}$$

with Jacobian

$$J(f) = \begin{bmatrix} 2x_1 & -1 \\ 1 & \frac{\pi}{2}\sin(\frac{\pi}{2}x_2) \end{bmatrix}.$$

The problem has solutions (-1,2), (0,1), and  $(-1/\sqrt{2},3/2)$ . The goal of this exercise is to show that the **root()** function might, or might not, converge depending on the initial guess that we choose.

a) Implement f and its Jacobian as Python functions called f and jac\_f.

For initial guesses, we will take points on the circle with radius 2 centered around (0,0). This circle can be parameterized by  $[2\cos(\rho), 2\sin(\rho)]$  for  $\rho \in [0, 2\pi]$ .

- b) Write a function circle()that takes as input a one-dimensional array  $\rho = [\rho_0, \dots, \rho_{k-1}]$  and outputs the points  $[2\cos(\rho_i), 2\sin(\rho_i)]$  on the rows of a  $k \times 2$  array. Do not use for-loops.
- c) For k=50 evenly spaced values of  $\rho \in [0,2\pi]$ , use root() to find a root of the system f(x)=0 with initial guess  $[2\cos(\rho),2\sin(\rho)]$ , Jacobian information, and the 'hybr' method. For every value of  $\rho$  the output should be a short message including 1) the value of  $\rho$  and 2) the root that

was found, or a message that no root was found. Have a look at the output message of the function  $\mathtt{root}()$  to see how you can determine whether the method found a root or not. For k=50 evenly spaced points in  $[0,2\pi]$  your output should look like this. As you can see, all of the three roots of the system are sometimes found, and also sometimes no root is found.

```
For guess = [2. 0.], found root is [1.5781495e-12 1.0000000e+00]
For guess = [1.98358003 \ 0.25575432], found root is [-2.57655834e-14 \ 1.00000000e+00]
For guess = [1.93458973 \ 0.50730917], found root is [1.23867782e-13 \ 1.00000000e+00]
For guess = [1.85383351 \ 0.75053401], found root is [1.57396466e-13 \ 1.00000000e+00]
For guess = [1.74263741 \ 0.9814351], found root is [9.05935849e-13 \ 1.00000000e+00]
For guess = [1.60282724 \ 1.19622106], found root is [1.38409487e-15 \ 1.00000000e+00]
For guess = [1.4366987 \ 1.3913651], found root is [-1.32700583e-12 \ 1.00000000e+00]
For guess = [1.2469796 \ 1.56366296], found root is [-3.80159246e-13 \ 1.00000000e+00]
For guess = [1.03678514 \ 1.71028553], found root is [8.04962149e-11 \ 1.00000000e+00]
For guess = [0.80956669 \ 1.82882525], found root is [1.54091134e-17 \ 1.00000000e+00]
For guess = [0.56905517 \ 1.91733571], found root is [-7.72638622e-15 \ 1.00000000e+00]
For guess = [0.31919979 \ 1.97436357], found root is [-5.90449013e-12 \ 1.00000000e+00]
For guess = [0.06410316 1.99897243], found root is [-0.70710678 1.5
For guess = [-0.19204605 \quad 1.99075823], found root is [-0.70710678 \quad 1.5]
                                                                                ]
For guess = [-0.44504187 \ 1.94985582], found root is [-1. \ 2.]
For guess = [-0.69073011]
                          1.87693684], found root is [-1. 2.]
For guess = [-0.92507658 \ 1.77319861], found root is [-1.
For guess = [-1.14423332 \ 1.64034451], found root is [-0.70710678]
                                                                                ]
For guess = [-1.34460178 \ 1.48055599], found root is [-0.70710678]
                                                                     1.5
                                                                                ]
For guess = [-1.52289192 \ 1.29645679], found root is [-0.70710678]
For guess = [-1.67617621 \ 1.0910698], found root is [-0.70710678]
                                                                                1
For guess = [-1.80193774 \ 0.86776748], found root is [-0.70710678 \ 1.5]
                                                                                ٦
For guess = [-1.89811149 \ 0.63021644], found root is [-1. \ 2.]
For guess = [-1.96311831]
                          0.38231726], found root is [4.60067409e-13 1.00000000e+00]
For guess = [-1.99589079 \quad 0.12814044], found root is [4.60088962e-14 \quad 1.00000000e+00]
For guess = [-1.99589079 -0.12814044], found root is [-4.39476516e-13 1.00000000e+00]
For guess = [-1.96311831 - 0.38231726], found root is [7.42968452e - 17 1.00000000e + 00]
For guess = [-1.89811149 - 0.63021644], found root is [8.20547063e-14 1.00000000e+00]
For guess = [-1.80193774 - 0.86776748], found root is [-5.80092403e-14 \ 1.00000000e+00]
For guess = [-1.67617621 -1.0910698], found root is [4.97457028e-14 1.00000000e+00]
For guess = [-1.52289192 -1.29645679], found root is [-2.90995034e-14 1.00000000e+00]
For guess = [-1.34460178 -1.48055599], found root is [2.29128908e-16 \ 1.00000000e+00]
For guess = [-1.14423332 - 1.64034451], found root is [-3.51617808e - 12 1.00000000e + 00]
For guess = [-0.92507658 -1.77319861], found root is [2.52638402e-13 1.00000000e+00]
For guess = [-0.69073011 - 1.87693684], found root is [1.66343153e - 11 1.00000000e + 00]
For guess = [-0.44504187 -1.94985582], found root is [-0.70710678]
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For guess = [-0.19204605 -1.99075823], found root is [-0.70710678]
                                                                                ]
For guess = [0.06410316 - 1.99897243], found root is [-0.70710678 1.5]
                                                                                ]
For guess = [0.31919979 - 1.97436357], no root was found
For guess = [0.56905517 - 1.91733571], no root was found
```

```
For guess = [ 0.80956669 -1.82882525], no root was found

For guess = [ 1.03678514 -1.71028553], found root is [-0.70710678 1.5 ]

For guess = [ 1.2469796 -1.56366296], found root is [-2.26536179e-16 1.00000000e+00]

For guess = [ 1.4366987 -1.3913651], found root is [9.29740721e-15 1.000000000e+00]

For guess = [ 1.60282724 -1.19622106], found root is [9.29961312e-13 1.00000000e+00]

For guess = [ 1.74263741 -0.9814351 ], no root was found

For guess = [ 1.85383351 -0.75053401], found root is [-7.74845403e-14 1.00000000e+00]

For guess = [ 1.93458973 -0.50730917], found root is [1.66798808e-12 1.00000000e+00]

For guess = [ 1.98358003 -0.25575432], found root is [-9.27679759e-14 1.00000000e+00]

For guess = [ 2.0000000e+00 -4.8985872e-16], found root is [1.57814971e-12 1.00000000e+00]
```

## Question 4

Let  $A \in \mathbb{R}^{n \times n}$  be a two-dimensional array,  $b \in \mathbb{R}^n$  a one-dimensional array and  $c \in \mathbb{R}$  a scalar. Consider the function

$$f(x) = f(x_0, \dots, x_{n-1}) = x^T A x + b^T x + c = \sum_{i=0}^{n-1} \sum_{j=0}^{n-1} a_{ij} x_i x_j + \sum_{i=0}^{n-1} b_i x_i + c$$

that takes as input a one-dimensional array x. The gradient of this function is given by  $\nabla f(x) = Ax + b$ .

Write a function quadratic\_minimizer that takes as input A, b and c. It should compute a (local) minimizer of the function f using minimize() with gradient information, the 'CG' method, and initial guess x = [1, 1, ..., 1].

It should give the following output on the given input.

```
# Example usage
A = np.array([[3, 0, 0], [0, 4, 0], [0, 0, 5]])
b = np.array([-1, -4, 1])
c = 6

result = quadratic_minimizer(A, b, c)
print("Optimal solution:", result.x)
print("Function value at optimal point:", result.fun)
```

Optimal solution: [ 0.33333333 1. -0.2 ] Function value at optimal point: 6.0

Because the matrix A is symmetric and has positive eigenvalues, the function f is convex, meaning that an optimal solution can be found at the point where the gradient is the all-zeros vector, that is,  $x^* = -A^{-1}b$ . This is indeed the solution that we found.

```
eig_A, _ = np.linalg.eig(A)
print(f"The eigenvalues of A are {eig_A}")
```

### Question 5

Write a function capacitated\_norm() taking as input vectors  $a, b, w \in \mathbb{R}^n_{\geq 0}$  and a capacity value W, that returns the solution to the problem

$$\begin{aligned} & \max & & \sqrt{\sum_{i=0}^{n-1} x_i^2} \\ & \text{s.t.} & & \sum_{i=0}^{n-1} w_i x_i \leq W \\ & & a_i \leq x_i \leq b_i \text{ for } i=0,\dots,n-1 \end{aligned}$$

using minimize() with the 'trust-constr' method. Take a as the initial guess. You may assume that  $\sum_i w_i a_i \leq W$ , so that this problem always has a solution. Hint: You may want to have a look at the constraints keyword argument in the documentation of minimize() to see how you can input additional arguments to a constraint.

```
n = 7
a = np.array([0,0,1,4,1,0,7])
b = np.array([1,4,3,6,8,1,9])
w = np.array([0,6,5,4,6,5,3])
W = 2*np.sum(a)
print(capacitated_norm(a,b,w,W))
```

[ 0.97618582 -0.89189189 0.25675676 3.40540541 0.10810811 -0.74324324 6.55405406]