Exercises for Lecture Course on Numerical Optimization (NUMOPT) Albert-Ludwigs-Universität Freiburg – Winter Term 2018-2019

Exercise 5: Exam Type Questions Inequality Constrained Optimization

(to be completed during the exercise session on Jan 11, 2018 or sent by email to messerer@tf.uni-freiburg.de before Jan 18, 2019)

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Goal of this sheet is to offer some exposure to exam-type questions. Further you will get more familiar with inequality constrained optimization and get to know more of CasADi's functionalities.

Exercise Tasks

1. **A sample exam question:** (your score in this question will be divided by 3) Regard the following minimization problem.

$$\lim_{x \in \mathbb{R}^2} x_2^4 + (x_1 + 2)^4 \quad \text{subject to} \quad \begin{cases} x_1^2 + x_2^2 \le 8 \\ x_1 - x_2 = 0 \end{cases}$$

(a) How many variables, how many equality, and how many inequality constraints does this problem have?

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(b) Sketch the feasible set $\Omega \in \mathbb{R}^2$ of this problem.

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(c) Bring this problem into the NLP standard form:

$$\min_{x \in \mathbb{R}^n} f(x) \quad \text{subject to} \quad \left\{ \begin{array}{ll} g(x) & = & 0 \\ h(x) & \geq & 0 \end{array} \right.$$

by defining the functions f, g, h appropriately.

3

| (d) | Is this optimization problem convex? Justify. | | |
|-----|---|--|---------|
| (e) | Write down the Lagrangian function of this optimization problem. | 2 | |
| (f) | A feasible solution of the problem is $\bar{x}=(2,2)^T.$ What is the active set $\mathcal{A}(\bar{x})$ at this | 2 poir | nt? |
| (g) | Is the linear independence constraint qualification (LICQ) satisfied at \bar{x} ? Justify. | 2 | |
| | | | |
| (h) | An optimal solution of the problem is $x^* = (-1, -1)^T$. What is the active set $\mathcal{A}(x)$ point? | $\begin{bmatrix} 3 \\ x^* \end{bmatrix}$ | at this |
| (i) | Is the linear independence constraint qualification (LICQ) satisfied at x^* ? Justify. | 1 | |
| (j) | Describe the tangent cone $T_{\Omega}(x^*)$ (the set of feasible directions) to the feasible set at x^* , by a set definition formula with explicitly computed numbers. | this | point |
| | | | |
| | | 2 | |

(k) Compute the Lagrange gradient and find the multiplier vectors λ^* , μ^* so that the above point x^* satisfies the KKT conditions.

(1) Describe the critical cone $C(x^*, \mu^*)$ at the point (x^*, λ^*, μ^*) in a set definition using explicitly computed numbers

3

(m) Formulate the second order necessary conditions for optimality (SONC) for this problem and test if they are satisfied at (x^*, λ^*, μ^*) . Can you prove that x^* is a local or global minimizer?

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2. Hanging chain, the last episode: Recall the optimization problem of the hanging chain with nonconvex inequality constraints and without considering a rest length:

minimize
$$\frac{1}{2} \sum_{i=1}^{N-1} D\left((y_i - y_{i+1})^2 + (z_i - z_{i+1})^2 \right) + g \sum_{i=1}^{N} m z_i$$
 (1a)

subject to
$$(y_1, z_1) = (-2, 1)$$
 (1b)

$$(y_N, z_N) = (2, 1)$$
 (1c)

$$z_i + y_i^2 \ge 0,$$
 for $i = 1, ..., N.$ (1d)

with parameters N = 40, D = 70, g = 9.81 and m = 0.1.

If you define this problem with opti = casadi.Opti(), and add decision variables and constraints with opti.variable() and opti.subject_to() respectively, CasADi will automatically bring it into the CasADi standard form:

minimize
$$f(x)$$
 (2a)
subject to $g_{lb} \le g(x) \le g_{ub}$ (2b)

subject to
$$g_{lb} \le g(x) \le g_{ub}$$
 (2b)

where for equality constraints the upper and lower bounds are identical. x is a collection of all decision variables, in order of their definition in the script. You can access these (symbolic) expressions and variables via opti.x, opti.f, opti.g, opti.lbg and opti.ubg. After solving the problem with sol = opti.solve(), you can access the value of any expression at the solution via e.g. fmin = sol.value (opti.f). Values obtained like this are still CasADi matrices. Therefore it might be necessary to first change them to a normal Matlab matrix using full () (e.g. fmin = full(fmin)) before further processing them.

(a) Solve problem (1) with Opti () and IPOPT as you did on the first exercise sheet. Find out the standard form into which CasADi has transformed the problem and write it down on paper, i.e. the relation between x and the y_i and z_i as well as the functions f(x), g(x) and the values of $g_{\rm lb}$ and $g_{\rm ub}$.

(2 bonus points)

(b) With opti.callback() you can pass a function handle @(i) DOSOMETHING with iteration index i as input. CasADi will call it every iteration. Create a function that makes a plot of the current chain position and pass it to CasADi via a function handle.

Hint: You can access the non-converged solution via opti.debug.value(). To be able to see something, use pause() after plotting.

(2 bonus points)

(c) Identify the active set at the solution. Does LICQ hold? Are there any weakly active constraints?

Note: You can consider a constraint $h(x) \ge 0$ *as active if* $|h(x)| \le 10^{-6}$

(2 bonus points)

- (d) Evaluate the Hessian of the Lagrangian at the solution. You may first build a CasADi expression for the Lagrangian. You can then obtain its Hessian via hessian (). You can access the Lagrange multipliers as optilam_g. Note that as opposed to the lecture, the Lagrangian is defined as $\mathcal{L}(x,\lambda) = f(x) + \lambda^{\top} g(x)$ in CasADi. This also changes the sign of the multipliers.
 - (2 bonus points)
- (e) Calculate the null-space of the Jacobian of the active constraints at the solution using QR decomposition. Use the result to calculate the reduced Hessian. Is the reduced Hessian positive definite? What does this mean?

Hints: Use qr(). If you do not succeed obtaining the null-space via QR decomposition, you can simply proceed by obtaining the null-space from null().

(2 bonus points)