Exercise: Time-optimal control

1 The effect of non-fixed time on problem structure.

In this section we will look at the dynamics of a point mass in a 2D x-y plane, controllable by forces acting along both axes.

The point mass starts from standstill at (0,1) and needs to visit a fine grid of waypoints.

The point mass has limits on position, speed and acceleration constraints.

Tasks:

- 1. Locate and run the reference multiple-shooting implementation template. Interpret the code and the plots. Why does the OCP solution deviate so much from the reference trajectory?
- 2. Change the integration method to explicit Euler, and change the NLP solver the sqpmethod. Verify that the objective (=tracking error) is 2.530909e-01. Explain why this takes only one SQP step.
- 3. Make the control horizon a decision variable, bounded between $0.5 \mathrm{~s}$ and $2 \mathrm{~s}$, and initialize with 1. Solve with ipopt solver and verify that you obtain a much improved tracking error $1.972783\mathrm{e}{-02}$.

Switch again to sqpmethod solver. Why does the SQP method take more than 1 step? The solver might fail to converge unless you specify a strategy to deal with non-convex subproblems (default is to ignore convexity):

```
opti.solver('sqpmethod',{"convexify_strategy": "eigen-clip"})
```

4. (extra) Explain the dense column in the constraint Jacobian. Make a re-implementation of 1.3 that does not have a dense column in the constraint Jacobian, but still yields the exact same tracking error. Compare the timing in the 'convexify' step with the previous exercise

2 Racing around obstacles

Tasks:

1. Start again from template. Introduce a circle in the scene, centered around (1,0.75) with a radius of 0.6. Remove the tracking objective. Find the time optimal trajectory that starts from standstill at (0,0) and ends at (2,1) without a prescribed final velocity.

Verify that you obtain an optimal time of either 1.0437847e+00 or 8.4867232e-01 seconds.

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- 2. (extra) Give the circle a radius of 0.601 and compute how much the optimal time T^\star changed. Use a finite difference to approximate $\frac{\partial T^\star}{\partial r}$.
 - Inspect the Lagrange multipliers (a.k.a dual variable variables) associated with the obstacle (see https://web.casadi.org/docs/#dual-variables for syntax). Explain why most multipliers are very small.
 - Can you find the relationship between the multiplier of the active constraint and $\frac{\partial T^{\star}}{\partial r}$?
- 3. Increase the control horizon to N=40. Find the time optimal trajectory that starts from standstill at (0,0), passes through transit point (2,1) at k=20 without a prescribed velocity, and return to (0,0) without a prescribed final velocity.
 - Find the clockwise solution. Verify that you obtain $T^{\star} = 2.2065$
- 4. Instead of prescribing that the transit point should be encountered exactly in the middle of the total control time, consider a variant in which the encounter may happen at any time instance the solver chooses. Do you expect T^* to increase or decrease? Verify that T^* changes by 0.2208.
- 5. (extra) Find a periodic trajectory around the obstacle that passes through (0,0) and (2,1) without prescribed velocity.

Verify that you obtain $T^* = 1.9430$.

3 (extra) Time-optimal path following

- 1. Start again from template. Change to a CasADi built-in integrator object (cvodes). Verify that the solution stays very much the same.
- 2. Make the control horizon a decision variable again. Verify that you obtain a tracking error of 2.7382665e-03
- 3. (extra) Change the problem as to require the end position to be $(\sin(2), \cos(2))$, and add a*T to the objective to effectively create a trade-off between time-optimality and tracking error. Compare for a=1 (leads to $T^*=0.96184$) and a=10 (leads to $T^*=0.62242$).
- 4. (extra) Trade-offs like this are easy enough to suggest and solve, but difficult to tune. The factor a has little intuitive meaning. Change to a more intuitive scheme: optimize purely for time, and prescribe a maximum deviation (0.2) of any point w.r.t to the reference ref. Verify that you obtain $T^* = 0.97318$.
- 5. (extra) In robotics, it is often required to follow a geometric path: ie a path without timing information, unlike our reference ref which speeds ahead at a constant pace. Adapt 3.4 such that the optimal trajectory stays within a limited bound to the geometric path $s \mapsto \left[\sin(2s) \cos(2s)\right]^T$, s = 0..1. Verify that you obtain $T^* = 0.66866$.

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