```
In [1]:
        import Pkg
        Pkg.activate(@__DIR__)
        Pkg.instantiate()
        import MathOptInterface as MOI
        import Ipopt
        import FiniteDiff
        import ForwardDiff
        import Convex as cvx
        import ECOS
        using LinearAlgebra
        using Plots
        using Random
        using JLD2
        using Test
        import MeshCat as mc
        using Statistics
```

Activating environment at `~/OCRL/HW3\_S24/Project.toml`

```
In [2]: include(joinpath(@__DIR__, "utils","fmincon.jl"))
  include(joinpath(@__DIR__, "utils","planar_quadrotor.jl"))
```

check dynamic feasibility (generic function with 1 method)

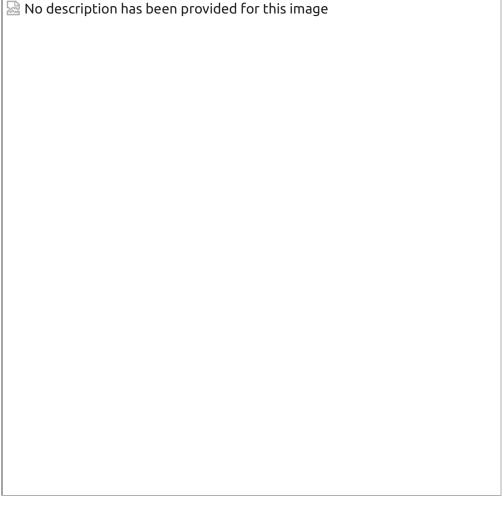
## Q3: Quadrotor Reorientation (40 pts)

In this problem, you will use the trajectory optimization tools you have demonstrated in questions one and two to solve for a collision free reorientation of three planar quadrotors. The planar quadrotor (as described in lecture 9) is described with the following state and dynamics:

$$x = egin{bmatrix} p_x \ p_z \ heta \ v_x \ v_z \ \omega \end{bmatrix},$$
 (1) $\dot{x}$  =

where  $p_x$  and  $p_z$  are the horizontal and vertial positions,  $v_x$  and  $v_z$  are the corresponding velocities,  $\theta$  for orientation,  $\omega$  for angular velocity,  $\ell$  for length of the quadrotor, m for mass, g for gravity acceleration in the -z direction, and a moment of inertia of J.

You are free to use any solver/cost/constraint you would like to solve for three collision free, dynamically feasible trajectories for these quadrotors that looks something like the following:



(if an animation doesn't load here, check out quadrotor reorient.gif.)

Here are the performance requirements that the resulting trajectories must meet:

- The three quadrotors must start at x1ic, x2ic, and x2ic as shown in the code (these are the initial conditions).
- The three quadrotors must finish their trajectories within .2 meters of x1g, x2g, and x2g (these are the goal states).
- The three quadrotors must never be within **0.8** meters of one another (use  $[p_x,p_z]$  for this).

There are two main ways of going about this:

- 1. **Cost Shaping**: Design cost functions for each quadrotor that motivates them to take paths that do not result in a collision. You can do something like designing a reference trajectory for each quadrotor to use in the cost. You can use iLQR or DIRCOL for this.
- 2. **Collision Constraints**: You can optimize over all three quadrotors at once by creating a new state  $\tilde{x}=[x_1^T,x_2^T,x_3^T]^T$  and control  $\tilde{u}=[u_1^T,u_2^T,u_3^T]^T$ , and then directly include collision avoidance constraints. In order to use constraints, you must use

DIRCOL (at least for now).

## Hints

- You should not use norm() >= R in any constraints, instead you should square the constraint to be  $norm()^2 >= R^2$ . This second constraint is still non-convex, but it is differentiable everywhere.
- If you are using DIRCOL, you can initialize the solver with a "guess" solution by linearly interpolating between the initial and terminal conditions. Julia let's you create a length N linear interpolated vector of vectors between a::Vector and b::Vector like this: range(a, b, length = N) (experiment with this to see how it works).

You can use either RK4 (iLQR or DIRCOL) or Hermite-Simpson (DIRCOL) for your integration. The dt = 0.2, and tf = 5.0 are given for you in the code (you may change these but only if you feel you really have to).

```
In [3]: | function single quad dynamics(params, x,u)
             # planar quadrotor dynamics for a single quadrotor
             # unpack state
             px, pz, \theta, vx, vz, \omega = x
             xdot = [
                 VX,
                 VΖ,
                 (1/params.mass)*(u[1] + u[2])*sin(\theta),
                 (1/params.mass)*(u[1] + u[2])*cos(\theta) - params.g,
                 (params.\ell/(2*params.J))*(u[2]-u[1])
             ]
             return xdot
        end
         function combined dynamics(params, x,u)
             # dynamics for three planar quadrotors, assuming the state is stacked
             # in the following manner: x = [x1;x2;x3]
             # NOTE: you would only need to use this if you chose option 2 where
             # you optimize over all three trajectories simultaneously
             # quadrotor 1
             x1 = x[1:6]
             u1 = u[1:2]
             xdot1 = single quad dynamics(params, x1, u1)
             # quadrotor 2
             x2 = x[(1:6) .+ 6]
             u2 = u[(1:2) + 2]
             xdot2 = single quad dynamics(params, x2, u2)
```

```
# quadrotor 3
x3 = x[(1:6) .+ 12]
u3 = u[(1:2) .+ 4]
xdot3 = single_quad_dynamics(params, x3, u3)

# return stacked dynamics
return [xdot1;xdot2;xdot3]
end
```

combined\_dynamics (generic function with 1 method)

```
In [70]:
        #Helper Functions
        function hermite simpson(params::NamedTuple, x1::Vector, x2::Vector, u, dt::
            # TODO: input hermite simpson implicit integrator residual
           x m = 0.5*(x1+x2) + (dt/8)*(combined dynamics(params, x1, u) - combined
           xk dot = combined dynamics(params,x m,u)
            res = x1 + dt .* (combined dynamics(params, x1, u)+4*xk dot + combined dyn
            return res
        end
        function compute quad cost(params::NamedTuple, Z::Vector)::Real
            idx, N, xg = params.idx, params.N, params.xg
           Q, R, Qf = params.Q, params.R, params.Qf
           # TODO: input cartpole LQR cost
           J = 0
           for i = 1:(N-1)
               xi = Z[idx.x[i]]
               ui = Z[idx.u[i]]
               x gi = transpose(xi-xg)*Q*(xi-xg)
               J += 0.5*x gi + transpose(ui)*R*ui
           end
           # dont forget terminal cost
           xN = Z[idx.x[N]]
           x gN = transpose(xN-xg)*Qf*(xN-xg)
            J += 0.5*x gN
            return J
        end
        function quad dynamic contraints(params::NamedTuple, Z::Vector)::Vector
            idx, N, dt = params.idx, params.N, params.dt
           # TODO: create dynamics constraints using hermite simpson
           # create c in a ForwardDiff friendly way (check HWO)
           c = zeros(eltype(Z), idx.nc)
            for i = 1:(N-1)
               xi = Z[idx.x[i]]
               ui = Z[idx.u[i]]
               xip1 = Z[idx.x[i+1]]
```

```
# TODO: hermite simpson
        \# c[idx.c[i]] = zeros(4)
        c[idx.c[i]] = hermite simpson(params,xi,xip1,ui,dt)
    end
    return c
end
function quad equality constraint(params::NamedTuple, Z::Vector)::Vector
    idx, N, dt = params.idx, params.N, params.dt
    x0 = Z[idx.x[1]]
    xN = Z[idx.x[N]]
    xic = [params.x1ic; params.x2ic;params.x3ic]
    xg = [params.x1g;params.x2g;params.x3g]
    \# c = zeros(eltype(Z), idx.nc)
    c = quad dynamic contraints(params, Z)
    res = [x0-xic; xN-xg; c]
    return res
end
function quad inequality constraint(params::NamedTuple, Z::Vector)::Vector
    idx, N = params.idx, params.N
    c = similar(Z, 3 * (N - 1))
    for i in 1:N-1
        xi idx = idx.x[i]
        xi = Z[xi idx]
        offsets = [0, 6, 12]
        for j in 1:3
            x1_idx, x2_idx = xi_idx + offsets[j], xi_idx + offsets[j] + 1
            x1 = Z[x1 idx][1:2]
            x2 = Z[x2 idx][1:2]
            dist_squared = norm(x1 - x2)^2
            c[3 * (i - 1) + j] = 0.8^2 - dist squared
        end
    end
    return c
function create idx(nx,nu,N)
    # This function creates some useful indexing tools for Z
   \# \times i = Z[idx.x[i]]
   \# u i = Z[idx.u[i]]
   # Feel free to use/not use anything here.
    # our Z vector is [x0, u0, x1, u1, ..., xN]
   nz = (N-1) * nu + N * nx # length of Z
    x = [(i - 1) * (nx + nu) .+ (1 : nx) for i = 1:N]
   u = [(i - 1) * (nx + nu) .+ ((nx + 1):(nx + nu))  for i = 1:(N - 1)]
    # constraint indexing for the (N-1) dynamics constraints when stacked up
```

```
c = [(i - 1) * (nx) .+ (1 : nx) for i = 1:(N - 1)]
    nc = (N - 1) * nx # (N-1)*nx
    return (nx=nx, nu=nu, N=N, nz=nz, nc=nc, x=x, u=u, c=c)
end
0.00
    quadrotor reorient
Function for returning collision free trajectories for 3 quadrotors.
Outputs:
    x1::Vector{Vector} # state trajectory for quad 1
    x2::Vector{Vector} # state trajectory for quad 2
    x3::Vector{Vector} # state trajectory for quad 3
    u1::Vector{Vector} # control trajectory for quad 1
    u2::Vector{Vector} # control trajectory for quad 2
    u3::Vector{Vector} # control trajectory for quad 3
    t vec::Vector
    params::NamedTuple
The resulting trajectories should have dt=0.2, tf = 5.0, N = 26
where all the x's are length 26, and the u's are length 25.
Each trajectory for quad k should start at `xkic`, and should finish near
`xkg`. The distances between each quad should be greater than 0.8 meters at
every knot point in the trajectory.
function quadrotor reorient(;verbose=true)
    # problem size
    nx = 18
    nu = 6
    dt = 0.2
    tf = 5.0
    t vec = 0:dt:tf
    N = length(t vec)
    # indexing
    idx = create idx(nx,nu,N)
    # initial conditions and goal states
    lo = 0.5
    mid = 2
    hi = 3.5
    x1ic = [-2, lo, 0, 0, 0, 0] # ic for quad 1
    x2ic = [-2, mid, 0, 0, 0, 0] # ic for quad 2
    x3ic = [-2,hi,0,0,0,0] # ic for quad 3
    xic = [x1ic; x2ic; x3ic]
    x1g = [2,mid,0,0,0,0] # goal for quad 1
    x2g = [2,hi,0,0,0,0] # goal for quad 2

x3g = [2,lo,0,0,0,0] # goal for quad 3
    xg = [x1g; x2g; x3g]
    Q = diagm(ones(nx))
    R = 0.1*diagm(ones(nu))
```

```
Qf = 10*diagm(ones(nx))
# load all useful things into params
# TODO: include anything you would need for a cost function (like a Q, R
# LQR cost)
params = (xlic=xlic,
          x2ic=x2ic
          x3ic=x3ic
          x1g = x1g,
          x2g = x2g
          x3g = x3g
          xic = xic,
          xg = xg,
          dt = dt,
          N = N,
          idx = idx,
          mass = 1.0, # quadrotor mass
          g = 9.81, # gravity
          \ell = 0.3, # quadrotor length
          J = .018, # quadrotor moment of inertia
          Q = Q
          Qf = Qf,
          R = R
# TODO: solve for the three collision free trajectories however you like
idx = params.idx
nu = idx.nu
nx = idx.nx
# TODO: primal bounds
# you may use Inf, like Inf*ones(10) for a vector of positive infinity
x l = -Inf*ones(idx.nz)
x u = Inf*ones(idx.nz)
# TODO: inequality constraint bounds
c l = 0.64*ones(3*(N-1))
c u = Inf*ones(3*(N-1))
#initial guess
z0 = zeros(idx.nz)
x0 = range(xic,xg, length=N)
for i=1:(N-1)
    z0[idx.x[i]] = x0[i]
end
diff type = :auto
Z = fmincon(quadrotor_cost,quad_equality_constraint,quad_inequality_cons
                x_l,x_u,c_l,c_u,z0,params, diff_type;
                tol = 1e-6, c_tol = 1e-6, max_iters = 10_000, verbose =
# pull the X and U solutions out of Z
X = [Z[idx.x[i]]  for i = 1:N]
U = [Z[idx.u[i]] \text{ for } i = 1:(N-1)]
# return the trajectories
```

```
x1 = [X[i][1:6] for i=1:N]
x2 = [X[i][7:12] for i=1:N]
x3 = [X[i][13:18] for i=1:N]
u1 = [U[i][1:2] for i=1:(N-1)]
u2 = [U[i][3:4] for i=1:(N-1)]
u3 = [U[i][5:6] for i=1:(N-1)]

return x1, x2, x3, u1, u2, u3, t_vec, params
end
```

quadrotor reorient

```
In [71]: @testset "quadrotor reorient" begin
             X1, X2, X3, U1, U2, U3, t vec, params = quadrotor reorient(verbose=true
             #-----testing-----
             # check lengths of everything
             Qtest length(X1) == length(X2) == length(X3)
             @test length(U1) == length(U2) == length(U3)
             @test length(X1) == params.N
             @test length(U1) == (params.N-1)
             # check for collisions
             distances = [distance between quads(x1[1:2],x2[1:2],x3[1:2]) for (x1,x2,
             @test minimum(minimum.(distances)) >= 0.799
             # check initial and final conditions
             @test norm(X1[1] - params.xlic, Inf) <= 1e-3</pre>
             [atest norm(X2[1] - params.x2ic, Inf) <= 1e-3]
             (3[1] - params.x3ic, Inf) <= 1e-3
             @test norm(X1[end] - params.x1g, Inf) <= 2e-1</pre>
             @test norm(X2[end] - params.x2g, Inf) <= 2e-1</pre>
             @test norm(X3[end] - params.x3g, Inf) <= 2e-1</pre>
             # check dynamic feasibility
             @test check dynamic feasibility(params,X1,U1)
             @test check dynamic feasibility(params, X2, U2)
             @test check dynamic feasibility(params, X3, U3)
             #-----plotting/animation-----
             display(animate planar quadrotors(X1,X2,X3, params.dt))
             plot(t vec, 0.8*ones(params.N),ls = :dash, color = :red, label = "collis")
                  xlabel = "time (s)", ylabel = "distance (m)", title = "Distance bet
             display(plot!(t vec, hcat(distances...)', label = ["|r 1 - r 2|" "|r 1 -
             X1m = hcat(X1...)
             X2m = hcat(X2...)
             X3m = hcat(X3...)
             plot(X1m[1,:], X1m[2,:], color = :red,title = "Quadrotor Trajectories",
             plot!(X2m[1,:], X2m[2,:], color = :green, label = "quad 2",xlabel = "p x"
             display(plot!(X3m[1,:], X3m[2,:], color = :blue, label = "quad 3"))
```

```
plot(t_vec, X1m[3,:], color = :red,title = "Quadrotor Orientations", lab
plot!(t_vec, X2m[3,:], color = :green, label = "quad 2",xlabel = "time (
    display(plot!(t_vec, X3m[3,:], color = :blue, label = "quad 3"))
end
```

```
-----checking dimensions of everything------
-----all dimensions good-----
-----diff type set to :auto (ForwardDiff.jl)----
-----testing objective gradient-----
-----testing constraint Jacobian-----
-----successfully compiled both derivatives-----
-----IPOPT beginning solve-----
This is Ipopt version 3.14.4, running with linear solver MUMPS 5.4.1.
Number of nonzeros in equality constraint Jacobian...:
                                                     300348
Number of nonzeros in inequality constraint Jacobian.:
                                                      46350
Number of nonzeros in Lagrangian Hessian....:
Total number of variables....:
                                                        618
                   variables with only lower bounds:
                                                          0
               variables with lower and upper bounds:
                                                          0
                   variables with only upper bounds:
                                                          0
Total number of equality constraints....:
                                                        486
Total number of inequality constraints.....:
                                                         75
       inequality constraints with only lower bounds:
                                                         75
  inequality constraints with lower and upper bounds:
                                                          0
       inequality constraints with only upper bounds:
                                                          0
iter
                            inf du lg(mu) ||d|| lg(rg) alpha_du alpha_pr
       objective
                   inf pr
ls
     4.1433000e+02 3.50e+00 3.50e+01
                                     0.0 0.00e+00
                                                       0.00e+00 0.00e+00
0
  1 4.1022142e+02 3.45e+00 5.39e+03 -5.8 5.98e+00
                                                    - 4.96e-02 1.54e-02
f
  1
  2 4.1052850e+02 3.45e+00 3.89e+04
                                     1.1 7.68e+04
                                                       5.08e-06 1.16e-06
  2
f
  3 4.1038180e+02 3.44e+00 4.43e+04
                                     0.1 8.20e+01
                                                    - 3.84e-03 7.69e-04
  1
h
  4 4.1049749e+02 3.43e+00 6.69e+04 -0.3 7.74e+01
                                                    - 2.72e-03 2.79e-03
f
  1
  5 4.1013508e+02 3.42e+00 8.09e+04
                                     0.0 4.07e+01
                                                    - 4.48e-03 4.18e-03
  1
    4.5862429e+02 3.01e+00 6.13e+05
                                     0.6 3.08e+01
                                                       2.79e-02 1.20e-01
  6
f
  1
  7
     5.5182365e+02 1.91e+00 7.32e+05
                                     0.3 9.01e+00
                                                    - 2.20e-01 5.00e-01
h
  2
  8
     5.5553020e+02 1.37e+00 1.19e+02 -0.4 5.43e+00
                                                       5.63e-01 1.00e+00
h
  1
     4.9665695e+02 6.19e-01 1.51e+01 -1.1 4.30e+00
  9
                                                    - 8.94e-01 1.00e+00
f
  1
iter
       objective
                   inf pr
                           inf du lg(mu) ||d|| lg(rg) alpha du alpha pr
ls
  10 4.7717237e+02 1.60e-01 4.31e+00 -6.5 3.30e+00
                                                       8.84e-01 8.03e-01
  11
    4.6632432e+02 1.14e-01 7.85e+00 -0.6 4.05e+00
                                                    - 6.14e-01 2.91e-01
f 1
  12 4.5729241e+02 9.66e-02 8.48e+00 -2.0 6.35e+00
                                                    - 2.85e-01 1.62e-01
 1
  13 4.5156316e+02 7.53e-02 4.67e+00 -1.7 5.56e+00
                                                    - 2.27e-01 2.27e-01
 1
  14 4.4642027e+02 6.22e-02 4.73e+00 -1.1 8.80e+00
                                                    - 6.84e-02 1.75e-01
```

f 1

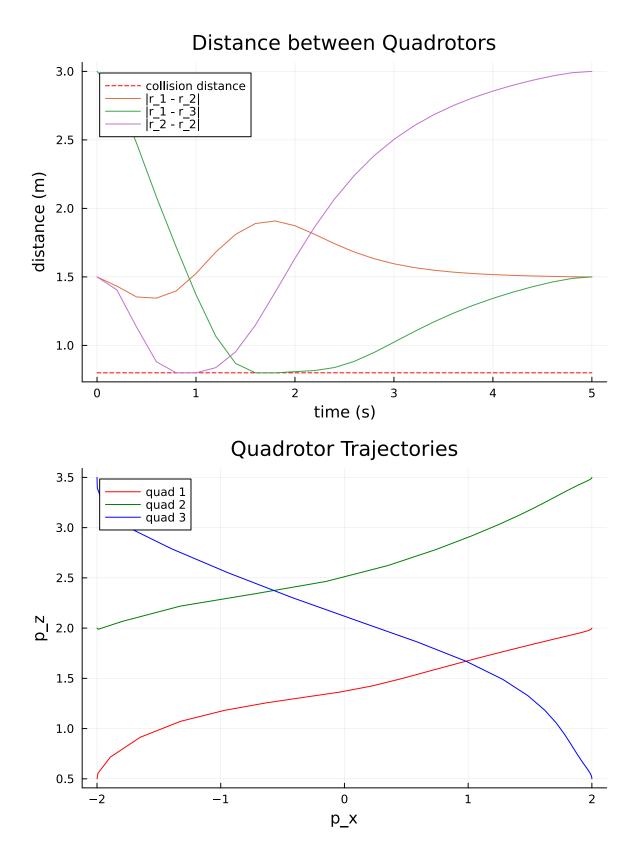
```
4.4411934e+02 4.62e-02 3.49e+00 -2.4 3.96e+00 - 3.86e-01 2.57e-01
 15
f 1
 16 4.4219688e+02 8.66e-03 3.00e+00 -2.3 1.17e+00 - 2.84e-01 8.27e-01
f 1
 17
    4.4173974e+02 5.65e-03 1.93e+00 -2.7 4.05e-01 - 4.84e-01 3.54e-01
h 1
    4.4167715e+02 5.14e-03 3.38e+00 -3.8 2.58e-01 - 4.15e-01 8.99e-02
h 1
 19 4.4135423e+02 5.87e-03 1.85e+00 -2.6 8.46e-01 - 1.90e-01 9.84e-01
f 1
      objective inf pr inf du lg(mu) ||d|| lg(rg) alpha du alpha pr
iter
ls
 20 4.4114857e+02 1.71e-03 2.92e-01 -3.2 5.90e-01 - 9.95e-01 7.71e-01
 21 4.4113255e+02 5.48e-04 2.59e-01 -4.5 1.51e-01 - 5.11e-01 9.61e-01
 22 4.4112482e+02 8.20e-05 7.89e-02 -3.6 7.23e-02 - 9.99e-01 1.00e+00
 23
    4.4112218e+02 1.01e-05 1.46e-02 -5.3 1.79e-02 - 1.00e+00 1.00e+00
h 1
 24 4.4112185e+02 9.36e-07 9.18e-03 -6.9 9.65e-03 - 1.00e+00 1.00e+00
h 1
 25 4.4112177e+02 1.17e-08 5.64e-03 -6.1 2.62e-02 - 1.00e+00 1.00e+00
 26 4.4112136e+02 2.93e-06 1.61e-02 -7.8 8.06e-03 - 1.00e+00 1.00e+00
f 1
 27 4.4112126e+02 1.23e-06 2.33e-03 -9.3 7.83e-03 - 1.00e+00 1.00e+00
h 1
 28
    4.4112128e+02 1.55e-08 2.29e-03 -11.0 1.52e-03 - 1.00e+00 1.00e+00
 29 4.4112127e+02 6.18e-09 1.73e-04 -11.0 6.10e-04 - 1.00e+00 1.00e+00
h 1
      objective inf pr inf du lg(mu) ||d|| lg(rg) alpha du alpha pr
iter
ls
 30 4.4112127e+02 1.92e-09 7.67e-05 -11.0 5.62e-04 - 1.00e+00 1.00e+00
 31 4.4112127e+02 6.28e-13 1.48e-04 -11.0 6.49e-04 - 1.00e+00 1.00e+00
 32 4.4112127e+02 1.57e-09 4.87e-05 -11.0 1.64e-04 - 1.00e+00 1.00e+00
 33
    4.4112127e+02 2.18e-10 9.30e-06 -11.0 1.14e-04 - 1.00e+00 1.00e+00
 34 4.4112127e+02 1.91e-11 8.35e-06 -11.0 3.82e-05 - 1.00e+00 1.00e+00
h 1
 h 1
 36 4.4112127e+02 2.22e-15 7.64e-06 -11.0 3.09e-05 - 1.00e+00 1.00e+00
    4.4112127e+02 3.20e-12 2.14e-06 -11.0 9.69e-06 - 1.00e+00 1.00e+00
 37
```

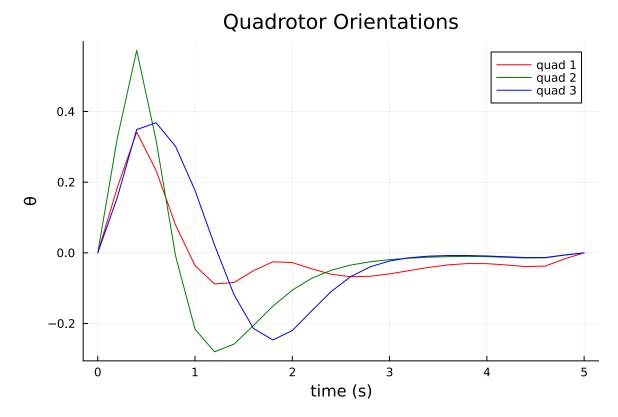
Number of Iterations....: 38

```
(scaled)
                                                          (unscaled)
Objective...... 4.4112127364108761e+02
                                                    4.4112127364108761e+02
Dual infeasibility....: 3.9802648754694303e-07
                                                    3.9802648754694303e-07
Constraint violation...: 3.5682568011452531e-13
                                                    3.5682568011452531e-13
Variable bound violation: 0.0000000000000000e+00
                                                    0.000000000000000e+00
Complementarity.....: 1.0000074458175432e-11
                                                    1.0000074458175432e-11
Overall NLP error.....: 3.9802648754694303e-07
                                                    3.9802648754694303e-07
Number of objective function evaluations
                                                  = 46
Number of objective gradient evaluations
                                                  = 39
Number of equality constraint evaluations
                                                 = 46
Number of inequality constraint evaluations
                                                 = 46
Number of equality constraint Jacobian evaluations = 39
Number of inequality constraint Jacobian evaluations = 39
Number of Lagrangian Hessian evaluations
                                                  = 0
Total seconds in IPOPT
                                                   = 8.673
EXIT: Optimal Solution Found.
\Gamma Info: Listening on: 127.0.0.1:8733, thread id: 1
@ HTTP.Servers /home/rsharde/.julia/packages/HTTP/enKbm/src/Servers.jl:369
\Gamma Info: MeshCat server started. You can open the visualizer by visiting the
following URL in your browser:
http://127.0.0.1:8733
```

L @ MeshCat /home/rsharde/.julia/packages/MeshCat/QXID5/src/visualizer.jl:64

Open Controls





In [ ]: