```
In [ ]: 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 project at `~/Desktop/2024Spring/CMU16745_OptimalControl/CMU16-745-Optimal-Control-HW/hw3`
In [ ]: include(joinpath(@_DIR__, "utils","fmincon.jl"))
        include(joinpath(@__DIR___, "utils","planar_quadrotor.jl"))
Out[]: check_dynamic_feasibility (generic function with 1 method)
```

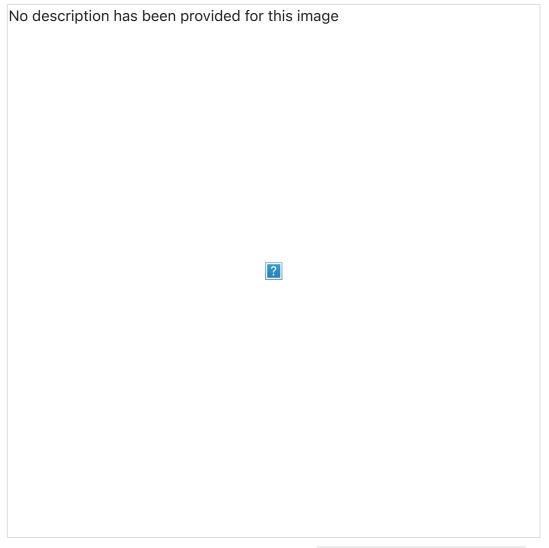
Q3: Quadrotor Reorientation (40 pts)

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:

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

where p_x and p_z are the horizontal and vertial positions, v_x and v_z are the corresponding velocities, θ for orientation, ω for angular velocity, ℓ 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 $^{\circ}0.8^{\circ}$ 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 []: function single_quad_dynamics(params, x,u)
    # planar quadrotor dynamics for a single quadrotor

# unpack state
px,pz,0,vx,vz,w = x

xdot = [
    vx,
    vz,
    w,
    (1/params.mass)*(u[1] + u[2])*sin(0),
    (1/params.mass)*(u[1] + u[2])*cos(0) - params.g,
```

```
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
Out[]: combined_dynamics (generic function with 1 method)
In [ ]: function task_cost(params::NamedTuple, Z::Vector)::Real
            # compute the cost
            J = 0.0
            for i = 1:(params.N-1)
                xi = Z[params.idx.x[i]]
                ui = Z[params.idx.u[i]]
                xi1 = xi[1:2]
```

 $(params. \ell/(2*params.J))*(u[2]-u[1])$

1

```
\theta 1 = xi[3]
                xi2 = xi[7:8]
                \theta 2 = xi[9]
                xi3 = xi[13:14]
                \theta3 = xi[15]
                J += norm(xi1 - params.x1q[1:2])^2 + norm(xi2 - params.x2q[1:2])^2 + norm(xi3 - params.x3q[1:2])^3
                J += 0.1 * (01^2 + 02^2 + 03^2)
            end
             return J
        end
Out[]: task_cost (generic function with 1 method)
In []: function hermite_simpson(params::NamedTuple, x1::Vector, x2::Vector, u, dt::Real)::Vector
            # TODO: input hermite simpson implicit integrator residual
            x mid =
                 0.5 * (x1 + x2) +
                0.125 * dt * (combined_dynamics(params, x1, u) - combined_dynamics(params, x2, u))
             return x1 +
                1 / 6 *
                 dt *
                     combined_dynamics(params, x1, u) +
                     4 * combined_dynamics(params, x_mid, u) +
                     combined dynamics(params, x2, u)
                 ) - x2
        end
        function eq_constrains(params::NamedTuple, Z::Vector)::Vector
            c = zeros(eltype(Z), 6*3*(params.N+1))
            # dynamic constrains
            for i = 1:(params.N-1)
                xi = Z[params.idx.x[i]]
                xip1 = Z[params.idx.x[i+1]]
                ui = Z[params.idx.u[i]]
                # dynamics constraints
                c[18*(i+1) + (1:18)] = hermite_simpson(params, xi, xip1, ui, params.dt)
            end
```

```
# initial condition
            x1 = Z[params.idx.x[1]]
            c[1:6] = x1[1:6] - params.x1ic
            c[7:12] = x1[7:12] - params.x2ic
            c[13:18] = x1[13:18] - params.x3ic
            # final condition
            xf = Z[params.idx.x[end]]
            c[19:24] = xf[1:6] - params.x1g
            c[25:30] = xf[7:12] - params.x2g
            c[31:36] = xf[13:18] - params.x3g
            return c
        end
Out[]: eq_constrains (generic function with 1 method)
In [ ]: function ineq_constrains(params::NamedTuple, Z::Vector)::Vector
            c = zeros(eltype(Z), 3*params.N)
            for i = 1:(params.N)
                xi = Z[params.idx.x[i]]
                xi1 = xi[1:2]
                xi2 = xi[7:8]
                xi3 = xi[13:14]
                d12 = norm(xi1 - xi2)^2
                d13 = norm(xi1 - xi3)^2
                d23 = norm(xi2 - xi3)^2
                D = params.R
                c[i] = D^2 - d12
                c[i + params.N] = D^2 - d13
                c[i + params.N*2] = D^2 - d23
            end
            return c
        end
```

```
In [ ]: function create_idx(nx,nu,N)
            # This function creates some useful indexing tools for Z
            \# x_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
        1111111
            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 guad 1
            u2::Vector{Vector} # control trajectory for guad 2
            u3::Vector{Vector} # control trajectory for guad 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
   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
   # load all useful things into params
   # TODO: include anything you would need for a cost function (like a Q, R, Qf if you were doing an
   # LOR cost)
    params = (x1ic=x1ic,
              x2ic=x2ic,
              x3ic=x3ic,
              x1q = x1q
              x2g = x2g
              x3q = x3q
              dt = dt,
              N = N
```

```
idx = idx,
         mass = 1.0, # quadrotor mass
         g = 9.81, # gravity
         J = .018.
         R = 0.8
         Rq = 0.2) # quadrotor moment of inertia
# TODO: solve for the three collision free trajectories however you like
x11_guess = range(params.x1ic[1], params.x1g[1], length=params.N)
x12_guess = range(params.x1ic[2], params.x1g[2], length=params.N)
x21_guess = range(params.x2ic[1], params.x2g[1], length=params.N)
x22_guess = range(params.x2ic[2], params.x2g[2], length=params.N)
x31 quess = range(params.x3ic[1], params.x3q[1], length=params.N)
x32_guess = range(params.x3ic[2], params.x3g[2], length=params.N)
z0 = zeros(params.idx.nz)
for i = 1:params.N
    z0[params.idx.x[i][1]] = x11_guess[i]
    z0[params.idx.x[i][2]] = x12\_guess[i]
    z0[params.idx.x[i][1] + 6] = x21\_guess[i]
    z0[params.idx.x[i][2] + 6] = x22 quess[i]
    z0[params.idx.x[i][1] + 12] = x31 quess[i]
    z0[params.idx.x[i][2] + 12] = x32 quess[i]
end
diff_type = :auto
Z = fmincon(
    task cost,
    eq constrains,
    ineg constrains,
    ones(params.idx.nz) * -10,
    ones(params.idx.nz) * 10,
    ones(3*N) .* (-Inf),
    ones(3*N) .* 0.0.
    z0,
    params,
    diff type;
    tol = 1e-6.
    c_{tol} = 1e-6
```

```
# return the trajectories
            x1 = [zeros(6) for _ = 1:N]
            x2 = [zeros(6) for _ = 1:N]
            x3 = [zeros(6) for = 1:N]
            u1 = [zeros(2) 	 for _ = 1:(N-1)]
            u2 = [zeros(2) 	 for _ = 1:(N-1)]
            u3 = [zeros(2) 	 for _ = 1:(N-1)]
            for i = 1:params.N
                x = Z[params.idx.x[i]]
                x1[i] = x[1:6]
                x2[i] = x[7:12]
                x3[i] = x[13:18]
                if i < params.N</pre>
                    u = Z[params.idx.u[i]]
                    u1[i] = u[1:2]
                    u2[i] = u[3:4]
                    u3[i] = u[5:6]
                end
            end
            return x1, x2, x3, u1, u2, u3, t_vec, params
        end
Out[]: quadrotor_reorient
In [ ]: @testset "quadrotor reorient" begin
            X1, X2, X3, U1, U2, U3, t_vec, params = quadrotor_reorient(verbose=true)
            #-----testing-----
            # check lengths of everything
            @test length(X1) == length(X2) == length(X3)
            @test length(U1) == length(U2) == length(U3)
            @test length(X1) == params.N
            @test length(U1) == (params.N-1)
```

max_iters=10_000,
verbose = true

```
# check for collisions
distances = [distance between quads(x1[1:2], x2[1:2], x3[1:2]) for (x1, x2, x3) in zip(X1, X2, X3)]
@test minimum(minimum.(distances)) >= 0.799
# check initial and final conditions
\operatorname{Otest} \operatorname{norm}(X1[1] - \operatorname{params.x1ic.} \operatorname{Inf}) <= 1e-3
\text{@test norm}(X2[1] - \text{params.x2ic, Inf}) \iff 1e-3
atest norm(X3[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 = "collision distance",
     xlabel = "time (s)", ylabel = "distance (m)", title = "Distance between Quadrotors")
display(plot!(t_vec, hcat(distances...)', label = ["|r_1 - r_2|" "|r_1 - r_3|" "|r_2 - r_2|"]))
X1m = hcat(X1...)
X2m = hcat(X2...)
X3m = hcat(X3...)
plot(X1m[1,:], X1m[2,:], color = :red,title = "Quadrotor Trajectories", label = "quad 1")
plot!(X2m[1,:], X2m[2,:], color = :green, label = "quad 2",xlabel = "p_x", ylabel = "p_z")
display(plot!(X3m[1,:], X3m[2,:], color = :blue, label = "quad 3"))
plot(t vec, X1m[3,:], color = :red,title = "Quadrotor Orientations", label = "quad 1")
plot!(t_vec, X2m[3,:], color = :green, label = "quad 2",xlabel = "time (s)", ylabel = "θ")
display(plot!(t vec, X3m[3,:], color = :blue, label = "quad 3"))
# plot U1, U2, U3
U1m = hcat(U1...)
```

```
U2m = hcat(U2...)
    U3m = hcat(U3...)
    plot(t_vec[1:end-1], U1m[1,:], color = :red,title = "Quadrotor Controls", label = "quad 1")
    plot!(t vec[1:end-1], U2m[1,:], color = :green, label = "quad 2",xlabel = "time (s)", ylabel = "u1")
    display(plot!(t vec[1:end-1], U3m[1,:], color = :blue, label = "quad 3"))
    plot(t vec[1:end-1], U1m[2,:], color = :red,title = "Quadrotor Controls", label = "quad 1")
    plot!(t_vec[1:end-1], U2m[2,:], color = :green, label = "quad 2",xlabel = "time (s)", ylabel = "u2")
    display(plot!(t vec[1:end-1], U3m[2,:], 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 3.14.14, running with linear solver MUMPS 5.6.2.
Number of nonzeros in equality constraint Jacobian...:
                                                     300348
Number of nonzeros in inequality constraint Jacobian.:
                                                      48204
Number of nonzeros in Lagrangian Hessian...:
                                                         0
Total number of variables....:
                                                       618
                   variables with only lower bounds:
                                                         0
              variables with lower and upper bounds:
                                                       618
                   variables with only upper bounds:
                                                         0
Total number of equality constraints....:
                                                       486
Total number of inequality constraints....:
                                                        78
       inequality constraints with only lower bounds:
                                                         0
  inequality constraints with lower and upper bounds:
                                                         0
       inequality constraints with only upper bounds:
                                                        78
       objective
                   inf_pr inf_du lg(mu) ||d|| lg(rg) alpha_du alpha_pr ls
iter
  0 5.4366000e+02 1.96e+00 8.00e+00 0.0 0.00e+00
                                                   - 0.00e+00 0.00e+00
  1 5.4475305e+02 1.93e+00 5.26e+03 -5.0 5.91e+00
                                                    - 4.54e-02 1.52e-02h 1
  2 5.4471199e+02 1.93e+00 1.81e+04 1.2 4.89e+03
                                                   - 1.30e-05 4.60e-06f 2
  3 5.4501683e+02 1.93e+00 2.10e+04 -1.1 4.62e+02
                                                    - 5.19e-03 5.91e-04h 1
  4 5.4515318e+02 1.93e+00 2.19e+04 -0.6 2.88e+01
                                                    - 3.11e-03 4.28e-04h 1
```

```
5 5.4564287e+02 1.93e+00 2.33e+04 -0.1 4.65e+01
                                                     - 2.80e-03 1.31e-03f 1
  6 5.5232401e+02 1.91e+00 4.36e+04 -0.2 3.87e+01
                                                     - 2.63e-03 9.42e-03f 1
  7 7.0106071e+02 1.79e+00 1.67e+05
                                    0.6 2.59e+01
                                                     - 4.52e-02 9.01e-02f 2
  8 4.3486358e+03 1.55e+00 2.18e+02
                                    0.4 8.63e+00
                                                     - 4.80e-01 1.00e+00h 1
  9 3.6398663e+03 1.01e+00 1.24e+02
                                    0.6 1.02e+01
                                                     - 5.17e-01 1.00e+00f 1
       objective
                   inf pr inf du lg(mu) ||d|| lg(rg) alpha du alpha pr ls
iter
 10 4.1491885e+03 3.03e-01 3.66e+01 -5.3 6.41e+00
                                                     - 6.25e-01 1.00e+00h 1
 11 4.0793823e+03 9.33e-02 2.06e+01 -1.0 3.48e+00
                                                     - 6.09e-01 8.89e-01f 1
 12 4.0786969e+03 3.22e-02 1.25e+01 -1.6 2.93e+00
                                                     - 4.79e-01 6.82e-01h 1
 13 4.0765822e+03 2.63e-02 1.02e+01 -1.1 2.01e+00
                                                     - 4.67e-01 1.78e-01f 1
 14 4.0753748e+03 1.90e-02 7.46e+00 -1.7 2.06e+00
                                                     - 3.00e-01 2.70e-01f 1
 15 4.0731490e+03 1.14e-02 2.02e+01 -0.5 4.34e+00
                                                     - 6.21e-02 4.48e-01f 1
 16 4.0720469e+03 9.51e-03 1.90e+01 -2.1 1.35e+00
                                                     - 4.30e-01 1.67e-01f 1
 17 4.0711893e+03 6.91e-03 1.31e+01 -6.9 9.98e-01
                                                     - 1.27e-01 2.85e-01h 1
 18 4.0708646e+03 5.84e-03 1.14e+01 -2.5 9.41e-01
                                                     - 3.94e-01 1.53e-01h 1
 19 4.0705068e+03 4.28e-03 9.58e+00 -1.9 1.16e+00
                                                     - 5.17e-01 2.65e-01f 1
iter
       objective
                   inf_pr inf_du lg(mu) ||d|| lg(rg) alpha_du alpha_pr ls
 20 4.0716478e+03 3.91e-03 4.54e+00 -1.8 3.11e-01
                                                     - 5.13e-01 1.00e+00h 1
 21 4.0716516e+03 2.88e-03 4.66e+00 -1.0 2.84e+00
                                                     - 5.19e-01 2.97e-01f 1
 22 4.0720863e+03 2.48e-03 5.44e+00 -1.3 5.83e-01
                                                     - 6.20e-01 1.00e+00f 1
 23 4.0713235e+03 1.03e-03 3.85e+00 -1.3 2.59e-01
                                                     - 1.00e+00 9.03e-01h 1
 24 4.0713377e+03 1.13e-04 1.78e+00 -2.4 3.95e-01
                                                     - 9.97e-01 9.95e-01h 1
 25 4.0712888e+03 6.72e-05 1.68e+00 -3.5 2.19e-01
                                                     - 1.00e+00 4.17e-01h 1
 26 4.0712816e+03 4.41e-05 6.37e-01 -4.0 1.27e-01
                                                     - 1.00e+00 7.04e-01h 1
 27 4.0712829e+03 1.61e-05 1.53e-01 -5.7 2.24e-02
                                                     - 1.00e+00 9.73e-01h 1
 28 4.0712886e+03 6.96e-07 5.77e-02 -7.4 7.00e-03
                                                     - 1.00e+00 9.85e-01h 1
 29 4.0712888e+03 1.98e-08 3.17e-02 -8.2 4.19e-03
                                                     - 1.00e+00 9.93e-01h 1
                   inf pr inf du lg(mu) ||d|| lg(rg) alpha du alpha pr ls
iter
       obiective
 30 4.0712888e+03 4.37e-09 8.39e-03 -10.1 4.00e-03
                                                     - 1.00e+00 9.99e-01h 1
 31 4.0712888e+03 1.74e-09 2.53e-04 -11.0 1.70e-04
                                                     - 1.00e+00 1.00e+00h 1
 32 4.0712888e+03 2.94e-10 2.83e-05 -11.0 1.03e-04
                                                     - 1.00e+00 1.00e+00h 1
 33 4.0712888e+03 1.51e-12 1.60e-05 -11.0 3.37e-05
                                                     - 1.00e+00 1.00e+00h 1
 34 4.0712888e+03 1.12e-12 4.62e-06 -11.0 2.89e-05
                                                     - 1.00e+00 1.00e+00h 1
 35 4.0712888e+03 7.64e-13 8.05e-06 -11.0 3.59e-06
                                                     - 1.00e+00 1.00e+00h 1
 36 4.0712888e+03 3.04e-13 6.73e-07 -11.0 2.09e-06
                                                     - 1.00e+00 1.00e+00h 1
```

Number of Iterations....: 36

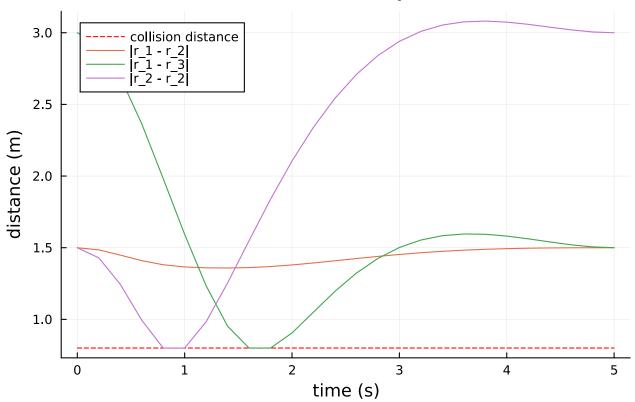
(scaled) (unscaled)
0bjective...... 4.0712887772704803e+03 4.0712887772704803e+03

```
Dual infeasibility....:
                            6.7271088196285582e-07
                                                     6.7271088196285582e-07
Constraint violation...:
                            3.0389579741552097e-13
                                                     3.0389579741552097e-13
Variable bound violation:
                            0.00000000000000000e+00
                                                     0.0000000000000000e+00
Complementarity....:
                           1.0000000000004423e-11
                                                     1.0000000000004423e-11
Overall NLP error...:
                           6.7271088196285582e-07
                                                     6.7271088196285582e-07
Number of objective function evaluations
                                                    = 41
Number of objective gradient evaluations
                                                    = 37
Number of equality constraint evaluations
                                                    = 41
Number of inequality constraint evaluations
                                                    = 41
Number of equality constraint Jacobian evaluations
                                                    = 37
Number of inequality constraint Jacobian evaluations = 37
Number of Lagrangian Hessian evaluations
                                                    = 0
                                                    = 8.687
Total seconds in IPOPT
```

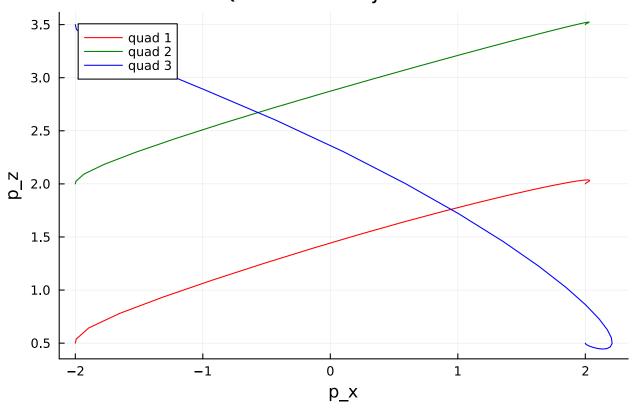
EXIT: Optimal Solution Found.

```
[ Info: Listening on: 127.0.0.1:8707, thread id: 1
r Info: MeshCat server started. You can open the visualizer by visiting the following URL in your browse
r:
http://127.0.0.1:8707
```

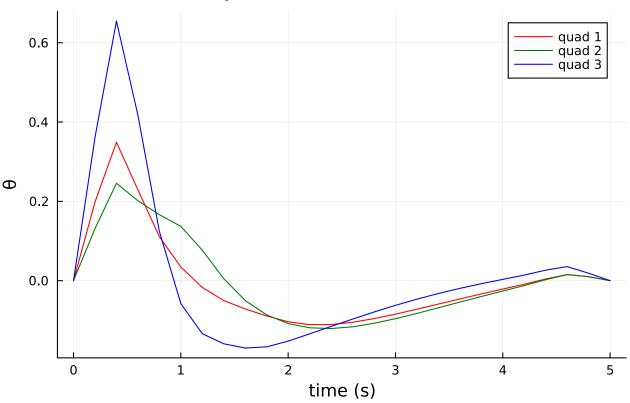
Distance between Quadrotors

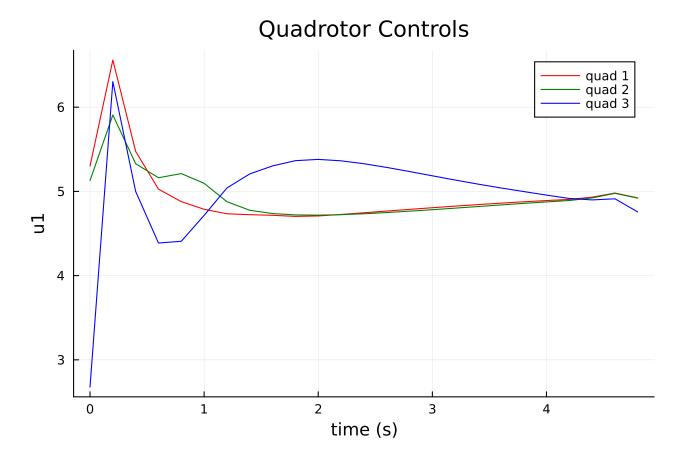


Quadrotor Trajectories

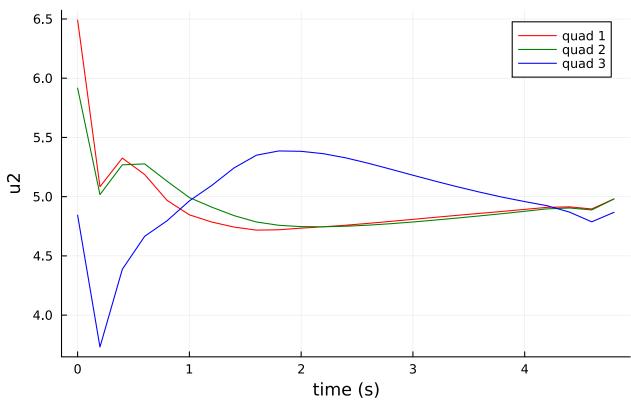


Quadrotor Orientations





Quadrotor Controls



Test Summary: | Pass Total Time quadrotor reorient | 14 14 10.1s

Out[]: Test.DefaultTestSet("quadrotor reorient", Any[], 14, false, false, true, 1.70924525430846e9, 1.709245264 373713e9, false)