NOTE: This question will have long outputs for each cell, remember you can use cell -> all output -> toggle scrolling to better see it all

Q1: Direct Collocation (DIRCOL) for a Cart Pole (30 pts)

We are now going to start working with the NonLinear Program (NLP) Solver IPOPT to solve some trajectory optimization problems. First we will demonstrate how this works for simple optimization problems (not trajectory optimization). The interface that we have setup for IPOPT is the following:

$$egin{array}{lll} \min_x & \ell(x) & ext{cost function} \ & ext{st} & c_{eq}(x) = 0 & ext{equality constraint} \ & c_L \leq c_{ineq}(x) \leq c_U & ext{inequality constraint} \ & x_L \leq x \leq x_U & ext{primal bound constraint} \ \end{array}$$

where $\ell(x)$ is our objective function, $c_{eq}(x)=0$ is our equality constraint, $c_L \leq c_{ineq}(x) \leq c_U$ is our bound inequality constraint, and $x_L \leq x \leq x_U$ is a bound constraint on our primal variable x.

Part A: Solve an LP with IPOPT (5 pts)

To demonstrate this, we are going to ask you to solve a simple Linear Program (LP):

$$egin{array}{ll} \min_x & q^T x \ & ext{st} & Ax = b \ & Gx \leq h \end{array}$$

Your job will be to transform this problem into the form shown above and solve it with IPOPT. To help you interface with IPOPT, we have created a function fmincon for you. Below is the docstring for this function that details all of the inputs.

```
In [3]:
        x = fmincon(cost, equality constraint, inequality constraint, x 1, x u, c 1, c u, x0,
        params,diff_type)
        This function uses IPOPT to minimize an objective function
        `cost(params, x)`
        With the following three constraints:
        `equality constraint(params, x) = 0`
         `c_l <= inequality_constraint(params, x) <= c_u`
        `x 1 <= x <= x u`
        Note that the constraint functions should return vectors.
        Problem specific parameters should be loaded into params::NamedTuple (things 1
        ike
        cost weights, dynamics parameters, etc.).
        args:
            cost::Function
                                               - objective function to be minimzed (ret
        urns scalar)
            equality_constraint::Function
                                               - c_eq(params, x) == 0
            inequality_constraint::Function - c_l <= c_ineq(params, x) <= c_u</pre>
            x_1::Vector
                                               - x_1 <= x <= x_u
                                               - x_1 <= x <= x_u
            x u::Vector
            c_l::Vector
                                               - c_1 <= c_ineq(params, x) <= c_u</pre>
                                               - c_l <= c_ineq(params, x) <= c_u</pre>
            c_u::Vector
            x0::Vector
                                               - initial guess
            params::NamedTuple

    problem parameters for use in costs/co

        nstraints
                                               - :auto for ForwardDiff, :finite for Fin
            diff_type::Symbol
        iteDiff
                                               - true for IPOPT output, false for nothi
            verbose::Bool
        ng
        optional args:
                                               - optimality tolerance
            tol
                                               - constraint violation tolerance
            c_tol
            max_iters
                                               - max iterations
            verbose

    verbosity of IPOPT

        outputs:
                                               - solution
            x::Vector
        You should try and use :auto for your `diff_type` first, and only use :finite
        absolutely cannot get ForwardDiff to work.
        This function will run a few basic checks before sending the problem off to IP
        solve. The outputs of these checks will be reported as the following:
        -----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-----

If you're getting stuck during the testing of one of the derivatives, try swit ching
to FiniteDiff.jl by setting diff_type = :finite.
""";
```

```
In [16]: @testset "solve LP with IPOPT" begin
             LP = jldopen(joinpath(@__DIR__,"utils","random_LP.jld2"))
             params = (q = LP["q"], A = LP["A"], b = LP["b"], G = LP["G"], h = LP["h"])
             # return a scalar
             function cost(params, x)::Real
                 # TODO: create cost function with params and x
                 return params.q'*x
             end
             # return a vector
             function equality constraint(params, x)::Vector
                 # TODO: create equality constraint function with params and x
                 return params.A*x - params.b
             end
             # return a vector
             function inequality constraint(params, x)::Vector
                 \# TODO: create inequality constraint function with params and x
                 return params.G*x - params.h
             end
             # TODO: primal bounds
             # you may use Inf, like Inf*ones(10) for a vector of positive infinity
             x l = -Inf*ones(20)
             x_u = Inf*ones(20)
             # TODO: inequality constraint bounds
             c 1 = -Inf*ones(20)
             c_u = zeros(20)
             # initial guess
             x0 = randn(20)
             diff type = :auto # use ForwardDiff.jl
               diff_type = :finite # use FiniteDiff.jl
             x = fmincon(cost, equality_constraint, inequality_constraint,
                         x_l, x_u, c_l, c_u, x0, params, diff_type;
                         tol = 1e-6, c tol = 1e-6, max iters = 10 000, verbose = true);
             @test isapprox(x, [-0.44289, 0, 0, 0.19214, 0, 0, -0.109095,
                                 -0.43221, 0, 0, 0.44289, 0, 0, 0.192142,
                                 0, 0, 0.10909, 0.432219, 0, 0], atol = 1e-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...:
                                                         80
Number of nonzeros in inequality constraint Jacobian.:
                                                        400
Number of nonzeros in Lagrangian Hessian....:
                                                          0
Total number of variables....:
                                                         20
                   variables with only lower bounds:
                                                          0
               variables with lower and upper bounds:
                                                          0
                   variables with only upper bounds:
                                                          0
Total number of equality constraints....:
                                                          4
Total number of inequality constraints....:
                                                         20
       inequality constraints with only lower bounds:
                                                          0
  inequality constraints with lower and upper bounds:
                                                          0
       inequality constraints with only upper bounds:
                                                         20
iter
       objective
                   inf pr
                           inf_du lg(mu) ||d|| lg(rg) alpha_du alpha_pr
1s
  0 -2.4542136e-01 5.54e+00 3.33e-01
                                     0.0 0.00e+00
                                                       0.00e+00 0.00e+00
0
                                                       4.08e-01 8.10e-01f
  1 4.5394613e+00 1.05e+00 1.51e+00 -0.3 1.78e+00
1
  2 3.0658391e+00 2.78e-16 6.56e-08 -2.0 8.37e-01
                                                       1.00e+00 1.00e+00h
1
  3 1.3978934e+00 2.22e-16 8.61e-08 -1.9 5.05e-01
                                                      9.33e-01 9.97e-01f
1
  4 1.2015779e+00 2.22e-16 6.34e-09 -3.2 7.99e-02
                                                       9.75e-01 9.26e-01f
1
  5 1.1767157e+00 2.22e-16 1.04e-10 -4.9 1.12e-02
                                                      1.00e+00 9.91e-01f
1
  6 1.1763495e+00 2.22e-16 3.17e-13 -10.6 1.98e-04
                                                    - 1.00e+00 1.00e+00f
1
Number of Iterations....: 6
                                 (scaled)
                                                        (unscaled)
Objective....:
                          1.1763494775164158e+00
                                                  1.1763494775164158e+00
Dual infeasibility....:
                          3.1707969583294471e-13
                                                  3.1707969583294471e-13
Constraint violation...:
                          2.2204460492503131e-16
                                                  2.2204460492503131e-16
Variable bound violation:
                          0.00000000000000000e+00
                                                  0.0000000000000000e+00
Complementarity....:
                          1.4344358625806248e-08
                                                  1.4344358625806248e-08
Overall NLP error....:
                          1.4344358625806248e-08
                                                  1.4344358625806248e-08
Number of objective function evaluations
                                                 = 7
Number of objective gradient evaluations
                                                 = 7
Number of equality constraint evaluations
Number of inequality constraint evaluations
Number of equality constraint Jacobian evaluations
                                                 = 7
Number of inequality constraint Jacobian evaluations = 7
```

Number of Lagrangian Hessian evaluations = 0 Total seconds in IPOPT = 0.014

EXIT: Optimal Solution Found.

Test.DefaultTestSet("solve LP with IPOPT", Any[], 1, false, false)

Part B: Cart Pole Swingup (20 pts)

We are now going to solve for a cartpole swingup. The state for the cartpole is the following:

$$x = [p, heta, \dot{p}, \dot{ heta}]^T$$

Where p and heta can be seen in the graphic <code>cartpole.png</code> .



where we start with the pole in the down position ($\theta = 0$), and we want to use the horizontal force on the cart to drive the pole to the up position ($\theta = \pi$).

$$egin{aligned} \min_{x_{1:N},u_{1:N-1}} && \sum_{i=1}^{N-1} \left[rac{1}{2}(x_i-x_{goal})^TQ(x_i-x_{goal}) + rac{1}{2}u_i^TRu_i
ight] + rac{1}{2}(x_N-x_{goal})^TQ_f(x_N-x_{goal}) \ & ext{st} && x_1 = x_{ ext{IC}} \ && x_N = x_{goal} \ && f_{hs}(x_i,x_{i+1},u_i,dt) = 0 \quad ext{for } i=1,2,\ldots,N-1 \ && -10 \leq u_i \leq 10 \quad ext{for } i=1,2,\ldots,N-1 \end{aligned}$$

Where $x_{IC}=[0,0,0,0]$, and $x_{goal}=[0,\pi,0,0]$, and $f_{hs}(x_i,x_{i+1},u_i)$ is the implicit integrator residual for Hermite Simpson (see HW1Q1 to refresh on this). Note that while Zac used a first order hold (FOH) on the controls in class (meaning we linearly interpolate controls between time steps), we are using a zero-order hold (ZOH) in this assignment. This means that each control u_i is held constant for the entirety of the timestep.

```
In [54]: # cartpole
          function dynamics(params::NamedTuple, x::Vector, u)
              # cartpole ODE, parametrized by params.
              # cartpole physical parameters
              mc, mp, 1 = params.mc, params.mp, params.1
              g = 9.81
              q = x[1:2]
              qd = x[3:4]
              s = sin(q[2])
              c = cos(q[2])
              H = [mc+mp mp*1*c; mp*1*c mp*1^2]
              C = [0 -mp*qd[2]*1*s; 0 0]
              G = [0, mp*g*1*s]
              B = [1, 0]
              qdd = -H \setminus (C*qd + G - B*u[1])
              xdot = [qd;qdd]
               return xdot
          function hermite_simpson(params::NamedTuple, x1::Vector, x2::Vector, u, dt::Re
          al)::Vector
              # TODO: input hermite simpson implicit integrator residual
              \dot{x}_k = dynamics(params, x1, u)
              \dot{x}_{kp1} = dynamics(params, x2, u)
              x_{kpm} = 1/2*(x1 + x2) + dt/8*(\dot{x}_k - \dot{x}_{kp1})
              \dot{x}_{kpm} = dynamics(params, x_{kpm}, u)
              res = x1 + dt/6*(\dot{x}_k + 4*\dot{x}_kpm + \dot{x}_kp1) - x2
               return res
          end
```

hermite_simpson (generic function with 1 method)

To solve this problem with IPOPT and $\$ fmincon , we are going to concatenate all of our x's and u's into one vector:

$$Z = egin{bmatrix} x_1 \ u_1 \ x_2 \ u_2 \ dots \ x_{N-1} \ u_{N-1} \ x_N \end{bmatrix} \in \mathbb{R}^{N \cdot nx + (N-1) \cdot nu}$$

where $x \in \mathbb{R}^{nx}$ and $u \in \mathbb{R}^{nu}$. Below we will provide useful indexing guide in <code>create_idx</code> to help you deal with Z.

It is also worth noting that while there are inequality constraints present ($-10 \le u_i \le 10$), we do not need a specific inequality_constraints function as an input to fmincon since these are just bounds on the primal (Z) variable. You should use primal bounds in fmincon to capture these constraints.

```
In [60]: 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
         function cartpole_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]]
                  J += 1 / 2 *( (xi - xg)' * Q * (xi - xg) + ui' * R * ui)
             end
             xn = Z[idx.x[N]]
             # dont forget terminal cost
             J += 1 / 2 * (xn - xg)' * Qf * (xn - xg)
             return J
         end
         function cartpole_dynamics_constraints(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 HW0)
             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]] = hermite simpson(params, xi, xip1, ui, dt)
             end
```

```
return c
end
function cartpole_equality_constraint(params::NamedTuple, Z::Vector)::Vector
   N, idx, xic, xg = params.N, params.idx, params.xic, params.xg
   # TODO: return all of the equality constraints
   x0 = Z[idx.x[1]]
   xN = Z[idx.x[N]]
   eq_cons = [cartpole_dynamics_constraints(params, Z); x0 - xic; xN - xg]
   return eq cons
end
function solve cartpole swingup(;verbose=true)
   # problem size
   nx = 4
   nu = 1
   dt = 0.05
   tf = 2.0
   t vec = 0:dt:tf
   N = length(t_vec)
   # LOR cost
   Q = diagm(ones(nx))
   R = 0.1*diagm(ones(nu))
   Qf = 10*diagm(ones(nx))
   # indexing
   idx = create idx(nx,nu,N)
   # initial and goal states
   xic = [0, 0, 0, 0]
   xg = [0, pi, 0, 0]
   # load all useful things into params
   idx,mc = 1.0, mp = 0.2, 1 = 0.5
   # TODO: primal bounds
   x l = -Inf*ones(idx.nz)
   x_u = Inf*ones(idx.nz)
   for i = 1:(N-1)
       x_1[idx.u[i]] = -10
       x_u[idx.u[i]] = 10
   # inequality constraint bounds (this is what we do when we have no inequal
ity constraints)
   c_1 = zeros(0)
   c u = zeros(0)
   function inequality_constraint(params, Z)
       return zeros(eltype(Z), 0)
   end
```

```
# initial quess
    z0 = 0.001*randn(idx.nz)
    # choose diff type (try :auto, then use :finite if :auto doesn't work)
    diff_type = :auto
    # diff_type = :finite
    Z = fmincon(cartpole cost, cartpole equality constraint, inequality constrai
nt,
               x_1,x_u,c_1,c_u,z0,params, diff_type;
               tol = 1e-6, c tol = 1e-6, max iters = 10 000, verbose = verbos
e)
    \# pull the X and U solutions out of Z
   X = [Z[idx.x[i]]  for i = 1:N]
   U = [Z[idx.u[i]]  for i = 1:(N-1)]
    return X, U, t vec, params
end
@testset "cartpole swingup" begin
   X, U, t_vec = solve_cartpole_swingup(verbose=true)
    # -----testing-----
   @test isapprox(X[1],zeros(4), atol = 1e-4)
   @test isapprox(X[end], [0,pi,0,0], atol = 1e-4)
   Xm = hcat(X...)
   Um = hcat(U...)
    # -----plotting-----
    display(plot(t vec, Xm', label = ["p" "\theta" "\theta" "\theta"], xlabel = "time (s)", t
itle = "State Trajectory"))
    display(plot(t_vec[1:end-1],Um',label="",xlabel = "time (s)", ylabel =
"u",title = "Controls"))
    # meshcat animation
    display(animate_cartpole(X, 0.05))
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...:
                                                       34272
Number of nonzeros in inequality constraint Jacobian.:
                                                          0
Number of nonzeros in Lagrangian Hessian....:
                                                          0
Total number of variables....:
                                                         204
                   variables with only lower bounds:
                                                          0
               variables with lower and upper bounds:
                                                          40
                   variables with only upper bounds:
                                                          0
Total number of equality constraints....:
                                                         168
Total number of inequality constraints....:
                                                          0
       inequality constraints with only lower bounds:
                                                          0
  inequality constraints with lower and upper bounds:
                                                          0
       inequality constraints with only upper bounds:
                                                          0
iter
       objective
                   inf_pr
                            inf_du lg(mu) ||d|| lg(rg) alpha_du alpha_pr
ls
                                     0.0 0.00e+00
                                                       0.00e+00 0.00e+00
     2.4670591e+02 3.14e+00 3.38e-04
a
    2.7497345e+02 2.38e+00 7.99e+00 -5.0 1.28e+01
                                                       4.90e-01 2.43e-01h
3
    2.9802689e+02 2.16e+00 1.03e+01
                                    -0.5 1.05e+01
                                                        6.11e-01 9.26e-02h
  2
4
     3.3422214e+02 1.87e+00 1.40e+01
                                    -0.4 1.29e+01
                                                       6.48e-01 1.33e-01h
3
  4 3.7116777e+02 1.61e+00 2.08e+01
                                    -0.5 1.19e+01
                                                       8.80e-01 1.40e-01h
3
  5 4.1966233e+02 1.33e+00 2.73e+01
                                    -0.8 1.00e+01
                                                       1.00e+00 1.74e-01h
3
  6 4.4384515e+02 1.20e+00 3.19e+01
                                     0.3 1.84e+01
                                                      6.33e-01 9.62e-02h
3
    4.7565058e+02 1.07e+00 3.53e+01
                                     0.2 1.80e+01
                                                       6.44e-01 1.12e-01h
  7
3
     5.1170702e+02 9.43e-01 3.89e+01
                                      0.3 2.25e+01
                                                       6.13e-01 1.17e-01h
3
     5.2136616e+02 8.54e-01 3.84e+01
                                      0.3 1.15e+01
                                                       8.80e-01 9.51e-02h
3
                            inf_du lg(mu) ||d|| lg(rg) alpha_du alpha_pr
iter
       objective
                   inf pr
ls
     5.1537777e+02 7.71e-01 4.12e+01
                                      0.4 2.61e+01
                                                       5.16e-01 9.70e-02f
 10
3
     5.0928663e+02 7.01e-01 4.40e+01
                                      0.5 2.67e+01
                                                       6.00e-01 9.08e-02f
 11
3
                                     0.4 3.52e+01
                                                       8.46e-01 1.07e-01f
     5.0689528e+02 6.26e-01 4.91e+01
 12
3
 13
     5.0854663e+02 5.59e-01 5.71e+01
                                     0.6 2.68e+01
                                                     - 3.34e-01 1.06e-01h
3
     5.3536126e+02 3.51e-01 7.08e+01
                                     0.4 1.99e+01
                                                       1.94e-01 3.71e-01h
1
```

```
5.3510308e+02 2.75e-01 7.28e+01
                                       0.2 1.78e+01
                                                     - 2.68e-01 2.18e-01h
     5.4198179e+02 1.90e-01 7.57e+01
                                       0.7 1.63e+01
                                                       - 3.88e-01 3.10e-01f
  16
1
     5.3767995e+02 1.18e-01 8.54e+01
                                       0.6 1.18e+01
                                                     - 7.65e-01 3.75e-01h
  17
1
 18
     5.4058740e+02 8.77e-02 7.83e+01
                                       0.6 9.64e+00
                                                       - 8.00e-01 5.53e-01h
     5.2854119e+02 9.22e-02 5.78e+01
                                       0.4 6.89e+00
                                                          8.16e-01 9.68e-01h
  19
1
                    inf_pr inf_du lg(mu) ||d|| lg(rg) alpha_du alpha_pr
iter
       objective
ls
     5.0257874e+02 3.98e-02 2.14e+01
                                       0.1 2.40e+00
                                                       - 9.70e-01 1.00e+00f
  20
  21
     4.8262126e+02 6.14e-02 2.48e+01
                                       0.1 1.02e+01
                                                       - 5.26e-01 3.36e-01f
1
     4.6958639e+02 4.33e-02 1.77e+01
                                     -0.1 5.13e+00
                                                       - 9.22e-01 4.65e-01f
  22
1
     4.6319795e+02 2.27e-01 4.29e+01 -0.1 3.13e+01
                                                       - 3.04e-01 2.92e-01f
  23
     4.6212204e+02 1.64e-01 5.90e+01
                                     0.4 1.92e+01
                                                       - 1.00e+00 3.20e-01f
1
     4.4562409e+02 7.11e-03 2.81e+01 -0.1 2.79e+00
                                                       - 9.95e-01 1.00e+00f
  25
1
  26
     4.4067786e+02 5.60e-03 2.00e+01 -0.9 1.73e+00
                                                       - 1.00e+00 1.00e+00f
  27
     4.3810335e+02 2.40e-03 1.89e+01
                                     -1.5 1.35e+00
                                                          1.00e+00 1.00e+00f
     4.3719618e+02 2.93e-04 1.78e+01 -2.2 7.29e-01
                                                          1.00e+00 1.00e+00f
  28
1
     4.3661795e+02 3.67e-03 1.85e+01 -2.8 1.44e+00
                                                          1.00e+00 1.00e+00f
1
                    inf pr inf du lg(mu) ||d|| lg(rg) alpha du alpha pr
iter
       objective
ls
 30
     4.3679242e+02 4.64e-02 1.70e+01 -0.8 1.19e+01
                                                          2.11e-01 7.18e-01f
     4.3235332e+02 4.23e-02 1.37e+01 -1.4 4.47e+00
                                                          1.00e+00 1.00e+00f
  31
                                                       - 1.00e+00 1.00e+00f
     4.3300423e+02 2.46e-02 1.33e+01 -0.7 3.46e+00
  32
     4.3352277e+02 8.08e-02 2.34e+01 -0.4 2.45e+01
                                                       - 6.78e-01 2.71e-01f
  33
2
     4.3329408e+02 8.96e-02 3.59e+01 -0.6 2.24e+01
                                                       - 4.88e-01 1.49e-01f
2
     4.2755856e+02 2.78e-02 2.94e+01 -0.6 3.58e+00
                                                       - 6.04e-01 9.45e-01f
  35
     4.2186006e+02 2.43e-03 3.43e+01 -1.2 3.14e+00
                                                       - 9.98e-01 1.00e+00f
  36
     4.2070840e+02 2.37e-02 1.74e+01
                                      -1.2 5.78e+00
                                                         9.91e-01 1.00e+00f
  37
                                                       - 1.00e+00 1.00e+00f
     4.1983045e+02 3.34e-03 1.87e+01 -0.8 2.37e+00
  38
1
 39
     4.2080422e+02 6.51e-03 2.67e+01 -0.2 4.65e+00
                                                       - 1.00e+00 1.00e+00F
1
       objective
                             inf_du lg(mu) ||d|| lg(rg) alpha_du alpha_pr
iter
                    inf pr
ls
     4.2102762e+02 2.07e-02 4.64e+01 -0.4 9.54e+00
                                                       - 1.00e+00 1.00e+00f
```

```
1
  41 4.1622536e+02 7.64e-03 2.07e+01 -0.4 2.90e+00
                                                          1.00e+00 1.00e+00f
1
  42 4.1496809e+02 3.39e-04 7.18e+00 -0.5 7.99e-01
                                                      - 9.46e-01 1.00e+00F
1
     4.1473592e+02 5.49e-03 1.12e+01 -0.6 2.07e+01
                                                      - 8.78e-01 7.06e-02f
  43
     4.1107889e+02 1.20e-02 2.64e+01 -0.3 5.59e+00
                                                        7.96e-01 1.00e+00F
     4.1170294e+02 1.33e-02 4.67e+01 -0.7 1.42e+01
                                                    - 8.77e-01 8.22e-01H
1
     4.1282382e+02 2.37e-02 2.93e+01 -0.4 5.86e+00
                                                          1.00e+00 1.00e+00f
  46
     4.0952613e+02 1.26e-03 2.04e+01 -1.4 6.50e-01
                                                       - 9.99e-01 1.00e+00f
     4.0858592e+02 8.00e-04 1.72e+01 -2.3 5.47e-01
                                                      - 9.99e-01 9.99e-01f
1
     4.0814405e+02 6.55e-04 2.20e+01 -1.5 2.76e+00
                                                      - 1.00e+00 1.00e+00f
1
iter
       objective
                    inf pr inf du lg(mu) ||d|| lg(rg) alpha du alpha pr
     4.0742372e+02 1.66e-04 1.65e+01 -2.5 6.46e-01
                                                          1.00e+00 9.92e-01f
     4.0224491e+02 1.33e-02 1.79e+01 -1.9 1.03e+01
                                                      - 1.00e+00 1.00e+00F
                                                       - 8.76e-01 1.00e+00f
     4.0836843e+02 8.95e-03 9.52e+00 -0.5 3.88e+00
     4.0447157e+02 9.07e-03 1.32e+01 -0.6 2.76e+00
                                                      - 1.00e+00 1.00e+00f
  53
1
     4.0215789e+02 2.13e-03 1.03e+01 -0.6 1.83e+00
  54
                                                      - 7.02e-01 1.00e+00F
     4.0175664e+02 5.30e-03 1.67e+01 -1.0 1.50e+01
                                                          9.94e-01 1.28e-01f
     4.0271557e+02 3.07e-02 4.16e+01 -0.3 2.16e+01
                                                      - 1.00e+00 7.82e-01F
  57
     4.0209456e+02 3.20e-02 2.33e+01 -0.4 6.08e+00
                                                      - 1.00e+00 1.00e+00f
     4.0022304e+02 2.40e-04 1.53e+01 -0.4 4.00e-01
                                                      - 1.00e+00 1.00e+00f
1
     3.9850002e+02 4.16e-04 1.06e+01 -0.8 1.04e+00
                                                         9.96e-01 1.00e+00f
1
                    inf_pr inf_du lg(mu) ||d|| lg(rg) alpha_du alpha_pr
iter
       objective
     3.9607492e+02 3.71e-03 1.86e+01 -1.5 3.45e+00
                                                          1.00e+00 1.00e+00F
    3.9714072e+02 3.03e-03 3.53e+01 -1.0 6.86e+00
                                                      - 1.00e+00 1.00e+00H
1
 62
     3.9517332e+02 8.75e-04 1.80e+01 -1.1 2.42e+00
                                                      - 9.98e-01 1.00e+00f
1
     3.9474061e+02 7.54e-04 3.13e+00 -1.6 7.75e-01
                                                          1.00e+00 1.00e+00f
     3.9460731e+02 1.88e-04 3.18e+00 -1.9 4.96e-01
                                                         1.00e+00 1.00e+00f
  64
     3.9434419e+02 3.03e-03 1.25e+01 -2.4 2.90e+00
                                                      - 1.00e+00 1.00e+00F
  65
     3.9603257e+02 3.32e-03 2.25e+01 -0.9 3.26e+01
                                                       - 1.00e+00 1.21e-01f
```

2

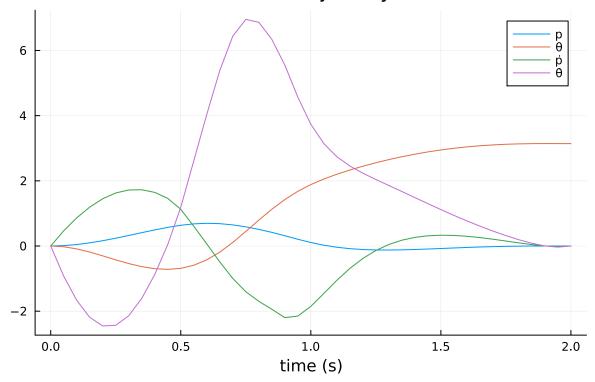
```
3.9449419e+02 4.37e-03 2.67e+01 -1.1 1.59e+00
                                                      - 1.00e+00 1.00e+00f
     3.9436426e+02 1.05e-04 2.17e+01 -1.1 4.98e-01
                                                        1.00e+00 1.00e+00f
  68
1
 69 3.9397136e+02 3.46e-03 1.41e+01 -1.6 1.58e+00
                                                      - 1.00e+00 1.00e+00f
1
                    inf pr
                             inf_du lg(mu) ||d|| lg(rg) alpha_du alpha_pr
iter
       objective
ls
    3.9354696e+02 3.95e-05 7.53e+00 -2.0 1.07e+00
                                                         1.00e+00 1.00e+00f
  70
1
 71
     3.9347592e+02 1.20e-05 3.47e+00 -2.5 2.28e-01
                                                         1.00e+00 1.00e+00f
1
  72 3.9345658e+02 4.66e-05 1.70e-01 -3.5 1.55e-01
                                                       - 1.00e+00 1.00e+00f
1
 73 3.9344871e+02 4.24e-06 3.31e-02 -5.1 5.19e-02
                                                      - 1.00e+00 9.95e-01h
1
 74 3.9344834e+02 1.26e-07 3.82e-04 -6.8 1.14e-02
                                                         1.00e+00 9.98e-01h
1
 75 3.9344834e+02 4.69e-08 1.33e-04 -8.9 5.68e-03
                                                         1.00e+00 1.00e+00h
1
  76 3.9344834e+02 2.58e-11 2.09e-05 -11.0 2.84e-04
                                                      - 1.00e+00 1.00e+00h
1
  77 3.9344834e+02 2.84e-12 3.34e-06 -11.0 7.93e-05
                                                      - 1.00e+00 1.00e+00h
1
 78 3.9344834e+02 1.39e-13 1.71e-06 -11.0 2.18e-05
                                                      - 1.00e+00 1.00e+00h
  79 3.9344834e+02 2.84e-14 2.51e-07 -11.0 9.49e-06
                                                      - 1.00e+00 1.00e+00h
1
Number of Iterations...: 79
                                  (scaled)
                                                           (unscaled)
Objective....:
                           3.9344833576222720e+02
                                                     3.9344833576222720e+02
Dual infeasibility....:
                           2.5090049761140400e-07
                                                     2.5090049761140400e-07
Constraint violation...:
                           2.8421709430404007e-14
                                                     2.8421709430404007e-14
Variable bound violation:
                           9.9997231828297117e-08
                                                     9.9997231828297117e-08
Complementarity....:
                           1.0000652361558553e-11
                                                     1.0000652361558553e-11
Overall NLP error....:
                           2.5090049761140400e-07
                                                     2.5090049761140400e-07
Number of objective function evaluations
                                                    = 172
Number of objective gradient evaluations
                                                    = 80
Number of equality constraint evaluations
                                                     172
Number of inequality constraint evaluations
                                                    = 0
Number of equality constraint Jacobian evaluations
                                                    = 80
Number of inequality constraint Jacobian evaluations = 0
Number of Lagrangian Hessian evaluations
                                                    = 0
```

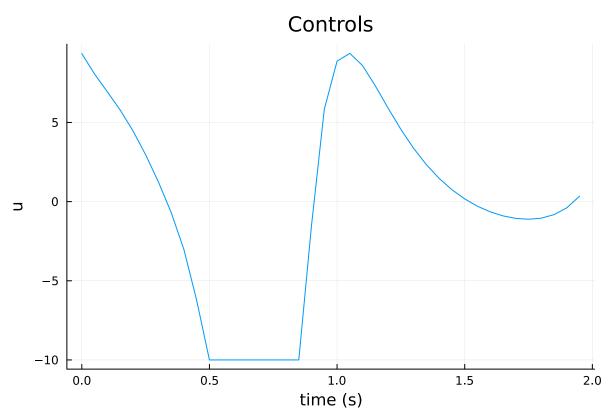
= 1.949

EXIT: Optimal Solution Found.

Total seconds in IPOPT

State Trajectory





Info: Listening on: 127.0.0.1:8720, thread id: 1

@ HTTP.Servers /root/.julia/packages/HTTP/enKbm/src/Servers.jl:369

Info: MeshCat server started. You can open the visualizer by visiting the following URL in your browser:

| http://127.0.0.1:8720

- @ MeshCat /root/.julia/packages/MeshCat/QXID5/src/visualizer.jl:64

Part C: Track DIRCOL Solution (5 pts)

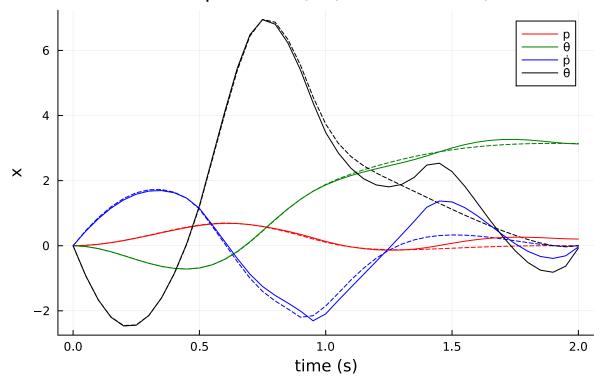
Now, similar to HW2 Q2 Part C, we are taking a solution X and U from DIRCOL, and we are going to track the trajectory with TVLQR to account for model mismatch. While we used hermite-simpson integration for the dynamics constraints in DIRCOL, we are going to use RK4 for this simulation. Remember to clamp your control to be within the control bounds.

10

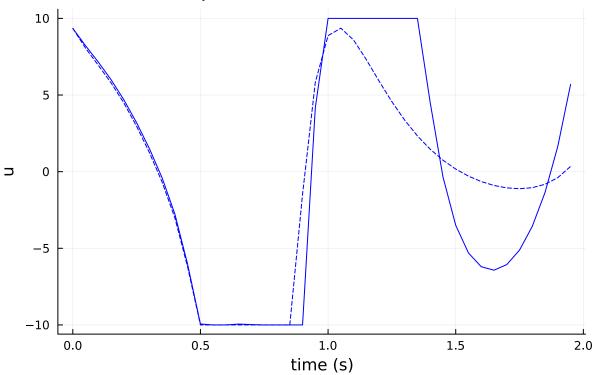
```
In [69]: | function rk4(params::NamedTuple, x::Vector,u,dt::Float64)
             # vanilla RK4
             k1 = dt*dynamics(params, x, u)
             k2 = dt*dynamics(params, x + k1/2, u)
             k3 = dt*dynamics(params, x + k2/2, u)
             k4 = dt*dynamics(params, x + k3, u)
             x + (1/6)*(k1 + 2*k2 + 2*k3 + k4)
         end
         @testset "track cartpole swingup with TVLQR" begin
             X dircol, U dircol, t vec, params dircol = solve cartpole swingup(verbose
         = false)
             N = length(X_dircol)
             dt = params_dircol.dt
             x0 = X_dircol[1]
             # TODO: use TVLQR to generate K's
             # use this for TVLQR tracking cost
             Q = diagm([1,1,.05,.1])
             Qf = 100*Q
             R = 0.01*diagm(ones(1))
             nx = 4
             nu = 1
             P = Of
             Ks = [zeros(nx) for i = 1:N-1]
             Dc = zeros(nx+nu,nx+nu)
             u = 0
             for ii = N:-1:2
                  # convert to dicrete dynamics
                 Ac = ForwardDiff.jacobian( x -> dynamics(params dircol, x, U dircol[i
         i-1]), X_dircol[ii])
                  Bc = ForwardDiff.jacobian( u -> dynamics(params dircol, X dircol[ii],
         _u), U_dircol[ii-1])
                  Dc[1:nx, 1:nx+nu] = [Ac Bc]
                  Dd = exp(Dc*dt)
                 A, B = (Dd[1:nx, 1:nx], Dd[1:nx, (nx+1):(nx+nu)])
                  # riccati recursion
                 K = (R + B'*P*B) \setminus B'*P*A
                 Ks[ii-1] = vec(K)
                  P = Q + K'*R*K + (A-B*K)'*P*(A-B*K)
             end
             # simulation
             Xsim = [zeros(4) for i = 1:N]
             Usim = [zeros(1) for i = 1:(N-1)]
             Xsim[1] = 1*x0
             # here are the real parameters (different than the one we used for DIRCOL)
             # this model mismatch is what's going to require the TVLQR controller to t
```

```
rack
   # the trajectory successfully.
   params_real = (mc = 1.05, mp = 0.21, l = 0.48)
   # TODO: simulate closed loop system with both feedforward and feedback con
trol
   # feedforward - the U dircol controls that we solved for using dircol
   # feedback - the TVLQR controls
   for i = 2:N
       # add controller and simulation step
       Usim[i-1] = clamp.(U_dircol[i-1]-[Ks[i-1]\cdot(Xsim[i-1] - X_dircol[i-1]))
1])], -10, 10)
       Xsim[i] = rk4(params real, Xsim[i-1],Usim[i-1], dt)
   end
   # -----testing-----
   xn = Xsim[N]
   @test norm(xn)>0
   @test 1e-6<norm(xn - X dircol[end])<.8</pre>
   @test abs(abs(rad2deg(xn[2])) - 180) < 5 # within 5 degrees</pre>
   @test maximum(norm.(Usim,Inf)) <= (10 + 1e-3)</pre>
   # -----plotting-----
   Xm = hcat(Xsim...)
   Xbarm = hcat(X dircol...)
   plot(t_vec,Xbarm',ls=:dash, label = "",lc = [:red :green :blue :black])
   display(plot!(t_vec,Xm',title = "Cartpole TVLQR (-- is reference)",
                xlabel = "time (s)", ylabel = "x",
                label = ["p" "\theta" "p" "\theta"], lc = [:red : green : blue : black]))
   Um = hcat(Usim...)
   Ubarm = hcat(U dircol...)
   plot(t vec[1:end-1],Ubarm',ls=:dash,lc = :blue, label = "")
   display(plot!(t_vec[1:end-1],Um',title = "Cartpole TVLQR (-- is referenc
e)",
                xlabel = "time (s)", ylabel = "u", lc = :blue, label = ""))
   # -----animate-----
   display(animate_cartpole(Xsim, 0.05))
end
```

Cartpole TVLQR (-- is reference)



Cartpole TVLQR (-- is reference)



Info: Listening on: 127.0.0.1:8722, thread id: 1
 @ HTTP.Servers /root/.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://127.0.0.1:8722

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

Test.DefaultTestSet("track cartpole swingup with TVLQR", Any[], 4, false, fal se)

1.