

# Q2e\_OptimalControl

April 16, 2024

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[ ]: using JuMP
      using Ipopt
      using Plots
      include("Q2c_VehicleDynamics.jl")

x0 = [-10.0, 0.0, 0.0, 0.0, 0.0, 10.0, 0.0] #TODO Initial Condition
XL = [-40, -20, -3, -pi/5, -pi/2, 5.0, -pi/12] # States Lower Bound
XU = [300, 20, 3, pi/5, pi/2, 15.0, pi/12] #TODO States Upper Bound
CL = [-2.6, -0.1] #TODO Control Lower Bound
CU = [2.6, 0.1] #TODO Control Upper Bound

model = Model(optimizer_with_attributes(Ipopt.Optimizer)) # Initialize JuMP
↳model

numStates = 7 #TODO number of states
numControls = 2 #TODO number of control
PredictionHorizon = 8 #TODO Prediction Time
numColPoints = 81 #TODO
Δt = PredictionHorizon/(numColPoints - 1)# Time interval

@variables(model, begin
    # Set xst as a numColPoints x numStates matrix that is between the upper
↳and lower states bounds
    XL[i] xst[j in 1:numColPoints, i in 1:numStates] XU[i]
    #TODO Similarly, set u as a numColPoints x numControls matrix that is
    # between the upper and lower control bounds
    CL[i] u[j in 1:numColPoints, i in 1:numControls] CU[i]
end)

# Fix initial conditions
fix(xst[1, 1], x0[1]; force = true) # set the initial condition for x-position
↳value
#TODO Follow the same way, set the remaining initial conditions,
# set x0[2] to xst[1,2],... and so on.
fix(xst[1,2],x0[2];force = true)
fix(xst[1,3],x0[3];force = true)
fix(xst[1,4],x0[4];force= true)
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fix(xst[1,5],x0[5];force= true)
fix(xst[1,6],x0[6];force= true)
fix(xst[1,7],x0[7];force= true)

# sa means steering angle, sr means steering rate
x = xst[:, 1]; y = xst[:, 2]; v = xst[:, 3]; r = xst[:, 4];  = xst[:, 5];
ux = xst[:, 6]; sa = xst[:, 7];
ax = u[:, 1]; # retract variable
sr = u[:, 2];

# xst = Matrix{Any}(undef, numColPoints, numStates)
# write the states derivative for all states & controls
xst = Matrix{Any}(undef, numColPoints, numStates)
for i = 1:1:numColPoints
    xst[i, :] = @expression(model, VehicleDynamics(xst[i, :], u[i, :]))
    # xst[i, :] = @expression(model, VehicleDynamics(reshape(xst[i, :],(1,7)),
    # reshape(u[i, :],(1,2))))
end

# add constraint to each state using backward Euler method
for j = 2:numColPoints
    for i = 1:numStates
        @constraint(model, xst[j, i] == xst[j - 1, i] +Δt * xst[j, i])
    end
end

# TODO write the cost function for each term - Lane change
y_cost = @expression(model, sum((y[j] - 5)^2 * Δt for j= 1:1:numColPoints))
    ↪#global y position of C.G Cost
sr_cost = @expression(model, sum((sr[j])^2 * Δt for j= 1:1:numColPoints))
sa_cost = @expression(model, sum((sa[j])^2 * Δt for j= 1:1:numColPoints))
ux_cost = @expression(model, sum((ux[j] - 13)^2 * Δt for j= 1:1:numColPoints))
ax_cost = @expression(model, sum((ax[j])^2 * Δt for j=1:1:numColPoints)) # ax
    ↪cost

#TODO define cost weight
# w_y = [1e-5, 1e-4, 1e-3, 1e-2, 1e-1, 1,10];
# w_y = 10
w_y = 0.05 # change later for 2f
w_sr = 2.0
w_ax = 0.2
w_ux = 0.2
w_sa = 1.0

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# Objective: Minimize cost function
@Objective(model, Min, w_y * y_cost + w_sr * sr_cost + w_ax * ax_cost + w_ux * ux_cost + w_sa * sa_cost) # objective value
optimize!(model) # optimize model
StatesHis = value.(model[:xst]) # retrieve data
if abs(objective_value(model) - 3.65) < 0.1 # check answer
    println("Congrats, your answer is correct")
else
    println("Something went wrong, please try again!")
end

# println("Objective value model = ",objective_value(model))
println("Your y cost is: ", round(value(y_cost); digits = 3))

#Plot
# plot(StatesHis[:, 1], StatesHis[:, 2], tickfontsize = 10, xlabel = "X (m)", ylabel = "Y (m)",guidefont=15) # path plot
# plot(0:Δt:PredictionHorizon, StatesHis[:, 6], tickfontsize = 10, xlabel = "time (s)", ylabel = "ux (m/s)",guidefont=15) # Speed plot

display(plot(StatesHis[:, 1], StatesHis[:, 2], tickfontsize = 10, xlabel = "X (m)", ylabel = "Y (m)",guidefont=15)) # path plot
display(plot(0:Δt:PredictionHorizon, StatesHis[:, 6], tickfontsize = 10, xlabel = "time (s)", ylabel = "ux (m/s)",guidefont=15)) # Speed plot

```

The states derivative is: [9.9 1.496 -1.004 2.587 0.1 1.0 0.1]

This is Ipopt version 3.14.14, running with linear solver MUMPS 5.6.2.

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Number of nonzeros in equality constraint Jacobian...:    2473
Number of nonzeros in inequality constraint Jacobian.:         0
Number of nonzeros in Lagrangian Hessian...:    3602

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Total number of variables...:    722
      variables with only lower bounds:         0
      variables with lower and upper bounds:    722
      variables with only upper bounds:         0
Total number of equality constraints...:    560
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	1.1142900e+02	9.50e+00	2.64e-01	-1.0	0.00e+00	-	0.00e+00	0.00e+00	0
1	1.1109759e+02	8.23e+00	6.83e+00	-1.0	3.13e+01	-	1.67e-02	1.33e-01f	1
2	1.1025958e+02	6.92e+00	5.68e+00	-1.0	2.83e+01	-	6.67e-02	1.60e-01f	1
3	1.0819221e+02	5.04e+00	4.04e+00	-1.0	2.49e+01	-	9.68e-02	2.71e-01f	1

4	1.0469717e+02	3.07e+00	2.31e+00	-1.0	1.94e+01	-	1.38e-01	3.90e-01f	1
5	9.8916387e+01	1.03e+00	1.23e+00	-1.0	1.29e+01	-	2.15e-01	6.66e-01f	1
6	9.1320242e+01	2.90e-03	7.91e-01	-1.0	5.54e+00	-	4.76e-01	1.00e+00f	1
7	7.7137482e+01	1.85e-03	8.01e-01	-1.0	4.29e+00	-	5.18e-01	1.00e+00f	1
8	3.2102711e+01	2.00e-02	3.97e-01	-1.0	2.02e+01	-	4.57e-01	1.00e+00f	1
9	1.0894542e+01	1.70e-02	9.82e-02	-1.0	1.50e+01	-	6.62e-01	1.00e+00f	1
iter	objective	inf_pr	inf_du	lg(mu)	d	lg(rg)	alpha_du	alpha_pr	ls
10	5.5093386e+00	4.06e-03	6.48e-02	-1.0	7.91e+00	-	1.00e+00	1.00e+00f	1
11	4.2433091e+00	2.22e-03	1.18e-02	-1.7	3.76e+00	-	1.00e+00	1.00e+00f	1
12	3.8205562e+00	2.76e-03	8.76e-03	-2.5	1.96e+00	-	9.80e-01	1.00e+00f	1
13	3.6906527e+00	2.33e-03	3.35e-02	-3.8	5.20e-01	-	8.17e-01	1.00e+00h	1
14	3.6620614e+00	6.66e-04	5.61e-04	-3.8	1.74e-01	-	1.00e+00	1.00e+00h	1
15	3.6533282e+00	2.26e-04	2.77e-03	-5.7	7.89e-02	-	8.10e-01	9.78e-01h	1
16	3.6519431e+00	2.36e-05	4.94e-04	-5.7	2.39e-02	-	9.47e-01	1.00e+00h	1
17	3.6517338e+00	2.47e-06	5.25e-06	-5.7	8.66e-03	-	1.00e+00	1.00e+00h	1
18	3.6516747e+00	4.23e-07	1.36e-05	-8.6	2.04e-03	-	9.85e-01	9.76e-01h	1
19	3.6516712e+00	8.27e-08	3.93e-07	-8.6	1.09e-03	-	1.00e+00	1.00e+00h	1
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20	3.6516708e+00	1.94e-08	9.49e-08	-8.6	5.38e-04	-	1.00e+00	1.00e+00h	1
21	3.6516707e+00	4.52e-09	2.21e-08	-8.6	2.59e-04	-	1.00e+00	1.00e+00h	1
22	3.6516707e+00	8.61e-10	4.21e-09	-8.6	1.13e-04	-	1.00e+00	1.00e+00h	1
23	3.6516706e+00	1.12e-10	5.46e-10	-9.0	4.07e-05	-	1.00e+00	1.00e+00h	1

Number of Iterations...: 23

	(scaled)	(unscaled)
Objective...:	3.6516706220464390e+00	3.6516706220464390e+00
Dual infeasibility...:	5.4583440954618428e-10	5.4583440954618428e-10
Constraint violation...:	1.1227421770065860e-10	1.1227421770065860e-10
Variable bound violation:	9.1102048405122815e-09	9.1102048405122815e-09
Complementarity...:	2.1361287151070329e-09	2.1361287151070329e-09
Overall NLP error...:	2.1361287151070329e-09	2.1361287151070329e-09

Number of objective function evaluations	= 24
Number of objective gradient evaluations	= 24
Number of equality constraint evaluations	= 24
Number of inequality constraint evaluations	= 0
Number of equality constraint Jacobian evaluations	= 24
Number of inequality constraint Jacobian evaluations	= 0
Number of Lagrangian Hessian evaluations	= 23
Total seconds in IPOPT	= 0.051

EXIT: Optimal Solution Found.

Congrats, your answer is correct

Your y cost is: 33.942

This is Ipopt version 3.14.14, running with linear solver MUMPS 5.6.2.

Number of nonzeros in equality constraint Jacobian...: 2473  
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