

AER1517 Control for Robotics

Assignment 1: Model-Based Iterative Linear Quadratic Control

Prof. Angela Schoellig

- Download handout and script templates from Quercus
 - Problem 1.1 Mobile robot
 - Problem 1.2 Quadrotor
- Due on Feb. 17 (Monday) 23:59
- Submission through *Gradescope*
 - A single PDF with solutions and requested scripts
 - Both typed and scanned handwritten solutions are accepted
- Office hour: Feb. 13 (Thursday) 14:00 @ UTIAS

Problem 1.1 Mobile Robot | Overview

- Goal: Control a unicycle-type mobile robot to move along a straight line
- Two components:
 - LQR
 - ILQC

Problem 1.1 Mobile Robot | Model

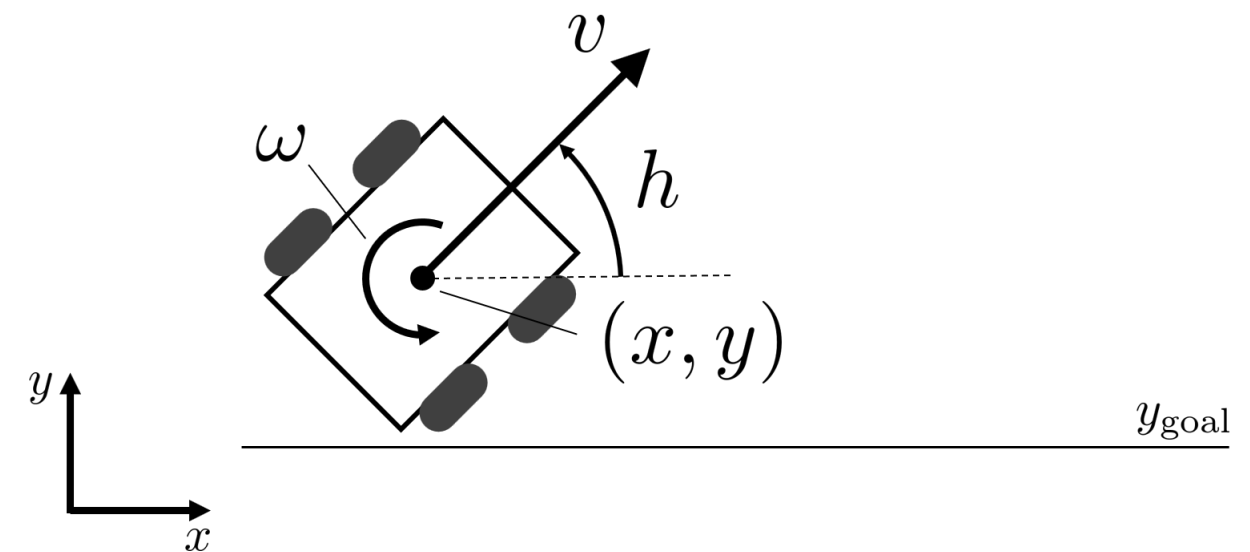
- State: $\mathbf{x} = [y, h]^T \in \mathbb{R}^2$
- Input: $\mathbf{u} = \omega \in \mathbb{R}^1$

- System dynamics:

$$\dot{y}(t) = v(t) \sin(h(t))$$

$$\dot{h}(t) = \omega(t)$$

(assume constant forward speed $v(t) = \bar{v}$)



Problem 1.1 Mobile Robot | Code Structure (LQR)

main_p1_lqr.m

- task_design()
- generate_model()
- LQR design
- mobile_robot_sim()
- plot_results()



```

EDITOR PUBLISH VIEW
main_p1_lqr.m x main_p2_ilqc.m x task_design.m x +
32 % define task
33 task_lqr = task_design();
34 N = length(task_lqr.start_time:task_lqr.dt:task_lqr.end_time);
35
36 % add model
37 const_vel = 1; % desired forward speed
38 model = generate_model(const_vel);
39
40 % initialize controller
41 controller_lqr = zeros(3, N-1);
42
43 % save directory
44 save_dir = './results/';
45
46 % flags
47 plot_on = true;
48 save_on = true;
49
50 %% [Problem 1.1 (c)] LQR Controller
51 % ===== [TODO] LQR Design =====
52 % Design a LQR controller based on the linearization of the system about
53 % an equilibrium point (x_eq, u_eq). The cost function of the problem is
54 % specified in 'task_lqr.cost' via the method 'task_design()'.
55 %
56 %
57 % =====
58
59 %% Simulation
60 sim_out_lqr = mobile_robot_sim(model, task_lqr, controller_lqr);
61 fprintf('--- LQR ---\n\n');
62 fprintf('trajectory cost: %.2f\n', sim_out_lqr.cost);
63 fprintf('target state [%3f; %3f]\n', task_lqr.goal_x);
64 fprintf('reached state [%3f; %3f]\n', sim_out_lqr.x(:,end));
65
66 %% Plots
67 if plot_on
68     plot_results(sim_out_lqr);
69 end

```

Problem 1.1 Mobile Robot | (a)-(c) LQR Design

- Assume infinite horizon
- Linearize dynamics about one point $(\mathbf{x}_{op}, \mathbf{u}_{op})$

$$\dot{\mathbf{x}}(t) = \mathbf{F}(\mathbf{x}(t), \mathbf{u}(t)) \quad \xrightarrow[\text{linearize}]{} \quad \delta\dot{\mathbf{x}}(t) = \mathbf{A}_{lin}\delta\mathbf{x}(t) + \mathbf{B}_{lin}\delta\mathbf{u}(t)$$

where $\delta\mathbf{x}(t) = \mathbf{x}(t) - \mathbf{x}_{op}$ and $\delta\mathbf{u}(t) = \mathbf{u}(t) - \mathbf{u}_{op}$

- Given the (Q, R) matrices defined in task_design(), design a LQR controller with lqr() function in Matlab
 - Note the convention of lqr() (see documentation)

Problem 1.1 Mobile Robot | (a)-(c) LQR Design

- Control law (static gains):

$$\mathbf{u}(t) = \mathbf{K}(\mathbf{x}(t) - \mathbf{x}_{\text{goal}}) = \underbrace{\begin{bmatrix} \boldsymbol{\theta}_{\text{ff}}^T & \boldsymbol{\theta}_{\text{fb}}^T \end{bmatrix}}_{\boldsymbol{\theta}^T} \begin{bmatrix} 1 \\ \mathbf{x}(t) \end{bmatrix} = \boldsymbol{\theta}_{\text{ff}}^T + \boldsymbol{\theta}_{\text{fb}}^T \mathbf{x}(t)$$

- ‘Repmat’ controller parameters for simulation: $\Theta = \underbrace{[\boldsymbol{\theta}, \boldsymbol{\theta}, \dots, \boldsymbol{\theta}]}_{(N-1) \text{ times}} \in \mathbb{R}^{3 \times (N-1)}$
- Complete `main_p1_lqr.m`
- Run and observe
 - Effect of changing Q and R matrices
 - Performance for different initial states

Problem 1.1 Mobile Robot | Code Structure (ILQC)

main_p1_ilqc.m

- task_design()
- generate_model()

- ILQC design

- mobile_robot_sim()

- plot_results()

```

EDITOR  PUBLISH  VIEW
main_p1_lqr.m  main_p2_ilqc.m  task_design.m  +
31      % add task
32      task_ilqc = task_design();
33      N = length(task_ilqc.start_time:task_ilqc.dt:task_ilqc.end_time);
34
35      % add model
36      const_vel = 1; % assume constant forward speed
37      model = generate_model(const_vel);
38
39      % save directory
40      save_dir = './results/';
41
42      % initialize controller
43      load(strcat(save_dir, 'lqr_controller'));
44      controller_ilqc = controller_lqr;
45
46      % flags
47      plot_on = true;
48      save_on = true;
49
50      %% [Problem 1.1 (j)] Iterative Linear Quadratic Controller
51      % ===== [TODO] ILQC Design =====
52      % Design a ILQC controller based on the linearized dynamics and quadratized
53      % costs. The cost function of the problem is specified in 'task_ilqc.cost'
54      % via the method 'task_design()'.
55      %
56      %
57      % =====
58
59      %% Simulation
60      sim_out_ilqc = mobile_robot_sim(model, task_ilqc, controller_ilqc);
61      fprintf('\n\ntarget state [%3f; %3f]\n', task_ilqc.goal_x);
62      fprintf('reached state [%3f; %3f]\n', sim_out_ilqc.x(:,end));
63
64      %% Plots
65      if plot_on
66          plot_results(sim_out_ilqc);
67      end

```


Problem 1.1 Mobile Robot | (d)-(j) ILQC Design

ILQC algorithm:

- Initial policy (LQR) - Θ_{lqr}
- Forward pass - $(\bar{\mathbf{x}}_0, \bar{\mathbf{u}}_0), (\bar{\mathbf{x}}_1, \bar{\mathbf{u}}_1), \dots, (\bar{\mathbf{x}}_{N-1}, \bar{\mathbf{u}}_{N-1}), \bar{\mathbf{x}}_N$
- (Initializations)
- Backward pass

at each time step, approximate system dynamics and cost function about $(\bar{\mathbf{x}}_k, \bar{\mathbf{u}}_k)$

- Linearizing system dynamics - $(\mathbf{A}_k, \mathbf{B}_k)$
- Quadratic approximation of costs - $(q_k, \mathbf{q}_k, \mathbf{Q}_k, \mathbf{r}_k, \mathbf{R}_k, \mathbf{P}_k)$
- Solve for optimal input at each time step and update policy - $\mathbf{u}_k = \bar{\mathbf{u}}_k + \delta \mathbf{u}_k^*$

repeat until convergence
criteria met

Problem 1.1 Mobile Robot | (d)-(j) ILQC Design

- Complete `main_p1_ilqc.m`
- Run and observe
 - Differences between LQR and ILQC
- Note: Model used in controller design and simulation are identical, but two functions `mobile_robot_sim.m` and `mobile_robot_sim_test.m` in `./simulation` can be used to explore model mismatch (see comments in code for details).

Problem 1.2 Quadrotor | Overview

- Goal: Control a quadrotor to reach a designated goal state and/or passing through a given via-point
- Four components (based on programming exercise from [1]):
 - LQR
 - LQR with via-point
 - ILQC
 - ILQC with via-point

Reference

[1] Jonas Buchli. *Optimal and Learning Control for Autonomous Robots*, ETH Zurich. Info: <http://www.adrlab.org/doku.php/adrl:education:lecture:fs2015>

Problem 1.2 Quadrotor | Model

- State: $\mathbf{x} = [x, y, z, \phi, \theta, \psi, \dot{x}, \dot{y}, \dot{z}, \dot{\phi}, \dot{\theta}, \dot{\psi}]^T \in \mathbb{R}^{12}$
- Input: $\mathbf{u} = [F_z, M_x, M_y, M_z]^T \in \mathbb{R}^4$
- System dynamics: $\dot{\mathbf{x}} = \mathbf{F}(\mathbf{x}, \mathbf{u}) = \mathbf{f}(\mathbf{x}) + \mathbf{g}(\mathbf{x})\mathbf{u}$
 - Nonlinear
 - Underactuated



AscTec Hummingbird [2]

References

[2] AscTec Hummingbird. Info: <http://www.asctec.de/en/uav-uas-drones-rpas-roav/asctec-hummingbird/#pane-0-0>

Problem 1.2 Quadrotor | Code Structure (LQR)

main_p2_lqr.m

- Task_Design()
- Model
- Cost_Design()
- LQR_Design()
- Quad_Simulator()
- Visualize2()

```
EDITOR PUBLISH VIEW
main_p2_lqr.m main_p2_ilqc.m Task_Design.m LQR_Design.m ILQC_Design.m Cost_Design.m +
32
33 %% General
34 % add subdirectories
35 addpath(genpath(pwd));
36
37 % define task
38 Task = Task_Design();
39
40 % load the dynamic model of the quadcopter
41 load('Quadrotor_Model.mat', 'Model'); % save as structure "Model"
42
43 % define cost function
44 Task.cost = Cost_Design( Model.param.mQ, Task );
45
46 % save directory
47 save_dir = './Results/';
48
49 % flags
50 plot_on = true;
51 save_on = true;
52
53 %% Initial LQR Controller Design
54 % [Problem 1.2 (a)-(c)] Fill in the missing parts in ...
55 % LQR_Design(Model, Task)
56 [Initial_Controller, Cost_LQR] = LQR_Design(Model, Task);
57
58 %% Simulation
59 Sim_Out_LQR = Quad_Simulator(Model, Task, Initial_Controller);
60 disp('LQR controller performance:');
61 fprintf('Cost with LQR controller (metric: LQR cost function!): J* = %.3f \n', Cost_LQR);
62 fprintf('Start Quadcopter position: x = %.3f, y = %.3f, z = %.3f \n', Sim_Out_LQR.x(1:3,1));
63 fprintf('Final Quadcopter position: x = %.3f, y = %.3f, z = %.3f \n\n', Sim_Out_LQR.x(1:3,end));
64
65 %% Visualization of LQR controller
66 if plot_on
67     Visualize2(Sim_Out_LQR, Model.param);
68 end
```

Problem 1.2 Quadrotor | (a) LQR Controller Gain

- LQR: Linearize system dynamics around one linearization point $(\mathbf{x}_{op}, \mathbf{u}_{op})$

$$\dot{\mathbf{x}}(t) = \mathbf{f}(\mathbf{x}(t)) + \mathbf{g}(\mathbf{x}(t))\mathbf{u}(t)$$



linearize

$$\delta\dot{\mathbf{x}}(t) = \mathbf{A}_{lin}\delta\mathbf{x}(t) + \mathbf{B}_{lin}\delta\mathbf{u}(t)$$

Useful functions

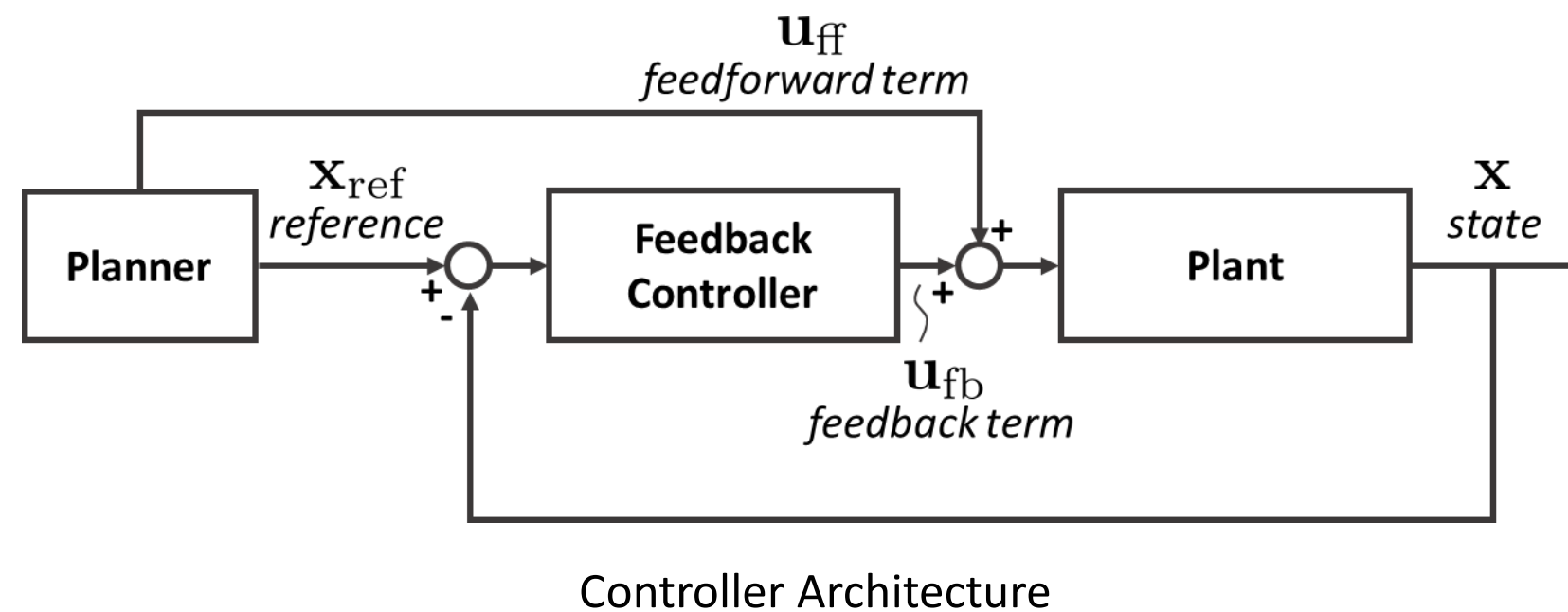
Model.Alin{1}(x_lin, u_lin, Model.param.syspar_vec)

Model.Blin{1}(x_lin, u_lin, Model.param.syspar_vec)

- Cost function (Q, R) matrices are specified in the struct Task.cost
 - Can be modified through Cost_Design()
- Matlab function lqr() for computing controller gain

Problem 1.2 Quadrotor | (b) LQR for Reaching Goal State

- Designing controller to reach goal state
- Controller has a feedforward component and a feedback component



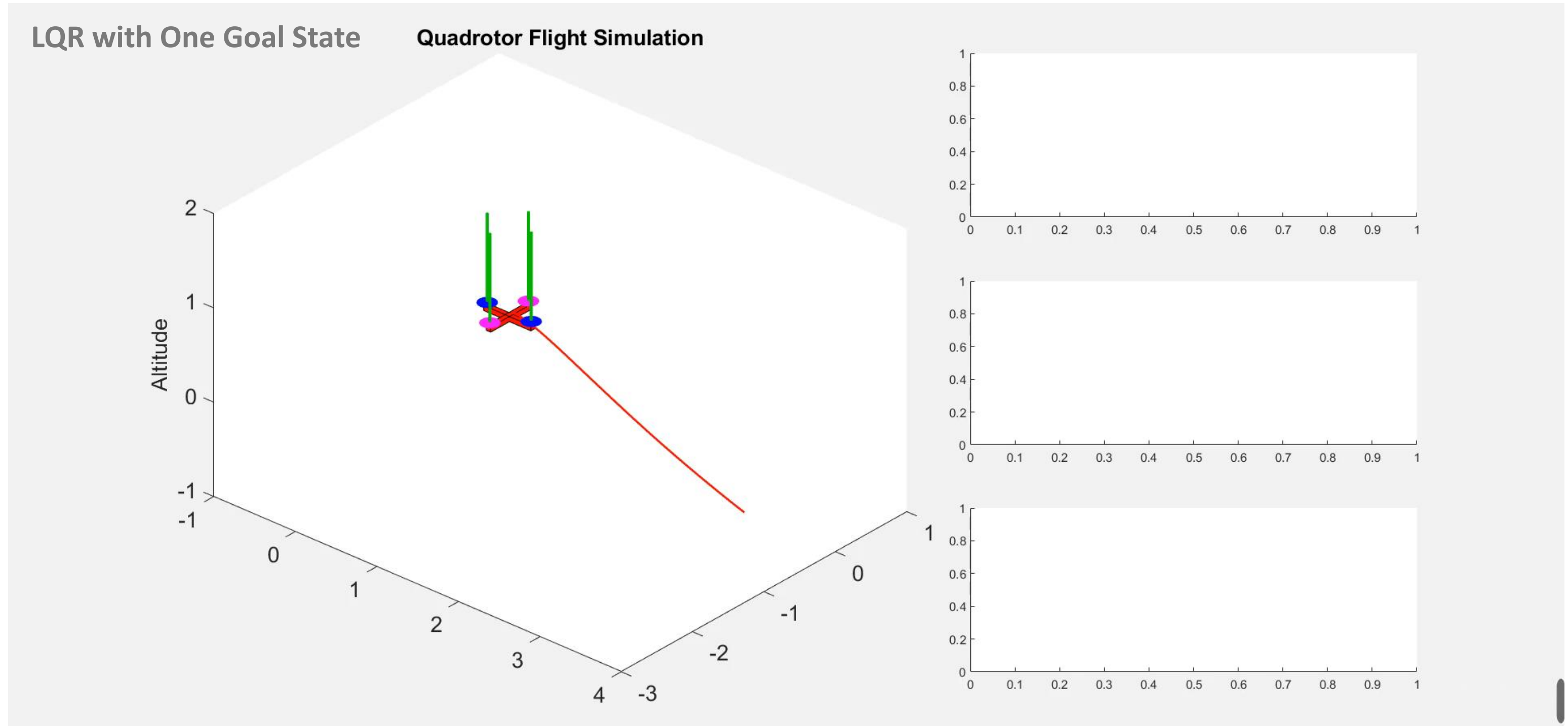
$$\begin{aligned}
 \mathbf{u}_k &= \mathbf{u}_{\text{ff}} + \mathbf{u}_{k,\text{fb}} \\
 &= \mathbf{u}_{\text{ff}} + \mathbf{K}(\mathbf{x}_k - \mathbf{x}_{\text{ref}}) \\
 &= \begin{bmatrix} \mathbf{u}_{\text{ff}} - \mathbf{K}\mathbf{x}_{\text{ref}} & \mathbf{K} \end{bmatrix} \begin{bmatrix} 1 \\ \mathbf{x}_k \end{bmatrix} \\
 &= \boldsymbol{\theta}_k^T \mathbf{x}_{k,\text{aug}}
 \end{aligned}$$

- Dimensions: $\boldsymbol{\theta}_k \in \mathbb{R}^{13 \times 4}$ (time step k), $\underbrace{\boldsymbol{\Theta} \in \mathbb{R}^{13 \times 4 \times (N-1)}}_{\text{LQR_Controller.theta}}$ (all time steps)

Problem 1.2 Quadrotor | (b) LQR for Reaching Goal State

- Complete corresponding parts in `LQR_Design()`
- Run and observe
 - Effect of changing (Q, R) matrices
 - Effect of changing goal locations
 - ...
- General note:
 - Private function `./Private/Design_functions/LQR_Design_Solution.p` is available for checking implementation

Problem 1.2 Quadrotor | (b) LQR for Reaching Goal State



Link to video: https://drive.google.com/open?id=1b2wJco_63T5li3juw85mi8wMTr6PB1JL

Problem 1.2 Quadrotor | (c) LQR with Via-Point

- Modify the controller such that the quadrotor passes through a prescribed via-point
 - Controller varies over time
- Complete `LQR_Design()`
- Run and observe
 - Performance as compared to LQR alone

Problem 1.2 Quadrotor | Code Structure (ILQC)

main_p2_ilqc.m

- Task_Design()
- Model
- Cost_Design()
- ILQC_Design()
- Quad_Simulator()
- Visualize2()

```

EDITOR PUBLISH VIEW
main_p2_lqr.m main_p2_ilqc.m Task_Design.m LQR_Design.m ILQC_Design.m Cost_Design.m +
36
37 % define task
38 Task = Task_Design();
39
40 % load the dynamic model of the quadcopter
41 load('Quadrotor_Model.mat', 'Model'); % save as structure "Model"
42
43 % define cost function
44 % [Problem 1.2 (e)] Fill in the missing parts under 'via_point' in ...
45 % Cost_Design( Model.param.mQ, Task )
46 Task.cost = Cost_Design( Model.param.mQ, Task );
47
48 % save directory
49 save_dir = './Results/';
50
51 % flags
52 plot_on = true;
53 save_on = true;
54 load_lqr = false;
55 ...
56 %% Problem 2: ILQC controller design
57 % [Problem 1.2 (d)] Fill in the missing parts in ...
58 % ILQC_Design(Model,Task,Initial_Controller,@Quad_Simulator)
59 [ILQC_Controller, Cost] = ILQC_Design(Model,Task,Initial_Controller,@Quad_Simulator);
60
61 %% Simulation
62 t_cpu = cputime;
63 Sim_Out_ILQC = Quad_Simulator(Model,Task,ILQC_Controller);
64 t_cpu = cputime - t_cpu;
65 fprintf('The ILQC algorithm found a solution in %fs \n\n',t_cpu);
66 fprintf('Final Quadcopter Position: xQ = %.3f, yQ = %.3f, zQ = %.3f \n', Sim_Out_ILQC.x(1:3,end));
67 fprintf('Final Quadcopter Velocity: xQ = %.3f, yQ = %.3f, zQ = %.3f \n', Sim_Out_ILQC.x(7:9,end));
68
69 %% Visualization of ILQC controller
70 if plot_on
71     Visualize2(Sim_Out_ILQC, Model.param);
72 end

```

LQR Controller

Problem 1.2 Quadrotor | (d) ILQC Controller Design

- ILQC: Linearize system dynamics around each discretized (state, input)
- Approximation to convert a continuous-time linearized system to a discrete-time linearized system:

$$\delta \dot{\mathbf{x}}(t) = \mathbf{A}_{\text{lin}} \delta \mathbf{x}(t) + \mathbf{B}_{\text{lin}} \delta \mathbf{u}(t)$$

↓ approx.

$$\frac{\delta \mathbf{x}_{k+1} - \delta \mathbf{x}_k}{\delta t} = \mathbf{A}_{\text{lin}} \delta \mathbf{x}_k + \mathbf{B}_{\text{lin}} \delta \mathbf{u}_k$$

$$\delta \mathbf{x}_{k+1} = \underbrace{(\mathbf{I} + \mathbf{A}_{\text{lin}} \delta t)}_{\mathbf{A}_k} \delta \mathbf{x}_k + \underbrace{(\mathbf{B}_{\text{lin}} \delta t)}_{\mathbf{B}_k} \delta \mathbf{u}_k$$

```

146 - for k = (length(sim_out.t)-1):-1:1
147
148     % state of system at time step n
149     x0 = X0(:,k);
150     u0 = U0(:,k);
151
152     % =====
153     % [Todo] Discretize and linearize continuous system dynamics Alin
154     % around specific pair (x0,u0). See exercise sheet Eqn. (18) for
155     % details.
156     %
157     % Alin = ...;
158     % Blin = ...;
159     % A = ...;
160     % B = ...;
161     % =====
162
163     % =====
164     % [Todo] quadratize cost function
165     % [Note] use function {q_fun, Qv_fun, Qm_fun, Rv_fun, Rm_fun,
166     % Pm_fun} provided above.
167     %
168     % t0 = T0(:,k);
169     % q = ...;
170     % Qv = ...;
171     % Qm = ...;
172     % Rv = ...;
173     % Rm = ...;
174     % Pm = ...;
175     % =====
    
```

Problem 1.2 Quadrotor | (d) ILQC Controller Design

- ILQC: Linearize system dynamics around each discretized (state, input)
- Approximation to convert a continuous-time linearized system to a discrete-time linearized system:

$$\delta \dot{\mathbf{x}}(t) = \mathbf{A}_{\text{lin}} \delta \mathbf{x}(t) + \mathbf{B}_{\text{lin}} \delta \mathbf{u}(t)$$

approx.

$$\delta \mathbf{x}_{k+1} = \underbrace{(\mathbf{I} + \mathbf{A}_{\text{lin}} \delta t)}_{\mathbf{A}_k} \delta \mathbf{x}_k + \underbrace{(\mathbf{B}_{\text{lin}} \delta t)}_{\mathbf{B}_k} \delta \mathbf{u}_k$$

```

146 - for k = (length(sim_out.t)-1):-1:1
147
148     % state of system at time step n
149     x0 = X0(:,k);
150     u0 = U0(:,k);
151
152     % =====
153     % [Todo] Discretize and linearize continuous system dynamics Alin
154     % around specific pair (x0,u0). See exercise sheet Eqn. (18) for
155     % details.
156     %
157     % Alin = ...;
158     % Blin = ...;
159     % A = ...;
160     % B = ...;
161     % =====
162
163     % =====
164     % [Todo] quadratize cost function
165     % [Note] use function {q_fun, Qv_fun, Qm_fun, Rv_fun, Rm_fun,
166     % Pm_fun} provided above.
167     %
168     % t0 = T0(:,k);
169     % q = ...;
170     % Qv = ...;
171     % Qm = ...;
172     % Rv = ...;
173     % Rm = ...;
174     % Pm = ...;
175     % =====

```

- Cost approximation: Jacobian and Hessian of the cost functions are provided in the script

Problem 1.2 Quadrotor | (d) ILQC Controller Design

- Similar to Problem 1.1, complete `ILQC_Design()`
 - Private function `./Private/Design_functions/ILQC_Design_Solution.p` is available for checking implementation
- Run and observe
 - Compare with LQR controller

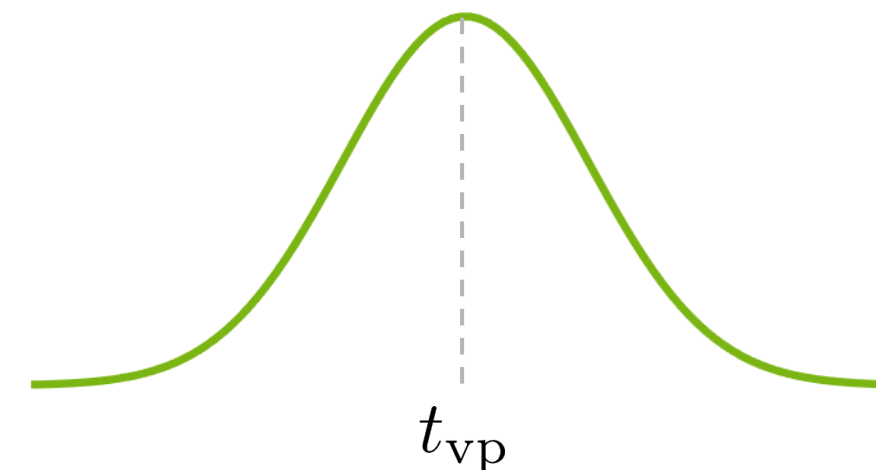
Problem 1.2 Quadrotor | (e) ILQC with Via-Point

- Add to ILQC cost function a term that penalizes deviation from via-point \mathbf{x}_{vp} in proximity of time t_{vp} :

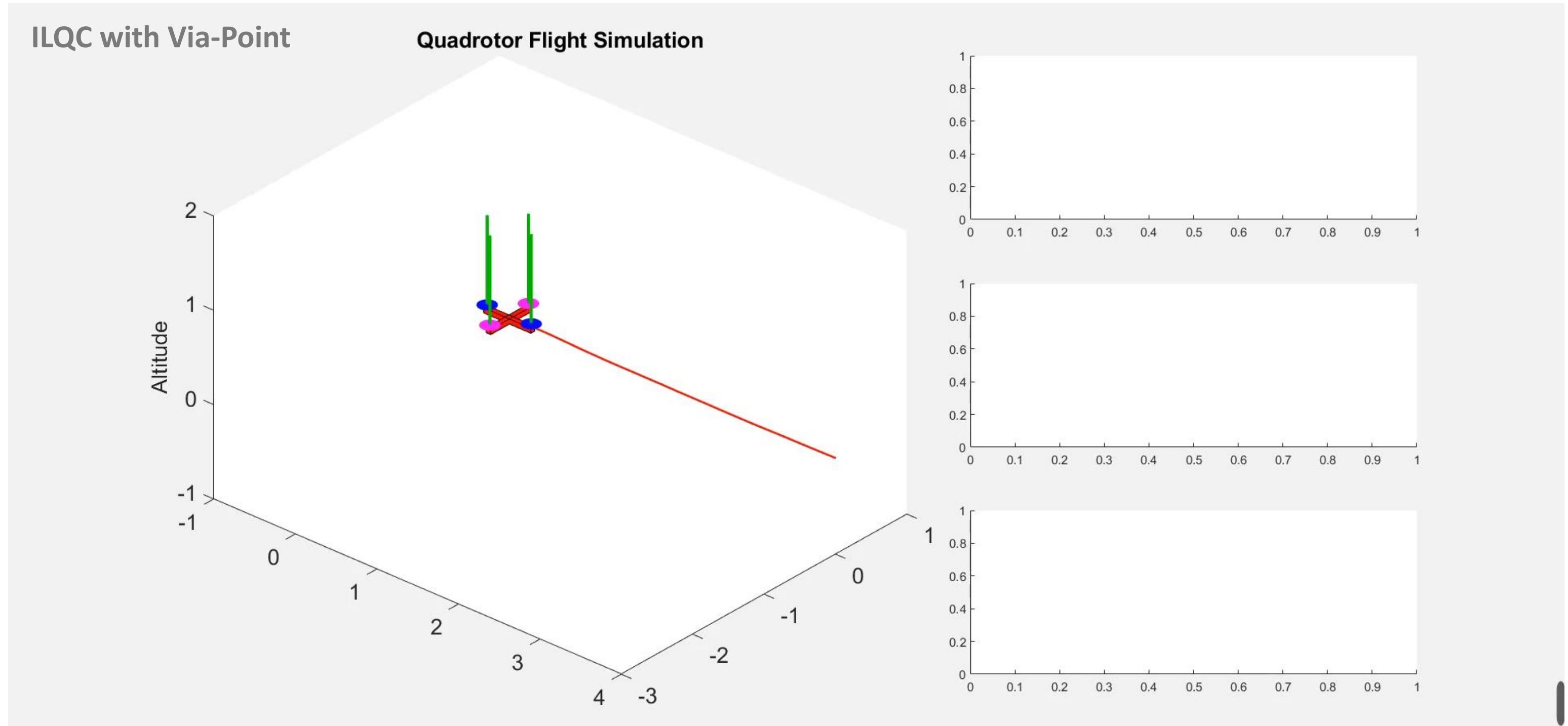
$$g_{vp}(t) = (\mathbf{x} - \mathbf{x}_{vp})^T \mathbf{Q}_{vp} (\mathbf{x} - \mathbf{x}_{vp}) \sqrt{\frac{\rho}{2\pi}} \exp\left(-\frac{\rho}{2}(t - t_{vp})^2\right)$$

weighted by ρ -steep bell curve

- Complete `Cost_Design()`
- Run and observe
 - Use different via-points
 - Compare with LQR



Problem 1.2 Quadrotor | (e) ILQC with Via-Point



Link to video: <https://drive.google.com/open?id=1awkzzF6MylcgP6hxMJwn7RcmvZobRFng>