```
1 # Fill in the respective functions to implement the
    LQR optimal controller
 2
 3 # Import libraries
 4 import numpy as np
 5 from base_controller import BaseController
 6 from scipy import signal, linalg
 7 from util import wrapToPi, closestNode
 8
 9 class CustomController(BaseController):
10
       def __init__(self, trajectory,
11
   look_ahead_distance=190):
12
           super().__init__(trajectory)
13
14
15
           # Define constants
16
           # These can be ignored in P1
17
           self.lr = 1.39
18
           self.lf = 1.55
19
           self.Ca = 20000
20
           self.Iz = 25854
21
           self.m = 1888.6
22
           self.g = 9.81
23
           # Add additional member variables according
24
    to your need here.
25
           self.look_ahead_distance =
   look_ahead_distance
26
           self.previous_psi = 0
27
           self.velocity_start = 58
           self.velocity_integral_error = 0
28
           self.velocity_previous_step_error = 0
29
30
       def update(self, timestep):
31
32
           trajectory = self.trajectory
33
34
           lr = self.lr
35
           lf = self.lf
36
           Ca = self.Ca
37
           Iz = self.Iz
```

```
38
           m = self.m
39
           g = self.g
40
41
           # Fetch the states from the BaseController
  method
42
           delT, X, Y, xdot, ydot, psi, psidot,
   obstacleX, obstacleY = super().getStates(timestep)
43
44
           # Set the look-ahead distance and find the
    closest index to the current position
45
           look_ahead_distance = 190
           _, closest_index = closestNode(X, Y,
46
   trajectory)
47
48
           # stop look-ahead distance from going out
  of bounds
49
           max_allowed_look_ahead = min(
   look_ahead_distance, len(trajectory) -
   closest_index - 1)
50
           look_ahead_distance = max(0,
   max_allowed_look_ahead)
51
52
          # Design your controllers in the spaces
   below.
53
           # Remember, your controllers will need to
  use the states
54
           # to calculate control inputs (F, delta).
55
           # -----|Lateral Controller
56
57
58
           # Please design your lateral controller
  below.
59
60
          # state space model for lateral control
61
           A = np.array(
               [[0, 1, 0, 0], [0, -4 * Ca / (m * xdot)]
62
   ), 4 * Ca / m, (-2 * Ca * (lf - lr)) / (m * xdot
   )], [0, 0, 0, 1],
63
                [0, (-2 * Ca * (lf - lr)) / (Iz * xdot
   ), (2 * Ca * (lf - lr)) / Iz,
```

```
64
                 (-2 * Ca * (lf ** 2 + lr ** 2)) / (
   Iz * xdot)]])
           B = np.array([[0], [2 * Ca / m], [0], [2
65
    * Ca * lf / Iz]])
           C = np.eye(4)
66
           D = np.zeros((4, 1))
67
68
69
           # discretize the state space model
70
           sys_continuous = signal.StateSpace(A, B, C
    D)
71
           sys_discretize = sys_continuous.
   to_discrete(delT)
72
           A_discretize = sys_discretize.A
73
           B_discretize = sys_discretize.B
74
75
           # calculate the desired heading angle (
   psi_desired) (referencing Project 2 solution)
           psi_desired = np.arctan2(trajectory[
76
   closest_index + look_ahead_distance, 1] -
   trajectory[closest_index, 1],
77
                                     trajectory[
   closest_index + look_ahead_distance, 0] -
   trajectory[closest_index, 0])
78
79
           # error calculation (referencing Project 2
    solution)
80
           e1 = (Y - trajectory[closest_index +
   look_ahead_distance, 1]) * np.cos(psi_desired) - (
                       X - trajectory[closest_index
81
    + look_ahead_distance, 0]) * np.sin(psi_desired)
           e2 = wrapToPi(psi - psi_desired)
82
           e1_dot = ydot + xdot * e2
83
84
           e2_dot = psidot
85
86
           # LQR controller design
87
           Q = np.eye(4)
88
           R = 40
89
90
           # solve for P and gain matrix K
91
           P = linalg.solve_discrete_are(A_discretize
     B_discretize, Q, R)
```

```
92
            K = linalg.inv(R + B_discretize.T @ P @
    B_discretize) @ (B_discretize.T @ P @ A_discretize
    )
 93
            # control delta calculation
 94
            delta = (-K @ np.array([[e1], [e1_dot], [
 95
    e2], [e2_dot]]))[0, 0]
 96
            delta = np.clip(delta, -np.pi / 6, np.pi
     / 6)
 97
            # -----|Longitudinal Controller
 98
 99
100
            # Please design your longitudinal
    controller below.
101
102
            # declaring PID variables
103
            Kp\_velocity = 95
104
            Ki_velocity = 1
105
            Kd_{velocity} = 0.005
106
107
            # velocity error calculation
            velocity = np.sqrt(xdot ** 2 + ydot ** 2
108
    ) * 3.6
109
            velocity_error = self.velocity_start -
    velocity
110
            self.velocity_integral_error +=
    velocity_error * delT
111
            velocity_derivative_error = (
    velocity_error - self.velocity_previous_step_error
    ) / delT
112
113
            # F with PID feedback control
114
            F = (velocity_error * Kp_velocity) + (self
    .velocity_integral_error * Ki_velocity) + (
                    velocity_derivative_error *
115
    Kd_velocity)
            # Return all states and calculated control
116
     inputs (F, delta) and obstacle position
            return X, Y, xdot, ydot, psi, psidot, F,
117
    delta, obstacleX, obstacleY
```