In [1]: import matplotlib.pyplot as plt import control as co import pandas as pd import numpy as np import sympy as sp import time from sympy.abc import a, b, q, s, z, omega, zeta from numpy.linalg import inv In [2]: u, uc, vs, y, ys, Gs, Hz, Hq = $sp.symbols('u(k),u_c(k)),v(s),y(k),y(s),G(s),H(z),H(q)')$ Ts = sp.symbols('T s') am1, am2, bm0, bm1 = $sp.symbols('a_{m_1},a_{m_2},b_{m_0},b_{m_1}')$ ao, a1, a2, b0, b1 = sp.symbols('a_o,a_1,a_2,b_0,b_1') r0, s0, s1, t0, t1 = $sp.symbols('r_0,s_0,s_1,t_0,t_1')$ Gs eq = sp.Eq(Gs, b/(s*(s + b)))The model is given by $G(s) = \frac{b}{s(b+s)}$. To discretize, $G(z) = (1 - z^{-1})Z\left\{\frac{b}{s^2(s+h)}\right\}$ $G(z) = (1 - z^{-1})Z\left\{\frac{1}{c^2} - \frac{1}{bs} + \frac{1}{b(b+s)}\right\}$ $G(z) = (1 - z^{-1}) \left(\frac{T_s z}{(z - 1)^2} - \frac{1}{b} \frac{z}{z - 1} + \frac{1}{b} \frac{z}{z - e^{-bT_s}} \right)$ In [3]: num, den = sp.fraction(Gz_eq_Ts_b) den_poly = sp.Poly(den, z) mono div = den_poly.coeffs()[0] sp.Eq(Hz, sp.collect(sp.simplify(Gz eq Ts b), z)) Out[3]: $H(z) = \frac{-T_s b + z \left(T_s b e^{T_s b} - e^{T_s b} + 1 \right) + e^{T_s b} - 1}{b \left(z^2 e^{T_s b} + z \left(-e^{T_s b} - 1 \right) + 1 \right)}$ **Pulse Function** In [4]: radius = 38/2000 # 38 mm diameter to radius in m mass = 4/7000 # grams per ballarea_ball = np.pi * radius**2 volume_ball = 4/3*np.pi*radius**3 density = 1.2 $\# kg/m^3$ veq = 2.8 # m/sTs val = 0.2b_nom = 2*9.81*(mass-density*volume_ball)/(mass*veq) In [5]: ## Example 3.1: since it has the same structure TF I used it to validate my approach, it's consistent # sub_vals = [(Ts, Ts_val), (b, 1)] ## This is for our project sub vals = [(Ts, Ts val), (b, b nom)]num mono z = sp.collect(num/mono div, z).subs(sub vals)den mono z = sp.collect(den/mono div, z).subs(sub vals)Hq eq = sp.Eq(Hq, num mono z/den mono z)Hq eq # np.roots(sp.Poly(num_mono_z).coeffs()) Out[5]: $H(q) = \frac{0.0888233086497573z + 0.0575823297146081}{1.0z^2 - 1.26797180817817z + 0.267971808178173}$ In [6]: n_coeffs_nom, d_coeffs_nom = sp.fraction(Hq_eq.rhs) pulse_coeffs = sp.Poly(d_coeffs_nom).coeffs()[1:] for bi in sp.Poly(n_coeffs_nom).coeffs(): pulse_coeffs.append(bi) zeros = np.roots(sp.Poly(n_coeffs_nom).coeffs()) poles = np.roots(sp.Poly(d_coeffs_nom).coeffs()) zeros[0] -0.6482794954381095 Out[6]: This is a sanity check with the nominal values to check the stability of the zeros. Here, the zero is unstable for a nominal value of b =-0.6482794954381095 and the poles are 1.0 and 0.26797180817817307. **Control Parameter Derivation** Given that the zeros are unstable, the parameters are derived without zero cancelation. Here we know from the compatability conditions that, $degA_m = degA = 2$ $degB_m = degB = 1$ Since the zeros in B are stable then $B^+ = q + \frac{b_1}{b_0}$ and $B^- = b_0$ Then, $degA_0 = degA - degB^+ - 1 = 2 - 1 - 1 = 0$ Using the Diophantine equation we get, $AR' + b_0 S = A_0 A_m$ Let $A_o = 1$ Since the process is second order, then, degR = degS = degT = 1, with R being monic. **Control Parameters** In [7]: # Process Values A = q**2 + a1*q + a2B = b0*q + b1 $_{\rm Bplus} = q + b1/b0$ Bminus = b0 # Model Values Am = q**2 + am1*q + am2 $\underline{Bm} = \underline{Am.subs(q, 1)*q}$ # Control Values R = q + r0 $_{S} = s0*q + s1$ T = t0*qDiophantine equation: In [8]: diophantine = $sp.Eq((_A*1 + b0*_S), (_Ao*_Am))$ dio LHS coeffs = sp.Poly(diophantine.lhs, q).coeffs() dio_RHS_coeffs = sp.Poly(diophantine.rhs, q).coeffs() In [9]: # for i in range(len(dio LHS coeffs)): # display([dio LHS coeffs[i], dio RHS coeffs[i]]) Finding s_0 In [10]: _s0 = sp.solve(dio_LHS_coeffs[1] - dio_RHS_coeffs[1], s0)[0] Out[10]: Finding s_1 In [11]: _s1 = sp.solve(dio_LHS_coeffs[2] - dio_RHS_coeffs[2], s1)[0] sp.Eq(s1, s1) $s_1 = \frac{-a_2 + a_0 a_{m_2}}{b_0}$ Out[11]: Finding T In [12]: $_{t0} = _{Ao*_Bm/b0}$ Out[12]: $\frac{a_o q \left(a_{m_1} + a_{m_2} + 1\right)}{b_0}$ In [13]: $R = _Bplus$ $S = _s0*q + _s1$ $T = _Ao*_Bm/_Bminus$ display(R) display(S) display(T) $\frac{q\left(-a_1 + a_0 a_{m_1}\right)}{b_0} + \frac{-a_2 + a_0 a_{m_2}}{b_0}$ $\underline{a_o q \bigg(a_{m_1} + a_{m_2} + 1 \bigg)}$ **Control Action** In [14]: uk1, uck1, yk1 = sp.symbols('u(k-1), u c{(k-1)}, y(k-1)') control_action = $(sp.expand(sp.Eq((_R*u)/q, (_T*uc - _S*y)/q)))$ control subs = [(1/q*u, uk1), (1/q*uc, uck1), (1/q*y, yk1)]control action sol = sp.Eq(u, sp.solve(control action.subs(control subs), u)[0]) control_action_sol.subs([(r0, b1/b0), $(s0, _s0),$ (s1, _s1), $(t0, _t0),$ (ao, 1)]) Out[14]: $u(k) = -\frac{b_1 u(k-1)}{b_0} + \frac{q u_c(k) \left(a_{m_1} + a_{m_2} + 1\right)}{b_0} - \frac{y(k) \left(-a_1 + a_{m_1}\right)}{b_0} - \frac{y(k-1) \left(-a_2 + a_{m_2}\right)}{b_0}$ **Simulation Process Deivation** $y(k) = -a_1 y(k-1) - a_2 y(k-2) + b_0 u(k-1) + b_1 u(k-2) = \phi(t-1)^T \theta$ where, $\phi(t-1) = \left[-y(k-1) - y(k-2) u(k-1) u(k-2) \right]^T$ and $\theta = \begin{bmatrix} a_1 & a_2 & b_0 & b_1 \end{bmatrix}^T$ In [15]: final_time = 90 t = np.arange(0, final time + Ts val, Ts val) def reference_signal(end_time=final_time, Ts_func=Ts_val, lower_set=0.2, upper_set=0.1, period=30): uc func = [] time = np.arange(0, end_time + Ts_func, Ts_func) for _t in time: rat = 2*np.pi/period **if** np.sin(rat * _t) >= 0: uc_func.append(upper_set) uc_func.append(lower_set) return np.array(uc_func, float) uc_val = reference_signal() # plt.plot(np.arange(0, 60 + Ts_val, Ts_val), uc) # plt.show() In [16]: $omega_n = 1$ zeta = 1 std = 0# std = 0.001Bmz tf, Amz tf = co.tfdata(co.sample system(co.tf([1], [1, 2*zeta*omega n, omega n**2]), method='zoh', Ts=Ts va AM1 = Amz tf[0][0][1] $AM2 = Amz_tf[0][0][2]$ # AM1 = $2*zeta*omega_n$ $\# AM2 = omega_n**2$ AM SUM = 1 + AM1 + AM2A0 = np.array([1])lam = 0.98initial_P_weights = [10000]*4# initial P weights = [100, 100, 10, 10] theta = np.array(pulse_coeffs, float).reshape(4, -1) In [17]: # Estimates k = 0time_ns = time.time_ns() theta_hat = np.array([-0.5, 0.5, 0.25, 0.25], float).reshape(4, -1) # 0: a1 1: a2 2: b0 3: b1 # theta_hat = np.array([-0.5, 0.5, 0.5, 0.5], float).reshape(4, -1) $# theta_hat = np.ones((4,1), float)$ theta_arr = theta_hat P = np.diag(initial P weights) phi = np.zeros((4,1))y_measure = (phi.T@theta + np.random.normal(0, std)).reshape(-1,) r0_est = theta_hat[3] $s0_{est} = A0*AM1 - theta_hat[0]$ $s1_{est} = A0*AM2 - theta_hat[1]$ t0_est = A0*AM_SUM $M = np.array([r0_est, s0_est, s1_est, t0_est], float).reshape(-1, 1)$ $N = np.array([0, -y_measure[0], 0, uc_val[0]], float).reshape(M.shape)$ u val = 1/theta[2]*(N.T@M).reshape(-1,) # Estimates k = 1 $phi = np.array([-y_measure[0], 0, u_val[0], 0], float).reshape(-1,1) \# phi \ of \ 0$ K = P@phi@inv(lam + phi.T@P@phi) y_measure = np.concatenate((y_measure, (phi.T@theta + np.random.normal(0, std)).reshape(-1,))) theta_hat = theta_hat + K@(y_measure[-1] - phi.T@theta_hat) theta arr = np.concatenate((theta arr, theta_hat.reshape(-1, 1)), axis=1) P = (np.eye(len(phi)) - K@phi.T) @P/lam r0_est = theta_hat[3] $s0_{est} = A0*AM1 - theta_hat[0]$ $s1_{est} = A0*AM2 - theta_hat[1]$ t0_est = A0*AM_SUM $M = np.array([r0_est, s0_est, s1_est, t0_est], float).reshape(-1, 1)$ $\label{eq:nparray} \texttt{N = np.array([-u_val[0], -y_measure[1], -y_measure[0], uc_val[1]], float).reshape(\texttt{M.shape})}$ u val = np.concatenate((u_val, 1/theta[2] * (N.T@M).reshape(-1,))) for k in range(2, len(t)): $phi = np.array([-y_measure[k-1], -y_measure[k-2], u_val[k-1], u_val[k-2]], float).reshape(-1,1)$ K = P@phi@inv(lam + phi.T@P@phi) y_measure = np.concatenate((y_measure, (phi.T@theta + np.random.normal(0, std)).reshape(-1,))) theta_hat = theta_hat + K@(y_measure[-1] - phi.T@theta_hat) theta_arr = np.concatenate((theta_arr, theta_hat.reshape(-1, 1)), axis=1) P = (np.eye(len(phi)) - K@phi.T)@P/lam r0_est = theta_hat[3] $s0_est = A0*AM1 - theta_hat[0]$ $s1_{est} = A0*AM2 - theta_hat[1]$ t0_est = A0*AM_SUM $M = np.array([r0_est, s0_est, s1_est, t0_est], float).reshape(-1, 1)$ $N = np.array([-u_val[k-1], -y_measure[k], -y_measure[k-1], uc_val[k]]).reshape(M.shape)$ u val = np.concatenate((u val, 1/theta[2] * (N.T@M).reshape(-1,))) $(time.time_ns() - time_ns)*1e-9/len(t)*1e6$ 67.62372505543237 Out[17]: In [18]: filepath = 'C:/Users/dboas065/University of Ottawa/ELG7113/system_models/derek_model/with_pictures/' colors = ['blue', 'orange', 'red', 'green'] for row in range(len(theta_arr)): plt.plot(t, theta_arr[row,:]) plt.hlines(theta[row][0], 0, t[len(t) - 1], colors=colors[row], linestyles='--') plt.title('Parameter Adaptation vs. Time, t') plt.xlabel('Time, t (sec)') plt.ylabel('Parameter Values') plt.legend(['\$a_1\$', '\$a_2\$', '\$b_0\$','\$b_1\$']) plt.tight_layout() # plt.savefig(filepath+'adaptation.pdf', format='pdf') plt.savefig(filepath+'adaptation noise.pdf', format='pdf') plt.show() st = 0 et = len(t)plt.plot(t[st:et], y_measure[st:et]) plt.plot(t[st:et], uc_val[st:et]) plt.title('System Response vs. Time, t') plt.xlabel('Time, t (sec)') plt.ylabel('Height from Bottom, h (m)') plt.legend(['\$y(t)\$', '\$u_c(t)\$']) plt.tight_layout() # plt.savefig(filepath+'output.pdf', format='pdf') plt.savefig(filepath+'output_noise.pdf', format='pdf') plt.show() plt.step(t[st:et], u_val[st:et]) plt.title('Input, $v_f(t) - v_{eq} \ vs. Time, t')$ plt.xlabel('Time, t (sec)') plt.ylabel('Air Speed, $v_f(t) - v_{eq}$ (\$m/s^2\$)') plt.legend(['\$u(t)\$']) plt.tight_layout() # plt.savefig(filepath+'input.pdf', format='pdf') plt.savefig(filepath+'input_noise.pdf', format='pdf') Parameter Adaptation vs. Time, t 0.50 0.25 0.00 Parameter Values -0.25-0.50-0.75-1.00-1.250 20 40 Time, t (sec) System Response vs. Time, t y(t)0.30 u_c(t) Height from Bottom, h (m) 0.25 0.20 0.15 0.10 0.05 0.00 0 40 Time, t (sec) Input, $v_f(t) - v_{eq}$ vs. Time, t u(t) 0.6 Air Speed, $v_r(t) - v_{eq} (m/s^2)$ 0.2 0.0 80 20 40 60 Time, t (sec) In []: In []: