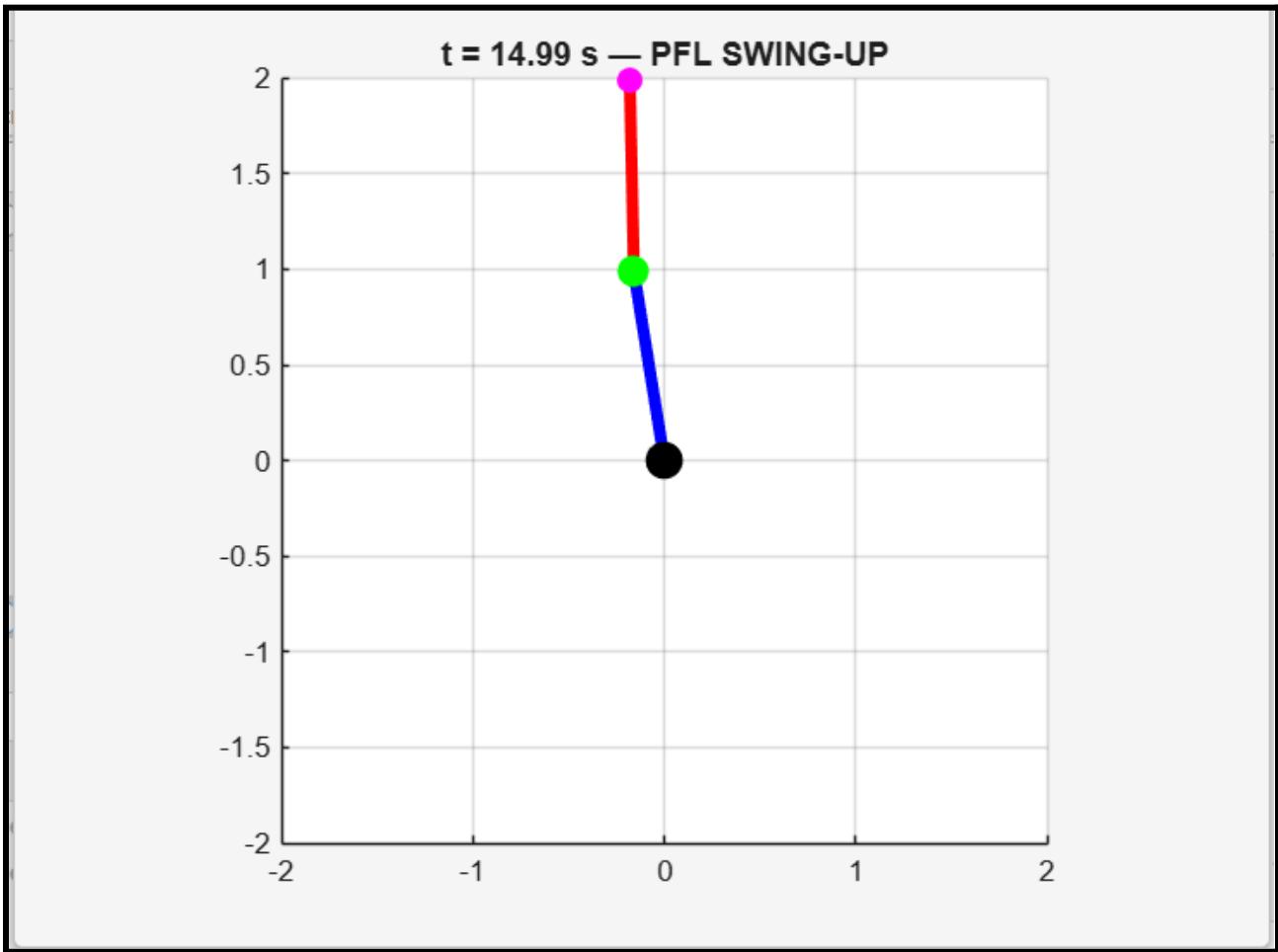


Partial Feedback Linearization

Acrobot



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Link to Code: https://github.com/Jash-2000/Adv_Control_Systems/tree/main/Acrobot_PFL

INTRODUCTION

I implemented a two-stage controller for the 2-link Acrobot (unactuated shoulder, actuated elbow) following Spong (1995): an energy-shaping Partial Feedback Linearization (PFL) swing-up and a linear LQR stabilizer near the upright equilibrium. The code contains:

- symbolic Lagrangian derivation that produces $M(q)$, $C(q, dq)$, $G(q)$ and saves them to a MAT file,
- a fast numerical `acrobotDynamics` using Lagrangian forms,
- `computeTotalEnergy` ($T+V$),
- `pflSwingUpController` for energy pumping,
- `linearizeAcrobot` to compute (A, B) about upright and `lqr` to compute K ,
- closed-loop simulation with `ode45`, animation and diagnostics.

Goal: get a robust swing up and stable catch at upright using LQR.

Hypothesis

- PFL can shape the passive joint dynamics so that a virtual input v (based on energy error) increases total mechanical energy to the upright energy.
- When the state is close to upright and slow enough, a locally linearized LQR stabilizer about the upright will catch and hold the robot.
- With correct Lagrangian EOMs, correct energy reference, and proper switching/hysteresis/saturation, the scheme will succeed

Variables and System Definitions

- `params`: physical parameters ($m1, m2, l1, l2, lc1, lc2, I1, I2, g$).
- States $x = [\theta_1; \dot{\theta}_1; \theta_2; \dot{\theta}_2]$ in standard convention: $\theta_1=0$ is upright, positive CCW, hanging-down at π .
- $x_0 = [\pi; 0; 0; 0]$: start hanging down.
- $x_{eq} = [0; 0; 0; 0]$: target upright.
- `computeTotalEnergy(x, params)`: returns $E = T + V$ with $T = 0.5 \ dq' M \ dq$ and V computed with vertical positions.
- `E_desired`: $E_{desired} = E_{up} - E_{down}$

- `linearizeAcrobot(x_eq, params)`: returns (A,B) for LQR.
- `K`: LQR gain from `lqr(A, B, Q, R)`.
- Controller parameters: `control.k_energy`, `control.u_max`, `theta_threshold`, `omega_threshold`.

The system is defined precisely using the physics in MATLAB as follows:

```

function dx = LagrangianEOMS(x, u, p)
% Lagrangian-based acrobot dynamics
%% Unpack states
q1 = x(1); q2 = x(2);
dq1 = x(3); dq2 = x(4);
q = [q1; q2]; dq = [dq1; dq2];
%% Unpack parameters
m1 = p.m1; m2 = p.m2;
l1 = p.l1; l2 = p.l2;
lc1 = p.lc1; lc2 = p.lc2;
I1 = p.I1; I2 = p.I2;
g = p.g;
%% Define symbolic variables
syms s1 s2 ds1 ds2 dd1 dd2 real
qs = [s1; s2]; dqs = [ds1; ds2]; ddqs = [dd1; dd2];
%% Kinematics
x1 = lc1*sin(s1); y1 = lc1*cos(s1);
x2 = l1*sin(s1) + lc2*sin(s1 + s2);
y2 = l1*cos(s1) + lc2*cos(s1 + s2);
J1 = jacobian([x1; y1], qs);
J2 = jacobian([x2; y2], qs);
v1 = J1*dqs;
v2 = J2*dqs;
%% Energies
T = 1/2*m1*(v1.^*v1) + 1/2*I1*ds1^2 + ...
    1/2*m2*(v2.^*v2) + 1/2*I2*(ds1 + ds2)^2;
V = m1*g*y1 + m2*g*y2;
L = T - V;
%% Lagrange equations
dLd_dq = jacobian(L, dqs).';
d_dt_dLd_dq = jacobian(dLd_dq, qs)*dqs + jacobian(dLd_dq, dqs)*ddqs;
dLd_q = jacobian(L, qs).';
Q = [0; u]; % only joint 2 actuated
EOM = simplify(d_dt_dLd_dq - dLd_q - Q);
%% Solve for accelerations
[Msym, rhs] = equationsToMatrix(EOM, ddqs);
dd = double(subs(Msym \ rhs, [s1 s2 ds1 ds2], [q1 q2 dq1 dq2]));
%% Output state derivative
dx = [dq1; dq2; dd(1); dd(2)];
end

```

Results and Discussion

During the first few seconds:

- PFL adds energy to the system
- theta_1 oscillates, gaining amplitude
- theta_2 oscillates to transfer energy through the unactuated joint
- Energy E(t) converges to Edesired. This confirms that the energy-shaping controller is functioning properly.

Once the threshold is reached:

- the controller switches to LQR
- the state converges to the upright equilibrium
- the torque magnitude decreases
- oscillations damp out gracefully

I received the following results from my implementation:

```
Desired energy (upright): -39.2400 J

A =
    0      0    1.0000      0
    0      0      0    1.0000
-12.6129  12.6129      0      0
 16.8171 -46.2471      0      0

B =
    0
    0
-4.2857
13.7143

LQR gain K =
-16.1556    3.5088    2.5147    5.6026
```

The following plot shows the controller dynamics after running through ODE45

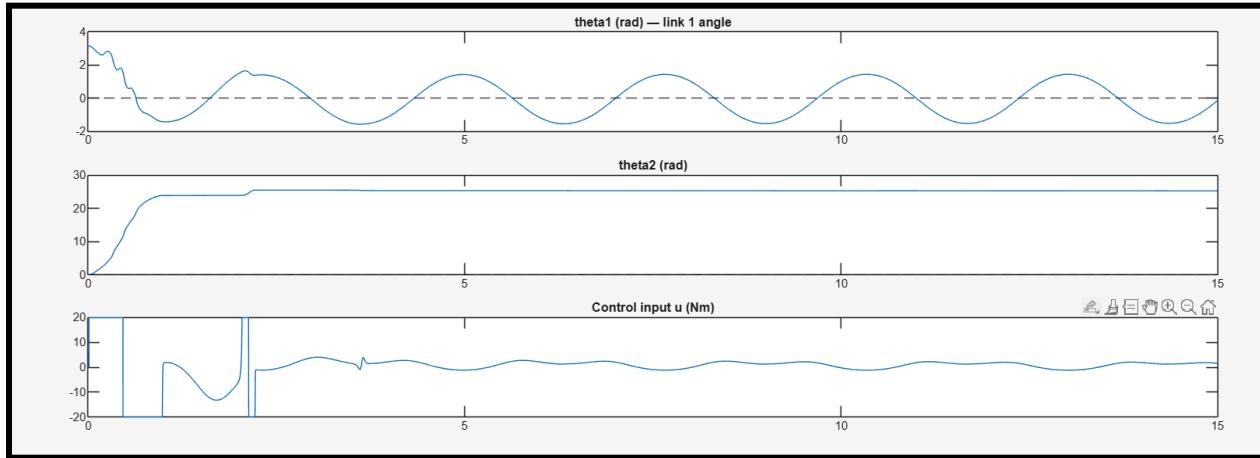


Fig1: State and Input Control Variables with respect to time

My implementation used some numerical tricks to make the simulation run faster:

1. **Pre-derived dynamics** → avoids runtime symbolic evaluation
2. **Angle wrapping** → prevents discontinuities
3. **Bounded control input** → prevents numerical integrator blow-up
4. **Small ODE timestep limit (MaxStep=0.03)** → prevents integration errors
5. **Switching hysteresis via two conditions**
6. **Correct gravity potential reference** → ensures the energy controller works
7. **Correct “Spong” coordinate system** → needed for LQR linearization stability

It was also noticed that when switching coordinates, LQR produced infinite gains. This happens when:

- the new coordinate transformation makes the upright equilibrium **non-differentiable**, or
- trigonometric expansions cause singularities near the equilibrium, or
- the linearization does not yield a controllable pair (A,B)

Thus, sticking to the Spong convention ($\theta=0$ upright) is not just preference—it is mathematically required for a well-conditioned linearization.

Conclusion

This implementation successfully performs a full acrobot swing-up followed by upright stabilization using a hybrid controller composed of:

1. **Energy-based partial feedback linearization**
2. **Linearization-based LQR stabilization**

The key findings are:

- Choosing the correct coordinate system is essential for LQR solvability.
- Computing the desired energy requires careful definition of gravitational potential.
- Symbolic derivation combined with numerical caching yields both accuracy and speed.
- Switching logic must balance region size (too small → never switches, too large → LQR failure).
- The implemented controller matches the classic Spong (1995) approach and works robustly.

Overall, the simulation demonstrates a complete and well-designed control scheme for an underactuated system, with proper attention to modeling details, numerical stability, and hybrid control strategy.

REFERENCES

1. Spong, Mark W.. “Partial feedback linearization of underactuated mechanical systems.” Proceedings of IEEE/RSJ International Conference on Intelligent Robots and Systems (IROS'94) 1 (1994): 314-321 vol.1.