Optimal Control and the Ricatti Equation

Calculus of Variations with Dynamic Systems

Recall, a function which minimizes the functional

$$J(x) = \int_{a}^{b} F(t, x, \dot{x}) dt$$

must also satisfy the Euler Legrange equation

$$F_x - \frac{d}{dt}(F_{\dot{x}}) = 0$$

Example: Find x(t) which minimizes the functional

$$J = \int_0^1 (x^2 + \dot{x}^2) dt$$

subject to the constraings that x(0) = 1, x(1) = 0

Solution: The Euler Legrange equation gives

$$F = x^{2} + \dot{x}^{2}$$
$$F_{x} - \frac{d}{dt}(F_{\dot{x}}) = 0$$
$$2x - \frac{d}{dt}(2\dot{x}) = 0$$
$$x - \ddot{x} = 0$$

Using LaPlace notation

$$(1-s^2)X=0$$

Either x = 0 (the trivial solution) or $s = \{+1, -1\}$. The general solution is then

$$X(t) = ae^t + be^{-t}$$

Plugging in the boundary conditions gives

$$x(0) = 1 = a + b$$

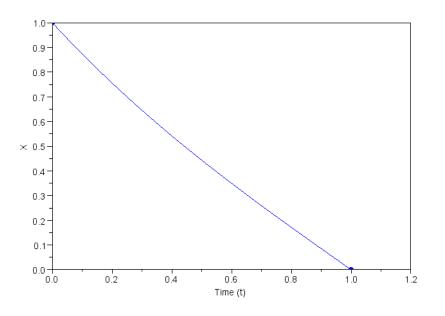
 $x(1) = 0 = 2.7183a + 0.3679b$

or

so the funciton which minimizes this funcitonal is

$$X(t) = -0.1565e^{t} + 1.1565e^{-t}$$

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-->t = [0:0.001:1]';
-->x = -0.1565*exp(t) + 1.1565*exp(-t);
-->plot(t,x);
-->xlabel('Time (t)');
-->ylabel('X');
```



Optimal path of x(t) with the cost function $J = \int_0^1 (x^2 + \dot{x}^2) dt$

Euler Legrange Equation with Two Dependent Variables

If you have *two* dependent variables:

$$J = \int_{a}^{b} F(t, x, \dot{x}, u, \dot{u}) dt$$

you have Itwo Euler Legrange equations to solve

$$F_x - \frac{d}{dt}(F_{\dot{x}}) = 0$$
$$F_u - \frac{d}{dt}(F_{\dot{u}}) = 0$$

Euler Legrange Equation with Contraints:

Finally, if you have constraints, such as

$$G(t, x, \dot{x}, u, \dot{u}) = 0$$

you can modify the const functional by adding a Legrange multiplier:

$$J = \int_{a}^{b} (F(t, x, \dot{x}, u, \dot{u}) + MG(t, x, \dot{x}, u, \dot{u})) dt$$

You can then solve this functional by plugging in the boundary conditions and the constraint on G(t,x,x').

Example 2: Find x(t) to minimize

$$J = \int_0^1 (x^2 + u^2) dt$$

subject to the constraints

$$\dot{X} = U$$

x(0) = 1
x(1) = 0

Solution: Add a Legrange multiplier so that F becomes

$$F = x^2 + u^2 + m(\dot{x} - u)$$

You now have three sets of Euler LaGrange equations to solve:

i) With respect to x:

$$F_x - \frac{d}{dt}(F_{\dot{x}}) = 0$$
$$2x - \frac{d}{dt}(m) = 2x - \dot{m} = 0$$

ii) With respect to u:

$$F_u - \frac{d}{dt}(F_u) = 0$$
$$2u - m = 0$$

iii) With respect to m:

$$F_m - \frac{d}{dt}(F_{\dot{m}}) = 0$$
$$\dot{x} - u = 0$$

Solving: From ii)

From iii)

$$u = \dot{x}$$

 $\dot{u} = \ddot{x}$

Substitute into i)

$$2x = \dot{m} = 2\dot{u} = 2\ddot{x}$$

or

$$\ddot{X} = X$$

or in LaPlace notation

$$\ddot{x} - x = 0$$
$$(s^2 - 1)x = 0$$

This has solutions of

• x = 0 (trivial solution), or

so

$$X(t) = ae^t + be^{-t}$$

Plugging in the constraints

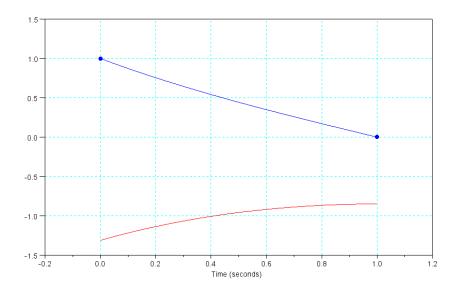
$$x(0) = 1 = a + b$$

 $x(1) = 0 = 2.7183a + 0.3679b$

results in

and

$$x(t) = -0.1565e^{t} + 1.1565e^{-t}$$
$$u(t) = \dot{x}(t) = -0.1565e^{t} - 1.1565e^{-t}$$



Optimal path for x(t) (blue) and input u(t) (red) for cost function $J = \int_0^1 \left(x^2 + u^2 \right) dt$

Note: If you change the funcitonal to weight x more heavily, it is driven to zero quicker:

$$J = \int_0^1 (100x^2 + u^2) dt$$

The functional becomes:

$$F = 100x^2 + u^2 + m(\dot{x} - u)$$

which results in the following three Euler LaGrange equations:

$200x - \dot{m} = 0$	(partials with respect to x)
2u - m = 0	(partials with respect to u)
$\dot{x} - u = 0$	(partials with respect to m)

which simplifies to:

$$\ddot{x} - 100x = 0$$

or

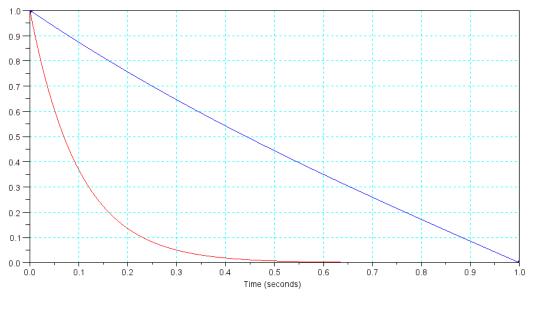
$$(s^2 - 100)x = 0$$
$$s = \pm 10$$

meaning

$$X(t) = ae^{10t} + be^{-10t}$$

which, after solving for the initial conditions, becomes:

$$x = 0.00000002e^{10t} + 1e^{-10t}$$



Optimal Path for $J = \int_0^1 \left(x^2 + u^2\right) dt$ (blue) and $J = \int_0^1 \left(100x^2 + u^2\right) dt$ (red)

Example 3: Find the functional to minimize

$$J = \int_{a}^{b} (X^{T} Q X + U^{T} R U) dt$$

subject to the constraint

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$$X = AX + BU$$

Solution: The functional becomes with a LaGrange multiplier:

$$F = (X^T Q X + U^T R U) + 2M^T \left(A X + B U - \dot{X}\right)$$

The Euler Legrange equations are then

$$2X^{T}Q + 2M^{T}A - \frac{d}{dt}(-2M^{T}) = 0$$
$$\dot{M}^{T} = X^{T}Q - M^{T}A$$
$$\dot{M} = -QZ - A^{T}M$$

and

$$2U^{T}R + 2M^{T}B = 0$$
$$RU = -B^{T}M$$

$$U = -R^{-1}B^{T}M$$

so you have the dyamic system

$$\begin{bmatrix} \dot{X} \\ \dot{M} \end{bmatrix} = \begin{bmatrix} A & -BR^{-1}B^T \\ -Q & -A^T \end{bmatrix} \begin{bmatrix} X \\ M \end{bmatrix}$$

which can be solved subject to the constraints on X(a) and X(b)

Full-State Feedback Formulation:

Assume that

$$M = PX$$

so that the full-state feedback gains are

$$K = R^{-1}B^T P$$

Then the dynamics become

$$X = (A - BR^{-1}B^{T}P)X$$
$$\dot{P}X + \dot{P}X = (-Q - A^{T}P)X$$
$$\dot{P}X = (\dot{P} - Q - A^{T}P)X$$

This implies that

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$$PA - PBR^{-1}B^{T}P = -\dot{P} - Q - A^{T}P$$

or

$II = -A I = IA = U \pm IDA DI$	Algebraic Ricatti equation for computing the time-varying
	feedback gains: U =-KX

This gives the optimal time-varying feedback gain. If the feedback gains are to be constants, then

$$\dot{P}=0$$

and

$$0 = -A^{T}P - PA - Q + PBR^{-1}B^{T}P$$

$$K = -R^{-1}B^{T}P$$
Algebraic Ricatti equation you'll see in most places

Example: For the first-order system

$$\dot{x} = u$$
$$J = \int_0^\infty (qx^2 + ru^2) dt$$

m is

$$0 = -m^2/r + q$$

or

$$m = \sqrt{qr}$$
$$k = \sqrt{q/r}$$

Note that

- Only the ratio of q/r matters not their absolute values. This is reasonable since U(t) minimizes a functional. The minimum of F() will also be the minimum of 10F().
- As Q increases, the poles shift left (faster) as the square root of Q
- As R incrases, the poles shift right (slower) as the square root of R

Example 4:

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$$\dot{x} = -x + u$$
$$J = \int_{0}^{\infty} (x^{2} + u^{2}) dt$$
$$q = 1$$
$$r = 1$$

The Ricatti equation becomes

$$0 = -A^{T}P - PA - Q + PBR^{-1}B^{T}P$$

$$0 = -2p - 1 + p^{2}$$

$$p = \{ 0.4142, -2.4142 \}$$

$$k = \{ 0.4142 - 2.4142 \}$$

This is a typical result.

- P (the Ricatti equation) is a quadratic equation hence generally there are two solutions
- One of these solutions will be a minimum, the other a maximum. Since the feedback gain of -2.4142 results in an unstable system, that is the wrong solution (the maximum). Select the one that stabilizes the system.

The optimal feedback gain is

$$k = 0.4142$$

 $u = -kx$