

Suggested solutions. Exam Maj 23 2013 in SF2852 Optimal Control.

1. **Solution:** We use PMP. The Hamiltonian is

$$H(x, u, \lambda) = u^2 - \lambda x + \lambda u$$

Pointwise minimization gives $u^*(t) = -\lambda(t)/2$. The adjoint equation is

$$\dot{\lambda} = \lambda$$

and there is no boundary constraint specified on λ . This implies $u^*(t) = -\lambda(t)/2 = -e^t \lambda_0$, where λ_0 is a free parameter that must be chosen such that the state constraint is satisfied (this means we have chosen $\lambda(t) = e^t \lambda(0)$ with $\lambda(0) = 2\lambda_0$). We have

$$x(1) = e^{-1} - \int_0^1 e^{-(1-t)} e^t dt \lambda_0 = e^{-1} - \frac{\lambda_0}{2} (e^1 - e^{-1}) = 0$$

This implies

$$\lambda_0 = \frac{1}{e \sinh(1)}.$$

The optimal control is $u^*(t) = -e^t \lambda_0$ and the optimal cost is

$$\int_0^1 (u^*(t))^2 dt = \frac{\lambda_0^2}{2} (e^2 - 1)$$

2. **Solution:**

(a) The ARE becomes $-2p + 1 = p^2$, which gives $p = -1 \pm \sqrt{2}$. The positive definite solution $p = \sqrt{2} - 1$ corresponds to the stabilizing solution. We get

i. The optimal stabilizing feedback control $u = (1 - \sqrt{2})x$.

ii. The optimal cost $J(x(0)) = x(0)^2 p = \sqrt{2} - 1$.

(b) The closed loop system becomes

$$\dot{x} = -x + (1 - \sqrt{2})x = -\sqrt{2}x$$

Hence, the closed loop pole is at $s = -\sqrt{2}$.

(c) HJBE gives rise to the Riccati equation

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

Separation of variables gives

$$\begin{aligned} \frac{dp}{(p+1+\sqrt{2})(p+1-\sqrt{2})} &= 2\sqrt{2}dt \\ \Leftrightarrow \ln\left(\frac{p(t)+1-\sqrt{2}}{p(t)+1+\sqrt{2}}\right) &= 2\sqrt{2}t+c \\ \Leftrightarrow \frac{p(t)+1-\sqrt{2}}{p(t)+1+\sqrt{2}} &= e^{2\sqrt{2}t+c} \end{aligned}$$

The boundary condition $p(T) = 0$ gives $c = -2\sqrt{2}T + \ln((1 - \sqrt{2})/(1 + \sqrt{2}))$. Hence,

$$p(t, T) = \frac{1 - e^{2\sqrt{2}(t-T)}}{1 + \sqrt{2} + (\sqrt{2} - 1)e^{2\sqrt{2}(t-T)}}$$

The optimal feedback solution is $u(t) = p(t, T)x(t)$ and the optimal cost-to-go is $J(t, x) = p(t, T)x^2$.

(d) We have

$$\lim_{T \rightarrow \infty} p(t, T) = \frac{1}{1 + \sqrt{2}} = \sqrt{2} - 1$$

which is the same as the stabilizing solution to the ARE in problem (a).

3. Solution:

(a) Taking the logarithm of the cost we want to maximize

$$\log L(x_1, \dots, x_T) = -\frac{T-1}{2} \log(2\pi) - \frac{1}{2} \sum_{t=1}^{T-1} (x_t - x_{t+1})^2, \quad (1)$$

or equivalently minimizing

$$\sum_{t=1}^{T-1} (x_t - x_{t+1})^2.$$

This can be written as the dynamic programming problem

$$\min \sum_{t=0}^{T-1} f_0(t, x_t, u_t) \quad \text{subject to} \quad \begin{cases} x_{t+1} = x_t + u_t \\ x_0 = 0 \\ x_t + u_t \in M_{t+1} \\ \text{for } t = 0, 1, \dots, T-1 \end{cases}$$

where

$$\begin{aligned}\phi(x_T) &= 0 \\ f_0(0, 0, u_0) &= 0 \\ f_0(t, x_t, u_t) &= u_t^2.\end{aligned}$$

The optimal cost is then obtained from (1).

(b) Let

$$J(k, x_k) = \min \sum_{t=k}^{T-1} f_0(t, x_t, u_t) \quad \text{subject to} \quad \begin{cases} x_{t+1} = x_t + u_t \\ x_t + u_t \in M_{t+1} \\ \text{for } t = k, k+1, \dots, T-1 \end{cases}$$

for $x_k \in M_k$ and $k \in 0, 1, \dots, T$.

Using dynamic programming, the backward recursion

$$\begin{aligned}J(k, x) &= \min_{u_k + x_k \in M_{k+1}} \{f_0(k, x_k, u_k) + J(k+1, x_k + u_k)\}, \quad k = T-1, T-2, \dots, 0 \\ J(T, x) &= 0\end{aligned}$$

gives the optimal solution. From this, we have

$$J(4, x) = 0, \quad x \text{ for } \in M_4$$

$$J(3, 4) = \min_{x \in M_4 = \{2, 7, 8\}} (x - 4)^2 + J(4, x) = 4, \quad (x_4 = 2),$$

$$J(3, 6) = \min_{x \in M_4 = \{2, 7, 8\}} (x - 6)^2 + J(4, x) = 1, \quad (x_4 = 7),$$

$$J(3, 8) = \min_{x \in M_4 = \{2, 7, 8\}} (x - 8)^2 + J(4, x) = 0, \quad (x_4 = 8),$$

$$J(2, 2) = \min_{x \in M_3 = \{4, 6, 8\}} (x - 2)^2 + J(3, x) = 8, \quad (x_3 = 4),$$

$$J(2, 5) = \min_{x \in M_3 = \{4, 6, 8\}} (x - 5)^2 + J(3, x) = 2, \quad (x_3 = 6),$$

$$J(2, 8) = \min_{x \in M_3 = \{4, 6, 8\}} (x - 8)^2 + J(3, x) = 0, \quad (x_3 = 8),$$

$$J(1, 2) = \min_{x \in M_2 = \{2, 5, 8\}} (x - 2)^2 + J(2, x) = 8, \quad (x_2 = 2),$$

$$J(1, 4) = \min_{x \in M_2 = \{2, 5, 8\}} (x - 4)^2 + J(2, x) = 3, \quad (x_2 = 5),$$

$$J(0, 0) = \min_{x \in M_1 = \{2, 4\}} J(1, x) = 3, \quad (x_1 = 4).$$

The optimal path is $(4 \rightarrow 5 \rightarrow 6 \rightarrow 7)$, and the likelihood is

$$L(x_1, \dots, x_T) = \frac{1}{(2\pi)^{3/2}} \exp\left(-\frac{1}{2}J(0, 0)\right) = \frac{e^{-3/2}}{(2\pi)^{3/2}}.$$

4. Solution:

The Hamiltonian is given by

$$H(x, u, \lambda) = (u - x)^2 + \lambda(ax + u),$$

and pointwise minimization gives

$$0 = \frac{\partial H}{\partial u} \implies u = x - \frac{\lambda}{2}.$$

The Hamiltonian is zero, hence we get

$$0 = H = \lambda \left((1 + a)x - \frac{\lambda}{4} \right)$$

which has the two solutions $\lambda \in \{0, 4(1 + a)x\}$. In (a) the solution $\lambda = 0$ is not stabilizing, and the optimal solution is

$$u = x - 2(1 + a)x \implies \dot{x} = -(1 + a)x$$

This gives $x(t) = e^{-(1+a)t}x_0$ and noting that $x - u = 2(1 + a)x$ we have that

$$\int_0^\infty (x - u)^2 dt = 4(1 + a)^2 x_0^2 \int_0^\infty e^{-2(1+a)t} dt = 2(1 + a)x_0^2.$$

In (b), $u = x$, which corresponds to $\lambda = 0$, is stabilizing and the optimal cost is hence 0.

In (c), the solution $u = x$ gives cost zero, but is not stabilizing since then $x \equiv x_0$. The cost can be made arbitrary close to zero by letting $u = (1 - \epsilon)x$ with $\epsilon > 0$ tending to zero.

5. Solution:

(a) Letting $y(t) = \int_0^t u(\tau)d\tau$, the optimization becomes

$$\max \int_0^T x(t)dt \quad \text{subject to} \quad \begin{cases} \dot{x} = -x^2 + u \\ \dot{y} = u \\ x(0) = 0 \\ y(0) = 0, \quad y(T) = K, \\ 0 \leq u(t) \leq 1. \end{cases}$$

The Hamiltonian corresponding to the problem is

$$H(x, u, \lambda) = x + \lambda(-x^2 + u) + \mu u$$

where (λ, μ) are the dual variables. The dynamics for the dual equations are given by

$$\begin{aligned}\dot{\lambda} &= -\frac{\partial H}{\partial x} = -1 + 2\lambda x \\ \dot{\mu} &= -\frac{\partial H}{\partial y} = 0.\end{aligned}$$

The boundary condition of the dual system is $\lambda(T) = 0$.

- (b) The optimal u is the maximizing argument the Hamiltonian, hence

$$\arg \max_{u \in [0,1]} H((x, y), u, (\lambda, \mu)) = \arg \max_{u \in [0,1]} u(\lambda + \mu) = \begin{cases} 1 & \lambda + \mu > 0 \\ ? & \lambda + \mu = 0 \\ 0 & \lambda + \mu < 0. \end{cases}$$

The control in the second case only affect the dynamics if $\lambda + \mu = 0$ on an interval I . Since $\dot{\mu} = 0$ we must have $0 = \dot{\lambda} = -1 + 2\lambda x$ on I , hence $\lambda x = 1/2$. For this to hold, we must have $\dot{x} = 0$ as well, hence $u = x^2$ on I (note that $\lambda \neq 0$ since $\dot{\lambda} = 0$ and $\lambda = 0$ cannot hold simultaneously). The maximizing control is therefore

$$\arg \max_{u \in [0,1]} H((x, y), u, (\lambda, \mu)) = \arg \max_{u \in [0,1]} u(\lambda + \mu) = \begin{cases} 1 & \lambda + \mu > 0 \\ x^2 & \lambda + \mu = 0 \\ 0 & \lambda + \mu < 0. \end{cases}$$

except possibly at isolated points.

- (c) Note first that $0 \leq x(t) < 1$ for any control on $t \in [0, T]$. Due to this, $\dot{\lambda} < 0$ for $\lambda < 1/2$. Therefore $\lambda(T) = 0$ implies that $\lambda(t) > 0$ for all $t \in [0, T)$. We now have $\mu < 0$ since otherwise $\lambda + \mu > 0$ for $t \in [0, T)$ and hence $u(t) = 1$ for $t \in [0, T)$, which is an infeasible control.

In the phase plane (x, λ) there are five regions of interest.

- (1) : $x\lambda < 1/2, \lambda + \mu \geq 0$ control $u = 1$
- (2) : $x\lambda = 1/2, \lambda + \mu = 0$ control $u = x^2$
- (3) : $x\lambda \leq 1/2, \lambda + \mu < 0$ control $u = 0$
- (4) : $x\lambda \geq 1/2, \lambda + \mu > 0$ invariant (cannot satisfy $\lambda(T) = 0$)
- (5) : $x\lambda > 1/2, \lambda + \mu \leq 0$ can never be reached.

The only regions that can be part of an optimal trajectory are the first three listed above. At $t = 0$, the state is in (1), since $x(0) = 0$. At time T , the dual state is in (3) since $\lambda(T) = 0$. Note

that here $x\lambda \leq 1/2$, hence $\dot{\lambda} \leq 0$. The only possible sequence of states are hence (1) \rightarrow (2) \rightarrow (3) or (1) \rightarrow (3). The possible switching sequences are hence $u = (1, x^2, 0)$ or $(1, 0)$.

(d) Let $K = 1$ and $T > K$. First we want to bound the cost using the control

$$u(t) = \begin{cases} 1 & t < 1 \\ 0 & t \geq 1. \end{cases} \quad (2)$$

First note that $\int_0^1 x dt \leq 1$ since $x(t) \leq 1$. Secondly, for $t \geq 1$, $\dot{x} = -x^2$ implies that

$$x(t) = \frac{1}{t - 1 + \frac{1}{x(1)}}.$$

Integrating, we get

$$\int_1^T x dt = \log((T - 1)x(1) + 1) \leq \log(T),$$

hence $\int_0^T x dt \leq 1 + \log(T)$ using this control.

Compare this to the control $u \equiv 1/T$. The ODE $\dot{x} = -x^2 + 1/T$ with $x(0) = 0$ has the solution

$$x(t) = \left(1 - 2 \left(1 + e^{2t/\sqrt{T}}\right)^{-1}\right) / \sqrt{T}.$$

When $t \geq \sqrt{T}$ it holds that

$$x(t) \geq \left(1 - \frac{1}{1 + e^2}\right) / \sqrt{T} \geq \frac{1}{2\sqrt{T}}$$

hence

$$\int_0^T x dt \geq \int_{\sqrt{T}}^T x dt \geq \frac{T - \sqrt{T}}{2\sqrt{T}} = (\sqrt{T} - 1)/2.$$

Since $\log(T)/\sqrt{T} \rightarrow 0$ as $T \rightarrow \infty$, it is possible to pick T such that $(\sqrt{T} - 1)/2 > 1 + \log(T)$, (e.g., $T = 1000$). For such T , the control (2) is not optimal, hence the optimal switching sequence for these parameters is $u = (1, x^2, 0)$.