

Homework to be done for the 14/10/2015 The three exercises are independent. When possible underline the final answers.

Exercise 1 (5pts). Consider a control dynamical system following

$$\mathbf{x}_{t+1} = f_t(\mathbf{x}_t, \mathbf{u}_t, \mathbf{w}_{t+1}), \quad (1)$$

where

- the state \mathbf{x}_t belongs to a Cartesian product finite set $\mathbb{X} \times \cdots \times \mathbb{X} = \mathbb{X}^{n_x}$;
- the control \mathbf{u}_t belongs to a Cartesian product finite set $\mathbb{U} \times \cdots \times \mathbb{U} = \mathbb{U}^{n_u}$;
- the noise \mathbf{w}_t belongs to a Cartesian product finite set $\mathbb{W} \times \cdots \times \mathbb{W} = \mathbb{W}^{n_w}$.

We suppose that $(\mathbf{w}_t)_{t \in \mathbb{N}}$ is a sequence of independent and identically distributed random variables.

We consider the following optimization problem

$$\min \quad \mathbb{E} \left[\sum_{t=0}^{T-1} L_t(\mathbf{x}_t, \mathbf{u}_t, \mathbf{w}_{t+1}) + K(\mathbf{x}_T) \right] \quad (2a)$$

$$s.t. \quad \mathbf{x}_{t+1} = f_t(\mathbf{x}_t, \mathbf{u}_t, \mathbf{w}_{t+1}), \quad \mathbf{x}_0 = x_0 \quad (2b)$$

$$\mathbf{u}_t \in \mathbb{U}, \quad \mathbf{u}_t \preceq \sigma(\mathbf{w}_1, \dots, \mathbf{w}_t) \quad (2c)$$

1. Write, in the form of a pseudo-code, the Dynamic Programming algorithm attached to the optimization problem.
2. What is the complexity of the Dynamic Programming algorithm in terms of the cardinals of the sets \mathbb{X}^{n_x} , \mathbb{U}^{n_u} , \mathbb{W}^{n_w} and of the horizon T ?
3. If the optimization problem were written on a tree (with one decision per node), how many decisions are there in total?

Solution 1. 1.

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V_T ≡ K ;
for t : T - 1 → 0 do
  for x ∈ X_t^{n_x} do
    v = -∞;
    for u ∈ U^{n_u} do
      Vmoy = 0;
      for w ∈ W^{n_w} do
        | Vmoy = Vmoy + P(w)(L_t(x, u, w_{t+1}) + V_{t+1} ∘ f_t(x, u, w_{t+1}));
      end
      v = min {v, Vmoy};
    end
    V_t(x) = v;
  end
end
end

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2. The number of operations is in $O(|\mathbb{X}|^{n_x} \times |\mathbb{U}|^{n_u} \times |\mathbb{W}|^{n_w})$

3. At stage t there is $(|\mathbb{W}|^{n_w})^t$ decisions, hence there is $\frac{|\mathbb{W}|^{(T+1)n_w} - 1}{|\mathbb{W}|^{n_w}}$.

Exercise 2 (8pts). Consider the controlled dynamic system given by

$$\mathbf{x}_{t+1} = \mathbf{x}_t + \mathbf{u}_t + \mathbf{w}_{t+1},$$

where (\mathbf{w}_t) is a sequence of independent, centered (i.e. $\mathbb{E}[\mathbf{w}_t] = 0$) exogeneous noises of standard-deviation σ_t . The control \mathbf{u}_t is taken knowing the past noises $\mathbf{w}_1, \dots, \mathbf{w}_t$ (in particular, \mathbf{u}_0 is deterministic and denoted u_0).

We consider the following optimization problem

$$\min_{u_0, \mathbf{u}_1} \mathbb{E} \left[u_0^2 + k\mathbf{u}_1^2 + \mathbf{x}_2^2 \right] \quad (3a)$$

$$s.t. \quad x_0 \text{ given} \quad (3b)$$

$$\mathbf{x}_1 = x_0 + u_0 + \mathbf{w}_1 \quad (3c)$$

$$\mathbf{x}_2 = \mathbf{x}_1 + \mathbf{u}_1 + \mathbf{w}_2 \quad (3d)$$

$$\mathbf{u}_1 \preceq \mathbf{w}_1 \quad (3e)$$

where $k > 0$ is a given parameter.

1. Identify the elements of a stochastic optimization control problem: what is the state variable? the control variable? the terminal time T ? the final cost function K ? the time-step cost functions L_t ? the initial state?
2. Why can we apply the Dynamic Programming approach to this problem? Hence, what is the form assumed by the optimal solution?
3. Determine the Bellman function V_1 at time 1, and the optimal strategy $\mathbf{u}_1^\#$ at time 1.
4. Determine the Bellman function V_0 at time 0, and the optimal control u_0 at time 0.
5. What is the optimal value of the optimization problem?
6. What are the optimal trajectories $t \mapsto \mathbf{x}_t^\#$ and $t \mapsto \mathbf{u}_t^\#$?
7. Interpret the behavior of the optimal trajectories $t \mapsto \mathbf{x}_t^\#$ and $t \mapsto \mathbf{u}_t^\#$, when $k \rightarrow +\infty$.

Solution 2. 1. The state variable is \mathbf{x}_t , the control variable \mathbf{u}_t , the terminal time $T = 2$, the final cost function $K(x) = kx^2$, the time-step cost functions $L_1(x, u, w) = x^2$, $L_0(x, u, w) = 0$, the initial state $x_0 = 0$.

2. The noises are independent, hence we can apply the DP approach and the optimal control are function of the current state.

3. By DP we have

$$\begin{aligned}
 V_1(x) &= \min_{u \in \mathbb{R}} \mathbb{E}[ku^2 + (x + u + \mathbf{w}_2)^2] \\
 &= \min_{u \in \mathbb{R}} (1+k)u^2 + x^2 + 2xu + 2(x+u)\mathbb{E}[\mathbf{w}_2] + \mathbb{E}[\mathbf{w}_2^2] && \text{by linearity of } \mathbb{E} \\
 &= \min_{u \in \mathbb{R}} (1+k)u^2 + x^2 + 2xu + \sigma^2 && \text{as } \mathbb{E}[\mathbf{w}_2] = 0 \\
 &= -\frac{1}{1+k}x^2 + x^2 + \sigma^2 && \text{with } u^\# = -\frac{x}{1+k} \\
 &= \frac{k}{1+k}x^2 + \sigma^2
 \end{aligned}$$

and $u_1^\# = -\frac{x_1}{1+k}$.

4. By DP we have with $\kappa = \frac{k}{1+k}$

$$\begin{aligned}
 V_0(x) &= \min_{u \in \mathbb{R}} \mathbb{E}[u^2 + \kappa(x + u + \mathbf{w}_2)^2] + \sigma^2 \\
 &= \min_{u \in \mathbb{R}} (1+\kappa)u^2 + \kappa x^2 + 2\kappa xu + (1+\kappa)\sigma^2 \\
 &= -\frac{\kappa^2}{1+\kappa}x^2 + x^2 + \sigma^2 && \text{with } u^\# = -\frac{\kappa x}{1+\kappa}
 \end{aligned}$$

and $u_0^\# = -\kappa x_0 / (1 + \kappa)$. The optimal value is given by $V_0(x_0)$.

5. If k tends to ∞ the second control is extremely costly and thus the optimal second control is 0.

Exercise 3 (7pts). You are the IT manager of a small company. You use a software for a period of 4 years. The software needs to be updated every year (else it becomes outdated). It has been bought (intalled up-to-date) on January 1st 2014, and will be changed January 1st 2018.

- If the software is up-to-date at the beginning of a year, its efficiency is evaluated at 100k\$ for the year.
- If the software is outdated at the beginning of a year, it is less efficient and only accounts for 80k\$.
- If the software is broken, reinstalling the software cost time and money and its value for one year is 0; however, the software is up-to-date at the end of the year.
- An outdated software has 25% chance of getting broken after one year of usage if nothing is done.
- When updating an outdated software there is 50% chance of breaking it (else it is up-to-date).

Suppose you want to optimize the expected value of the software on between January 1st 2014, and January 1st 2018.

1. Find a Controlled Markov chain representing the problem.
2. Solve the problem by dynamic programming: give the Bellman value function, the optimal policy, and the expected value of the software over the period. The Bellman function and policy can be presented as a table.