

Trace property: For a gaussian random variable a with mean \bar{a} and covariance A

$$\mathbb{E}(a^T M a) = \bar{a}^T M \bar{a} + \text{Tr}(M A)$$

Looking at $t = T - 1$

$$\begin{aligned}
J_t(\pi_t) &= \min_{\mu_t: T-1} \sum_{k=t}^{T-1} \mathbb{E}(\|x_k\|_{Q_k}^2 + \|u_k\|_{R_k}^2) + \mathbb{E}\|x_T\|_{Q_T}^2 \\
J_{T-1}(\pi_{T-1}) &= \mathbb{E}(\|x_k\|_{Q_k}^2 + \|u_k\|_{R_k}^2) + \mathbb{E}\|X_t\|_{Q_t}^2 \\
Q_T = S_T \implies J_{T-1}(\pi_{T-1}) &= \mathbb{E}(\|x_k\|_{Q_k}^2 + \|u_k\|_{R_k}^2) + \mathbb{E}(x_T^T S_T x_T) \\
x_{t+1} = A_t x_t + B_t u_t + w_t \implies \mathbb{E}(x_T^T S_T x_T) &= \mathbb{E}((A_{T-1} x_{T-1} + B_{T-1} u_{T-1} + w_{T-1})^T S_T (A_{T-1} x_{T-1} + B_{T-1} u_{T-1} + w_{T-1})) \\
\mathbb{E}(w_t) = 0 \implies \mathbb{E}(x_T^T S_T x_T) &= \mathbb{E}((A_{T-1} x_{T-1} + B_{T-1} u_{T-1})^T S_T (A_{T-1} x_{T-1} + B_{T-1} u_{T-1})) + \text{Tr}(S_T W_{T-1}) \\
\implies J_{T-1}(\pi_{T-1}) &= \mathbb{E}(x_{T-1}^T Q_{T-1} x_{T-1} + u_{T-1}^T R_{T-1} u_{T-1} + (A_{T-1} x_{T-1})^T S_T (A_{T-1} x_{T-1}) + 2(B_{T-1} u_{T-1})^T S_T (A_{T-1} x_{T-1}) \\
&\quad + (B_{T-1} u_{T-1})^T S_T B_{T-1} u_{T-1}) + \text{Tr}(S_T W_{T-1}) \\
M_t = B_t^T S_t B_t + R_t \implies J_{T-1}(\pi_{T-1}) &= \mathbb{E}(x_{T-1}^T Q_{T-1} x_{T-1} + u_{T-1}^T M_{T-1} u_{T-1} + (A_{T-1} x_{T-1})^T S_T (A_{T-1} x_{T-1}) \\
&\quad + 2(B_{T-1} u_{T-1})^T S_T (A_{T-1} x_{T-1}) + \text{Tr}(S_T W_{T-1}) \\
&\quad \text{Completing the square} \\
&\quad u_{T-1}^T M_{T-1} u_{T-1} + 2(B_{T-1} u_{T-1})^T S_T (A_{T-1} x_{T-1}) \\
&= (u_{T-1} + M_{T-1}^{-1} B_{T-1}^T S_T A_{T-1} x_{T-1})^T M_{T-1} (u_{T-1} + M_{T-1}^{-1} B_{T-1}^T S_T A_{T-1} x_{T-1}) \\
&\quad - (B_{T-1}^T S_T A_{T-1} x_{T-1})^T M_{T-1}^{-1} (B_{T-1}^T S_T A_{T-1} x_{T-1}) \\
&\quad \text{know } S_t = A_t^T S_{t+1} A_t - A_t^T S_{t+1} B_t M_t^{-1} B_t^T S_{t+1} A_t + Q_t \\
\implies J_{T-1}(\pi_{T-1}) &= \mathbb{E}(x_{T-1}^T S_{T-1} x_{T-1} + (u_{T-1} + M_{T-1}^{-1} B_{T-1}^T S_{T-1} A_{T-1} x_{T-1})^T M_{T-1} (u_{T-1} + M_{T-1}^{-1} B_{T-1}^T S_{T-1} A_{T-1} x_{T-1})) \\
&\quad + \text{Tr}(S_T W_{T-1}) \\
&\quad \text{know } u^* = -M_t^{-1} B_t^T S_t A_t \hat{x}_t \\
&\implies J_{T-1}(\pi_{T-1}) = \mathbb{E}(x_{T-1}^T S_{T-1} x_{T-1} \\
&\quad + (M_{T-1}^{-1} B_{T-1}^T S_{T-1} A_{T-1} (\hat{x}_{T-1} - x_{T-1}))^T M_{T-1} (M_{T-1}^{-1} B_{T-1}^T S_{T-1} A_{T-1} (\hat{x}_{T-1} - x_{T-1}))) \\
&\quad \text{know } \Theta_t = K_t^T M_t K_t \text{ and } K_t = -M_t^{-1} B_t^T S_t A_t \\
\implies J_{T-1}(\pi_{T-1}) &= \mathbb{E}(x_{T-1}^T S_{T-1} x_{T-1} + (\hat{x}_{T-1} - x_{T-1})^T \Theta_{T-1} (\hat{x}_{T-1} - x_{T-1})) + \text{Tr}(S_T W_{T-1}) \\
&\quad \text{From trace rules } \mathbb{E}(x_{T-1}^T S_{T-1} x_{T-1}) = \hat{x}_{T-1}^T S_{T-1} \hat{x}_{T-1} + \text{Tr}(S_{T-1} P_{T-1|T-1}) \\
\implies J_{T-1}(\pi_{T-1}) &= \hat{x}_{T-1}^T S_{T-1} \hat{x}_{T-1} + \text{Tr}(S_{T-1} P_{T-1|T-1}) + \text{Tr}(\Theta_{T-1} P_{T-1|T-1}) + \text{Tr}(S_T W_{T-1})
\end{aligned}$$

Now we need to get the value function for $t = T - 2, T - 3, \dots, 1, 0$. This can be done in a similer way to in exercise 2.4) Finally,

$$J_t(\pi_t) = \|\hat{x}_t\|_{S_t}^2 + \text{Tr}(S_t P_{t|t}) + \sum_{k=1}^{T-1} (\text{Tr}(W_k S_{k+1}) + \text{Tr}(\Theta_k P_{k|k}))$$