

Maintaining Robust Stability and Performance through Sampling and Quantization

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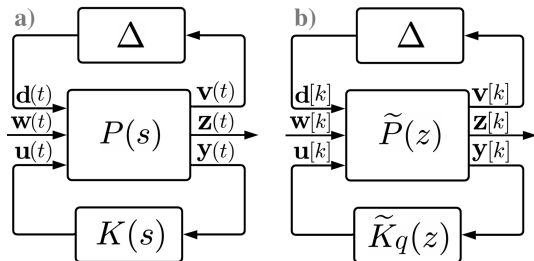
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Robust Control Context



The numeric regulator $\tilde{K}_q(z)$ is obtained through:

- **sampling with a fixed period** $\tau \in \mathbb{R}_+$, with discretization;
- **uniform coefficient quantization** with fixed step $q \in \mathbb{R}_+$.

Denote $\xi = (\tau, q) \in \mathbb{R}_+^2$ single two-dimensional variable.

Robust Control: μ -Synthesis

Structured singular value (SSV):

$$\mu_{\Delta}(\text{LLFT}(P, K)) = \sup_{\omega \in \mathbb{R}_+} \frac{1}{\min_{\Delta \in \Delta} \{\bar{\sigma}(\Delta), \det(I - M_{\omega}\Delta) = 0\}}}, \quad (1)$$

with $M_{\omega} = \text{LLFT}(P, K)(j\omega)$.

μ -synthesis problem:

$$K \text{ stabilizing s.t. } \mu_{\Delta}(\text{LLFT}(P, K)) < 1. \quad (2)$$

Assumed closed-loop properties imposed through $K(s)$:

- **robust stability (RS)**;
- **robust performance (RP)**.

Key Questions

- ① How to guarantee that $\tilde{K}_q \in \mathcal{G}_D$ still maintains RS and RP?
- ② How to select such $\tilde{K}_q \in \mathcal{G}_D$ configurations?

Contributions

Lemma

Given the continuous models G_n , U , W , P and mappings \mathcal{T} , \mathcal{A} , the discrete augmented plant counterpart \tilde{P} can be computed based on the discretization of its individual components \tilde{G}_n , \tilde{U} , \tilde{W} through $\mathcal{D}\{\cdot, \tau\}$, $\tau > 0$. Moreover, the block Δ is invariant from the continuous domain to the discrete domain. As such:

$$\tilde{G} = \mathcal{T}(\mathcal{D}\{G_n, \tau\}, \mathcal{D}\{U, \tau\}), \Delta \in \Delta; \quad (3)$$

$$\tilde{P} = \mathcal{D}\{P, \tau\} = \mathcal{A}(\tilde{G}, \mathcal{D}\{W, \tau\}). \quad (4)$$

Main advantage: Reuse the models and weights from the continuous to discrete case.
Reformulate the μ -synthesis problem in terms of the equivalent discrete-time models:

$$\mu_{\Delta}(\text{LLFT}(\tilde{P}, \tilde{K}_q)) = \sup_{\omega \in \Omega_N} \frac{1}{\min_{\Delta \in \Delta} \{\bar{\sigma}(\Delta), \det(I - \tilde{M}_{\omega} \Delta) = 0\}} < 1 \text{ (mandatory constraint)},$$

with $\tilde{M}_{\omega} = \text{LLFT}(\tilde{P}, \tilde{K}_q)(e^{j\omega\tau})$ and domain $\Omega_N = [0, \omega_N)$, where $\omega_N = \pi/\tau$ is the Nyquist frequency for the period τ .

Optimization Functionals

In addition to the mandatory SSV constraint, we are left with a **separate degree of freedom** to select the pairs $\xi = (\tau, q) \in \mathbb{R}_+^2$. Two dichotomic approaches are suggested:

- **Implementability functional:**

$$\min_{\xi \in \mathbb{R}_+^2} f_1(\xi) = -\xi_1 \xi_2. \quad (5)$$

- **Fidelity functional:**

$$\min_{\xi \in \mathbb{R}_+^2} f_2(\xi) = \mathcal{J}(\xi), \quad (6)$$

$$\mathcal{J}(\xi) = \int_{\Omega} \left| \bar{\sigma}(K) - \bar{\sigma}(\tilde{K}_q) \right| (1 + \|\nabla^2 K\|) d\omega. \quad (7)$$

Hands-On Summary

Preconditions:

- ① establish $G_n, \Delta, U \rightarrow \mathcal{T}, W \rightarrow \mathcal{A} \Rightarrow K$;
- ② define discretization methods $\mathcal{D}_p, \mathcal{D}_c$;
- ③ define quantization method \mathcal{Q} for \tilde{K}_q ;
- ④ discretize plant \tilde{P} (Lemma 1);
- ⑤ define cost functional $f : \mathbb{R}_+^2 \rightarrow \mathbb{R}_+$;
- ⑥ define nonlinear constrained optimization problem:

$$\min_{\xi \in \mathbb{R}_+^2} f(\xi) \text{ s.t. } \mu_\Delta(\text{LLFT}(\tilde{P}, \tilde{K}_q)) < 1,$$

\Leftrightarrow the interior-point function, $\rho > 0$:

$$F(\xi) = f(\xi) - \rho \ln \left(1 - \mu_\Delta(\text{LLFT}(\tilde{P}, \tilde{K}_q)) \right).$$

Algorithm 1: Optimal selection of sampling rate and quantization step for a continuous-time regulator K

Input: $K, G_n, U, W, \mathcal{T}, \mathcal{A}, \Delta, F \in \{F_1, F_2\}, \Omega, \rho$
 Discretization operators $\mathcal{D}_p\{\cdot, \tau\}, \mathcal{D}_c\{\cdot, \tau\}$.
 {For the plant and controller models}
 Solver algorithm $\xi_{k+1} = \Sigma(F, \xi_k)$.

Output: Optimum $\xi^* = (\tau^*, q^*) \in \mathbb{R}_+^2$.

- 1 Initialize $\xi \leftarrow \xi_0 = (\tau_0, q_0) \in \mathbb{R}_+^2$.
 - 2 **while** *stopping criterion is not satisfied* **do**
 - 3 $\tilde{G}_n \leftarrow \mathcal{D}_c\{G_n, \xi_1\}$.
 - 4 $\tilde{U} \leftarrow \mathcal{D}_c\{U, \xi_1\}; \tilde{W} \leftarrow \mathcal{D}_c\{W, \xi_1\}$.
 - 5 $\tilde{G} \leftarrow \mathcal{T}\{\tilde{G}_n, \tilde{U}\}; \tilde{P} \leftarrow \mathcal{A}\{\tilde{G}, \tilde{W}\}$ {Lemma 1}.
 - 6 $\tilde{K}_q \leftarrow \mathcal{Q}\{\mathcal{D}\{K, \xi_1\}, \xi_2\}$ {Eq. (9), (10)}.
 - 7 Compute $\mu_\Delta(\text{LLFT}(\tilde{P}, \tilde{K}_q))$ approximation.
 - 8 Compute $F(\xi)$ {Eq. (16) or (21)}.
 - 9 Update $\xi \leftarrow \Sigma(F, \xi)$.
 - 10 Verify stopping criterion.
 - 11 **end**
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In our experiments: $\Sigma = \text{fmincon} + \text{GlobalSearch}$ initialization.