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Performance Control for Interconnection of Identical Systems: Application to PLL network design

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Abstract

In this paper, the problem of the control law design for interconnected identical

systems ensuring the global stability and the global performance properties is under consideration. Inspired by the decentralized control law design methodology using the dissipativity input-output approach, the problem is reduced to the problem of satisfying two conditions: (i) the condition on the interconnection and (ii) the condition on the local subsystem dynamics. Both problems are efficiently solved applying a (quasi-) convex LMI optimization and standard H_∞ synthesis. The proposed design methodology is applied to the control law design of a synchronous PLL network.

Keywords Multi-Agent systems, decentralized control, H_∞ control, dissipativity input-output approach, LMI optimization, PLL network synchronization.

1 Introduction

The *design* problem of a *large-scale system* represented as an interconnection of *identical subsystems* is under consideration. Each subsystem is a feedback interconnection of the *plant* and the *controller*. The problem of system design is: given the plant and the interconnection topology, find the subsystem controllers ensuring an appropriate *global behavior* of the overall system, referred to as the *global system performance*.

This design problem turns out to be relevant in various fields of science and engineer applications [1]. In this paper, as an example of application, we investigate the control law design for an active clock distribution network, that is, interconnected Phase-Locked Loops (PLLs) interconnected in an array. This device was proposed as an alternative to the classical clock distribution trees in order to distribute the clock signal in a synchronous multi-core microprocessor system with a better level of performance in terms of perturbation

rejection, robustness and power consumption [2–5]. The active clock distribution network design consists in computing the controller (filter) of the (feedback) PLL loop in order to ensure a high level of global system performance. Indeed, the global synchronization of the phase and frequency in a specified amount of time is crucial to ensure the correct operation of the clock network. Issues of minimal global system bandwidth, maximal control signal level, noise and perturbation rejection must be addressed as well. The two first requirements enforce the convergence speed and limit the power consumption of the whole system, while the two last ones are critical to distribute the clock properly in a noisy and often perturbed chip environment.

In the Microelectronics field, the traditional design procedure of the PLL network system usually consists of two steps: *(i) local design* of the PLL (filter) and *(ii) global analysis* of the PLL network. During the first step, the network interconnection is neglected, that is, the subsystems (PLLs) are assumed independent. The local subsystem design, usually based on the classical frequency design methods, is then performed such that the local subsystem behavior or *local subsystem performance* are satisfactory. The second step consists in the performance analysis of the global system composed of the designed subsystems interconnected with the given topology. Usually, to perform this step, some local and global simulations are performed in order to test if the local performance ensured by the local design is achieved for the overall system as well. If it is not the case, the *difference* between the local performance and global one is then estimated *i.e.* the global performance is rather evaluated in the *relative* than in the *absolute* way. This design procedure is a reminiscence of some iterative, “trial and error” approaches to decentralized control [6].

The same design procedure, *i.e.* the local subsystem design followed by the global system analysis, can be applied to the design of other large-scale interconnected systems. However, due to the mutual coupling, the multiple interconnections and feedbacks inside the network, the traditional design method can fail. In general, it cannot ensure that the local performance of the independent subsystem will not be significantly degraded when the subsystem is connected to the global network. Moreover, even the global system stability is not ensured. Hence, the impact of the global network interconnections can be very important: it is then necessary to design the control law under the constraint of global stability and global performance.

Though the (heuristic) traditional design method does not ensure *a priori* appropriate global system behavior, there are some motivations for its use. A first motivation is the inherent *complexity* of the networked system design, due to the decentralization and the possibly important dimension. Since the network topology does not connect all the subsystems with all the other ones and since for practical realization reasons, the controllers are decentralized, the design problem is a particular case of the decentralized control design problem. The structural (decentralized) constraint implies the complexity of the design problem and the inefficiency of its resolution algorithms [7]. Second, since the overall system can have a large number of subsystems, its direct global design using optimization can lead to a large number of decision variables. In the traditional approach, the design is first performed for a single local subsystem, that is an optimization problem with a significantly *reduced number* of decision variables. Moreover, since in the considered application the subsystems are identical, they are all designed at once. Another motivation is the existence of the standard

and well-known *local subsystem design tools*, which is also the case for fields different from Microelectronics.

The main goal of this paper is to propose a new design method which is close to the Microelectronics traditional design procedure but solves its major drawbacks *i.e.* the local design of the subsystem does not guarantee the stability and the performance of the network. A strong motivation of procedure close to the traditional one is to improve the acceptability of the proposed design method. Furthermore, it has to be general enough to be applied to other similar design problems of networked systems, in other application fields.

Let us first discuss the existing interconnected system design methods in Control Theory with respect to our goals. A similar control problem was largely investigated in the field of *Multi-Agent system*. A Multi-Agent system is a network of intelligent subsystems called agents, where each agent exchanges some information with its neighbour ones, in order to achieve some specified global network behaviour. The agent is the feedback of a to-be-controlled plant and a controller.

The most fundamental result in the case of Linear Time Invariant (LTI) identical Multi-Agent system is the necessary and sufficient *stability conditions* presented in [8]. By some aspects, this result is a generalization of the Nyquist stability criterion to Multi-Agent systems. Based on the Graph Theory, the authors *transform* the *global stability condition* of the N interconnected identical Multi-Agents into *local conditions of simultaneous stability* for the N independent different subsystems, see Theorem 3 in [8].

Based on this result, a large number of papers were published during the last years for sometimes particular classes of interconnections (Laplacian, Adjacency, symmetric or

normal matrices). The static feedback control of [9] or the dynamic observer based solution presented in [10–12] and [13] are such examples. However, these results cannot be applied to our application of PLL network design. First, for the control law discussed in [10–13] a communication network which is able to transmit arbitrary information is assumed: plant state vector for the static control case and observer based controller state vectors for the dynamic control case. Furthermore, the computation of these controllers and observers is not investigated. The reason is the intrinsic complexity of the simultaneous stabilization problem for three and more subsystems, as discussed in [14]. This fact is confirmed by the results presented in [15, 16] where the control law computation is based on Bilinear Matrix Inequality (BMI) optimization which, in general, cannot be efficiently solved.

Another drawback of these results which is important for our application, is that the proposed control law design methods do not address general performance specifications. The focus is only on the consensus or coordination with a fixed convergence rate, inspired by Theorem 3 in [8]. For example, the specification of the reference ramp tracking, necessary for the PLL synchronization, cannot be considered. The minimization of the H_∞ norm of a global transfer function in [17] is an interesting perspective for considering some performance specifications. However, the solution presented in [17] could be extremely conservative since it is based on a quadratic Lyapunov function which is common to all the subsystems with a *block diagonal structure*. A less conservative result is presented in [18], where, if the interconnection satisfies a structural property, the global system could be equivalently decomposed into independent subsystems. In this case, the design of controller ensuring the global stability and an H_∞ norm inequality on a global transfer function is investigated.

This inequality ensures that the magnitude of the frequency response of the global transfer function is below a frequency *independent* constraint. As in most of the applications, the performance specifications of the PLL network have to be expressed as "the magnitude of the frequency response of the global transfer function is below a frequency *dependent* constraint". The introduction of performance inputs and outputs to the global system associated to weighting functions [19] in order to deal with frequency *dependent* constraint could destroy the system structural property so that the global system might not be decomposed. In this case the approaches presented in [17, 18] cannot be directly applied.

The second possible approach for the large-scale system design is the Decentralized Control design methods, see [20] and [21]. In these approaches, a system is represented as the interconnection of subsystems. A local controller is designed for each subsystem such that the global and sometimes local system objectives are ensured. These (local) controllers form hence the decentralized controller.

Despite a large amount of works, the design of such controllers is still an open problem for the general case [22–24]. Under some assumptions, it is proved to be NP-hard [7] and boils down to a non convex BMI optimization problem [24, 25]. A first approach to solve the problem is to propose a direct but heuristic resolution of the BMI problem [24] or to over constraint the problem to obtain a convex formulation. The last approach is adopted in the paper [22] where a quadratic Lyapunov function is fixed to have a block diagonal matrix structure. A second approach consists in a relaxation of the complex structural constraints imposed by the network topology in order to obtain an efficient design algorithm as in [26]. The authors of [26] propose to solve the problem introducing the link between global system

and local subsystem behaviours based on a subsystem dissipativity characterisation used in the input-output approaches [27].

Nevertheless, the approach presented in [26] is not suitable for our application. In [26], the subsystems are not necessarily identical and thus the complexity of the design conditions strongly depends on the number of subsystems. This potentially leads to large optimization problems since, for every single subsystem, a set of Linear Matrix Inequality (LMI) constraints has to be introduced. The application to the PLL network design can be conservative since this method does not consider identical subsystems, in contrast with the Multi-Agent approaches as in [8]. Furthermore, the proposed approach does not satisfy our goal, that is, propose a procedure where the design of a local controller implies the stability and the performance of the global system.

In this paper, we propose a design method for the control of interconnections of identical subsystems. The proposed approach is to design the (local) controller of a subsystem in order to achieve the stability and the specifications of performance for the interconnections of identical subsystems (the global system). A first result reveals how to compute the local controller such that the global system is stable, with the H_∞ norm of a global transfer function less than a given level η (section 2). The input-output behavior of the subsystem is described by a general dissipative property [27], that is, the input signals and the output signals satisfy an (integral) quadratic constraint. Such an approach is suitable to obtain stability criteria for interconnected systems, using separation of graph arguments [27–30]. The nice feature is that the local controller can be designed by the standard H_∞ method with some additional LMI constraints due to the interconnection. Based on this first result,

a second one reveals how to obtain a similar result with the H_∞ norm inequality replaced by a frequency *dependent* constraint on the frequency response magnitude of the global transfer function (section 3). The nice feature is that this kind of constraints allows to address typical design specifications. The proposed method is then applied to the design of a PLL network (section 4). The nice feature is that we succeed in improving the traditional design process by a priori ensuring the global stability and performance.

Notations The superscript “ T ” defines a real matrix transpose while the superscript “ $*$ ” defines the complex conjugate transpose. I_N denotes a square $N \times N$ identity matrix while $0_{n \times m}$ is a $n \times m$ zero matrix. The dimension of the identity or zero matrix is omitted (I and 0) if it is clear from the context. The diagonal aggregation of two matrices A and B is denoted by $\text{diag}(A, B)$. The *Kronecker product* between two matrices A and B denoted by \otimes is defined as:

$$A \otimes B = [a_{ij}B]$$

With $G = \begin{bmatrix} G_{11} & G_{12} \\ G_{21} & G_{22} \end{bmatrix}$, $G \star K$ denotes $G \star K = G_{11} + G_{12}K(I - G_{22}K)^{-1}G_{21}$ with \star the *Redheffer (star)* product. Similarly, $K \star G$ denotes $K \star G = G_{22} + G_{21}K(I - G_{11}K)^{-1}G_{12}$. For a stable LTI system G , $\|G\|_\infty$ denotes the H_∞ norm of G . For a matrix P , $\bar{\sigma}(P)$ denotes the maximal singular value of the matrix P . \mathcal{L}_2 is the space of \mathbb{R}^n square integrable valued functions defined on \mathbb{R} , where the norm is defined by $\|f\|_2 = (\int \|f(t)\|^2 dt)^{1/2}$.

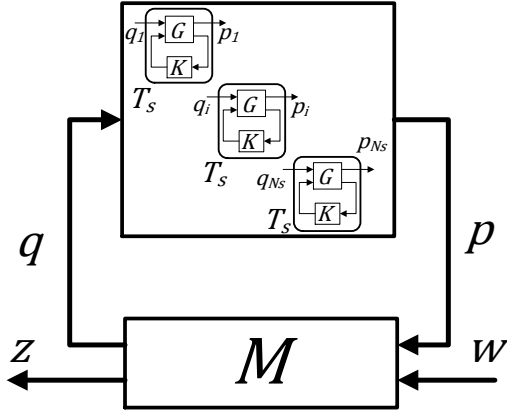


Figure 1: Considered global LTI large-scale system

2 Large-scale control

2.1 Problem formulation

The Linear Time Invariant (LTI) large-scale system with identical subsystems is defined as the interconnection M of N_s identical subsystems T_s with n_s inputs and m_s outputs, that is :

$$\begin{aligned}
 p &= (I_{N_s} \otimes T_s)(q) \\
 \begin{bmatrix} q \\ z \end{bmatrix} &= M \left(\begin{bmatrix} p \\ w \end{bmatrix} \right)
 \end{aligned} \tag{1}$$

where

- $M = \begin{bmatrix} M_{11} & M_{12} \\ M_{21} & M_{22} \end{bmatrix}$ is a (finite dimensional) stable LTI system;
- $T_s = G \star K$ where the (finite dimensional) LTI systems G and K are referred to as the local plant and the local controller;

- $w(t) \in \mathbb{R}^{n_w}$ is the system input, $z(t) \in \mathbb{R}^{n_z}$ the system output and $q(t) \in \mathbb{R}^{n_s N_s}$, $p(t) \in \mathbb{R}^{m_s N_s}$ are internal signals.

In the sequel, this system, denoted $(I_{N_s} \otimes T_s) \star M$ and represented Fig. 1, is referred to as the *global system* and $T_s = G \star K$ as the *local system*. Both systems are assumed to be well posed and causal.

Note that an active clock distribution network can be represented as a Linear Time Invariant (LTI) large-scale system with identical subsystems where the local subsystem T_s represents a Phase-Locked Loops, K the PLL filter and M the interconnection of the PLLs.

Remark 1. The proposed general large-scale system model (1) can describe the network of the interconnected Multi-Agents. In this particular case, the agent dynamic is represented by T_s and their interconnection by a Laplacian (or an Adjacency) matrix M (see [8,31]). In contrast with, in the sequel, the proposed approach does not assume any particular structure of the interconnection.

Problem 1 (Decentralized control problem). *Given $\eta > 0$, the interconnection M and the local plant G , find the local controller K such that the global system $(I_{N_s} \otimes T_s) \star M$ is stable and*

$$\|(I_{N_s} \otimes T_s) \star M\|_\infty < \eta. \quad (2)$$

Remark 2. The proposed decentralized control problem recovers well-known Multi-Agent problems such as consensus, synchronization, coordination, reference tracking, *etc.* [8,9,11,12,31,32]. In this case, the minimization of the H_∞ -norm (2) can be interpreted as the global agent network performance goal: minimization of the synchronisation, tracking

or coordination error [17, 33]. The example of the Multi-Agent synchronization will be illustrated in Section 4; the other problems can be addressed in a similar way.

2.2 Proposed solution to the decentralized control problem

The proposed solution is based on the description of systems using a concept of dissipativity.

Definition 1 (Dissipativity). A causal operator H with the input q and the output p is *strictly* $\{X, Y, Z\}$ -*dissipative*, where $X = X^T$, Y and $Z = Z^T$ are real matrices with $\begin{bmatrix} X & Y \\ Y^T & Z \end{bmatrix}$ full rank if there exists a real $\varepsilon > 0$ such that for any $q \in \mathcal{L}_2$

$$\int_0^\tau \begin{bmatrix} q(t) \\ p(t) \end{bmatrix}^T \begin{bmatrix} X & Y \\ Y^T & Z \end{bmatrix} \begin{bmatrix} q(t) \\ p(t) \end{bmatrix} dt \leq -\varepsilon \int_0^\tau q(t)^T q(t) dt \quad (3)$$

if the inequalities (3) is satisfied with $\varepsilon = 0$ the operator is said to be $\{X, Y, Z\}$ -*dissipative*.

If the operator H is stable and LTI then equation (3) leads to the following frequency domain condition [34]: for almost $\forall \omega \in \mathbb{R}^+$ and a real $\varepsilon > 0$

$$\begin{bmatrix} I \\ H(j\omega) \end{bmatrix}^* \begin{bmatrix} X & Y \\ Y^T & Z \end{bmatrix} \begin{bmatrix} I \\ H(j\omega) \end{bmatrix} \leq -\varepsilon I. \quad (4)$$

The next theorem presents a sufficient condition for (2) based on the dissipativity properties of the interconnection M and the local system T_s .

Theorem 1 (Global Performance). *Given $\eta > 0$, a stable interconnection M and a local plant G and the real matrices $X = X^T \in \mathbb{R}^{m_s \times m_s} \geq 0$, $Y \in \mathbb{R}^{m_s \times n_s}$, $Z = Z^T \in \mathbb{R}^{n_s \times n_s}$, if there exists*

(i) a positive definite matrix $P \in \mathbb{R}^{N_s \times N_s}$ such that M is $\{diag(P \otimes X, -\eta^2 I), diag(P \otimes Y, 0), diag(P \otimes Z, I)\}$ –dissipative;

(ii) a local controller K such that $T_s = G \star K$ is strictly $\{-Z, -Y^T, -X\}$ dissipative

then the local controller K ensures that the large scale system $(I_{N_s} \otimes T_s) \star M$ is stable and

$$\|(I_{N_s} \otimes T_s) \star M\|_\infty \leq \eta. \quad (5)$$

Discussion Condition (i) and condition (ii) of Theorem 1 can be tested using convex optimization involving Linear Matrix Inequality constraints. To this purpose, let us introduce the matrices A, B, C, D which define a minimal state-space representation of $M(s) = C(sI - A)^{-1}B + D$ with $A \in \mathbb{R}^{n \times n}$. The matrix P is such that the condition (i) holds if and only if there exists a positive definite matrix $\hat{P} \in \mathbb{R}^{n \times n}$ such that

$$\begin{bmatrix} I & 0 \\ A & B \\ 0 & I \\ C & D \end{bmatrix}^T \left[\begin{array}{c|ccc} 0 & \hat{P} & & \\ \hat{P} & 0 & & \\ \hline & & P \otimes X & 0 & P \otimes Y & 0 \\ & & 0 & -\eta^2 I & 0 & 0 \\ 0 & & P \otimes Y^T & 0 & P \otimes Z & 0 \\ & & 0 & 0 & 0 & I \end{array} \right] \begin{bmatrix} I & 0 \\ A & B \\ 0 & I \\ C & D \end{bmatrix} \leq 0. \quad (6)$$

which is a Linear Matrix Inequality constraint. This inequality is directly obtained by applying the Kalman-Yakubovich-Popov (KYP) lemma [35]. For the sake of clearness and without loss of generality, in the sequel, the interconnection M is considered static, that is a gain matrix. Thanks to (6), the presented results can be straightforwardly extended to the case of dynamical M .

Furthermore, previously published results reveal that find a controller K such that condition **(ii)** of Theorem 1 is satisfied can be tested using convex optimization involving Linear Matrix Inequality constraints (see *e.g.* [36] or [26]). Moreover, when $X > 0$, using the following loop shifting transformation:

$$\hat{T} = (-X)^{\frac{1}{2}} (T_s + X^{-1}Y) (Z - Y^T X^{-1}Y)^{-\frac{1}{2}} \quad (7)$$

the computation of the controller K such that $T_s = G \star K$ is strictly $\{-Z, -Y^T, -X\}$ -dissipative is equivalent to the computation of K such that the $\|\hat{T}\|_{\infty} < 1$ which is a standard H_{∞} control problem [19]. Such a loop shifting transformation is well-posed since, by applying the Schur Lemma [37], $X > 0$ and (4) imply that $Z - Y^T X^{-1}Y > 0$.

Theorem 1 is now specialized to the stability problem.

Corollary 1 (Global Stability). *Given $\eta > 0$, a stable interconnection M , a plant G , and real matrices $X = X^T \in \mathbb{R}^{m_s \times m_s} \geq 0$, $Y \in \mathbb{R}^{m_s \times n_s}$, $Z = Z^T \in \mathbb{R}^{n_s \times n_s}$, if there exists*

- (i) *a positive definite matrix $P \in \mathbb{R}^{N_s \times N_s}$ such that M_{11} is $\{P \otimes X, P \otimes Y, P \otimes Z\}$ -dissipative;*
- (ii) *a local controller K such that $T_s = G \star K$ is strictly $\{-Z, -Y^T, -X\}$ -dissipative*

then the local controller K ensures that the large scale system $(I_{N_s} \otimes T_s) \star M$ is stable.

Remark 3. For different choices of the matrices X , Y and Z , Corollary 1 recovers some existing results. Note that $\{0, I, 0\}$ -dissipativity corresponds to passivity, while $\{I, 0, I\}$ -dissipativity corresponds to (small) \mathcal{L}_2 -gain. In this sense and when the non-linear counterparts of T_s and M_{11} are considered, Corollary 1 is an generalization of respectively the passivity theorem

and the small gain one. The theorems were applied for ensuring the stability of Multi-Agent systems in *e.g.* [38–40] and [41] respectively. As our approach is not restricted to passivity or \mathcal{L}_2 -gain, the matrices X , Y and Z introduce extra degrees of freedom in order to reduce the conservatism. In the next subsection, a choice of these matrices is investigated.

2.3 Proof of Corollary 1 and Theorem 1

We first prove Corollary 1, which focuses on the stability, in order to afterwards prove Theorem 1 which focuses on the H_∞ norm performance defined by (5).

Proof of Corollary 1 (stability). Note that due to the stability of M , the stability of

$$(I_{N_s} \otimes T_s) \star M = M_{22} + M_{21} (I_{N_s} \otimes T_s) (I - M_{11} (I_{N_s} \otimes T_s))^{-1} M_{12} \quad (8)$$

is equivalent to the stability of $(I_{N_s} \otimes T_s) (I - M_{11} (I_{N_s} \otimes T_s))^{-1}$, that is the feedback connection of $(I_{N_s} \otimes T_s)$ and M_{11} . The proof is based on the application of the separation of graph theorem presented in [27, Theorem 1] (see also [42]) which is presented here in a slightly different way.

Theorem 2. *The negative feedback of a linear $\{-R, -S^T, -Q\}$ -dissipative operator L and a static interconnection H is stable if the following matrix is positive definite, i.e.:*

$$\widehat{Q} = SH + H^T S^T - H^T R H - Q > 0 \quad (9)$$

Condition **(i)** of Corollary 1 states that the interconnection M_{11} is $\{P \otimes X, P \otimes Y, P \otimes Z\}$ -dissipative. First let us prove that if condition **(ii)** of Corollary 1 holds (T_s is strictly $\{-Z, -Y^T, -X\}$ -dissipative) then for any positive definite matrix P , $I_{N_s} \otimes T_s$ is strictly

$\{-P \otimes Z, -P \otimes Y^T, -P \otimes X\}$ -dissipative. Therefore the stability is deduced by applying Theorem 2 due to the fact that condition **(i)** of Corollary 1 implies condition (9) of Theorem 2 with $H = -M_{11}$, $L = I_N \otimes T_s$ and $Q = P \otimes X$, $S = P \otimes Y$, $R = P \otimes Z$.

Condition **(ii)** (the strict $\{-Z, -Y^T, -X\}$ -dissipativity of T_s) ensures that $(I_{N_s} \otimes T_s)$ is strictly $\{-I_{N_s} \otimes Z, -I_{N_s} \otimes Y^T, -I_{N_s} \otimes X\}$ -dissipative, *i.e.* for almost all $\omega \in \mathbb{R}$ and a real $\varepsilon_1 > 0$

$$\begin{aligned} & \begin{bmatrix} I_{n_s} \\ T_s(j\omega) \end{bmatrix}^* \begin{bmatrix} -Z & -Y^T \\ -Y & -X \end{bmatrix} \begin{bmatrix} I_{n_s} \\ T_s(j\omega) \end{bmatrix} \leq -\varepsilon_1 I_{n_s} \\ \Leftrightarrow & \begin{bmatrix} I_{N_s} \otimes T_s(j\omega) \\ I_{N_s} \otimes I_{n_s} \end{bmatrix}^* \begin{bmatrix} I_{N_s} \otimes X & I_{N_s} \otimes Y \\ I_{N_s} \otimes Y^T & I_{N_s} \otimes Z \end{bmatrix} \begin{bmatrix} I_{N_s} \otimes T_s(j\omega) \\ I_{N_s} \otimes I_{n_s} \end{bmatrix} \geq \varepsilon_1 I_{N_s} \otimes I_{n_s} \end{aligned} \quad (10)$$

By post- and pre-multiplying the previous inequality by an invertible full rank matrix $D \otimes I_{n_s}$ with $D \in \mathbb{R}^{N_s \times N_s}$ and its transpose, by noting that

$$(I_{N_s} \otimes T_s(j\omega)) \times (D \otimes I_{n_s}) = (D \otimes I_{m_s}) \times (I_{N_s} \otimes T_s(j\omega))$$

and by using the properties of the Kronecker product [43], one obtains for almost all $\omega \in \mathbb{R}$

$$\begin{bmatrix} I_{N_s} \otimes T_s(j\omega) \\ I_{N_s} \otimes I_{n_s} \end{bmatrix}^* \begin{bmatrix} (D^T D \otimes I_{m_s}) (I_{N_s} \otimes X) & (D^T D \otimes I_{m_s}) (I_{N_s} \otimes Y) \\ (D^T D \otimes I_{n_s}) (I_{N_s} \otimes Y^T) & (D^T D \otimes I_{n_s}) (I_{N_s} \otimes Z) \end{bmatrix} \begin{bmatrix} I_{N_s} \otimes T_s(j\omega) \\ I_{N_s} \otimes I_{n_s} \end{bmatrix} \geq \varepsilon_1 D^T D \otimes I_{n_s}$$

which for $P = D^T D > 0$ is equivalent to for almost all $\omega \in \mathbb{R}$ a real $\varepsilon_1 > 0$

$$\begin{bmatrix} I_{N_s} \otimes T_s(j\omega) \\ I_{n_s \times N_s} \end{bmatrix}^* \begin{bmatrix} P \otimes X & P \otimes Y \\ P \otimes Y^T & P \otimes Z \end{bmatrix} \begin{bmatrix} I_{N_s} \otimes T_s(j\omega) \\ I_{n_s \times N_s} \end{bmatrix} \geq \varepsilon_1 P \otimes I_{n_s} \geq \varepsilon_2 I_{n_s \times N_s} \quad (11)$$

with new real $\varepsilon_2 > 0$ such that $\varepsilon_2 \leq \varepsilon_1 \bar{\sigma}(P \otimes I_{n_s}) = \varepsilon_1 \bar{\sigma}(P)$. The last inequality implies

that $I_{N_s} \otimes T_s$ is strictly $\{-P \otimes Z, -P \otimes Y^T, -P \otimes X\}$ -dissipative which concludes the proof.

Proof of Theorem 1 (performance). Note that condition (i) and condition (ii) of Theorem 1 imply condition (i) and condition (ii) of Corollary 1, which ensures stability.

Let us now prove the inequality (5), that is the inequality on the H_∞ norm. By definition, $\{\text{diag}(P \otimes X, -\eta^2 I), \text{diag}(P \otimes Y, 0), \text{diag}(P \otimes Z, I)\}$ – dissipativity of M is equivalent to the following quadratic condition $\forall w, \forall \tau > 0$:

$$\int_0^\tau \begin{bmatrix} p(t) \\ w(t) \end{bmatrix}^T \begin{bmatrix} I \\ M \end{bmatrix}^T \left[\begin{array}{cc|cc} P \otimes X & 0 & P \otimes Y & 0 \\ 0 & -\eta^2 I & 0 & 0 \\ \hline P \otimes Y^T & 0 & P \otimes Z & 0 \\ 0 & 0 & 0 & I \end{array} \right] \begin{bmatrix} I \\ M \end{bmatrix} \begin{bmatrix} p(t) \\ w(t) \end{bmatrix} dt \leq 0$$

From equation (1) (definition of the global system), one obtains:

$$\begin{aligned} & \int_0^\tau \begin{bmatrix} z(t) \\ w(t) \end{bmatrix}^T \begin{bmatrix} I & 0 \\ 0 & -\eta^2 I \end{bmatrix} \begin{bmatrix} z(t) \\ w(t) \end{bmatrix} dt \\ & \leq - \int_0^\tau \begin{bmatrix} p(t) \\ q(t) \end{bmatrix}^T \begin{bmatrix} P \otimes X & P \otimes Y \\ P \otimes Y^T & P \otimes Z \end{bmatrix} \begin{bmatrix} p(t) \\ q(t) \end{bmatrix} dt \end{aligned} \quad (12)$$

$\forall \tau > 0, \forall w$ and p, q, z defined by (1).

The left hand part of (12) expresses the relation between the input w and the output z of the global system (1) while its right hand part is an integral quadratic constraints on the internal signals q and p , that is, the input/output of $I_{N_s} \otimes T_s$ (see Fig.1). Since condition (11) ensures strict $\{-P \otimes Z, -P \otimes Y^T, -P \otimes X\}$ -dissipativity of $I_{N_s} \otimes T_s$, for some $\varepsilon > 0$ and

$\forall P = P^T > 0$, with $p = I_{N_s} \otimes T_s(q)$, we have:

$$\int_0^\tau \begin{bmatrix} p(t) \\ q(t) \end{bmatrix}^T \begin{bmatrix} -P \otimes X & -P \otimes Y \\ -P \otimes Y^T & -P \otimes Z \end{bmatrix} \begin{bmatrix} p(t) \\ q(t) \end{bmatrix} dt \leq -\varepsilon \int_0^\tau q(t)^T q(t) dt \leq 0 \quad (13)$$

From condition (13) and condition (12), one obtains

$$\int_0^\tau \begin{bmatrix} z(t) \\ w(t) \end{bmatrix}^T \begin{bmatrix} I & 0 \\ 0 & -\eta^2 I \end{bmatrix} \begin{bmatrix} z(t) \\ w(t) \end{bmatrix} dt \leq 0$$

$\forall \tau > 0$, $\forall w$ and z defined by (1).

Applying Fourier transform, the last equation becomes

$$\int_{-\infty}^{+\infty} \begin{bmatrix} z(j\omega) \\ w(j\omega) \end{bmatrix}^* \begin{bmatrix} I & 0 \\ 0 & -\eta^2 I \end{bmatrix} \begin{bmatrix} z(j\omega) \\ w(j\omega) \end{bmatrix} d\omega \leq 0$$

Leaving out the integral and noting that $z = (I_{N_s} \otimes T_s) \star M(w)$ the following condition on external input output signals in frequency domain is obtained: for almost $\forall \omega \in \mathbb{R}^+$

$$\begin{bmatrix} (I_{N_s} \otimes T_s) \star M(j\omega) \\ I(t) \end{bmatrix}^* \begin{bmatrix} I & 0 \\ 0 & -\eta^2 I \end{bmatrix} \begin{bmatrix} (I_{N_s} \otimes T_s) \star M(j\omega) \\ I(t) \end{bmatrix} \leq 0$$

which is equivalent to $\|(I_{N_s} \otimes T_s) \star M\|_\infty \leq \eta$, *i.e.* the inequality (5).

2.4 How to choose X , Y , Z ?

In order to apply Theorem 1 and Corollary 1, it is necessary to choose the matrices X , Y and Z . For the sake of clarity, we focus on static interconnections, that is M is a real matrix, with $n_s = 1$, $m_s = 1$, *i.e.* X , Y , Z are scalars, referred to as in the sequel x , y and

z , and T_s SISO. The case of a dynamical LTI interconnections M can be investigated in a similar way. However, the choice of the matrices in the Multi Input Multi Output (MIMO) case is the perspective of a future work.

The main idea of the choice of x , y and z is to satisfy the condition (i) of Corollary 1 and Theorem 1 and to relax condition (ii) as much as possible. Indeed, it allows to obtain the least restrictive constraint $\|\widehat{T}\|_\infty < 1$ (see (7)) in order to compute the local controller by solving a standard H_∞ problem. We first discuss the choice of x , y and z for the application of Corollary 1 and then for the application of Theorem 1.

Conditions of Corollary 1 (Stability)

Condition (ii) of Corollary 1 enforces the transfer function T_s to be strictly $\{-z, -y, -x\}$ -dissipative, that is:

$$\begin{bmatrix} T_s(j\omega) \\ 1 \end{bmatrix}^* \begin{bmatrix} x & y \\ y & z \end{bmatrix} \begin{bmatrix} T_s(j\omega) \\ 1 \end{bmatrix} > 0 \quad (14)$$

$$\iff (T_s(j\omega) - c)^*(T_s(j\omega) - c) < r^2 \quad (15)$$

with $c = -\frac{y}{x}$ and $r = \sqrt{\frac{y^2}{x^2} - \frac{z}{x}}$. The last inequality means that the Nyquist plot of the transfer function T_s is inside the circle with center c and radius r . Fig.2 presents the Nyquist plot of T_s in the case of a typical PLL design (blue line), with the circle plotted in full red line. To relax the constraint (15) for a given center c , the radius r has to be maximized. Note that the radius r can be increased by increasing the positive $-\frac{z}{x}$. Similarly, condition (i) of Corollary 1 which enforces the matrix M_{11} to be $\{xP, yP, zP\}$ -dissipative with

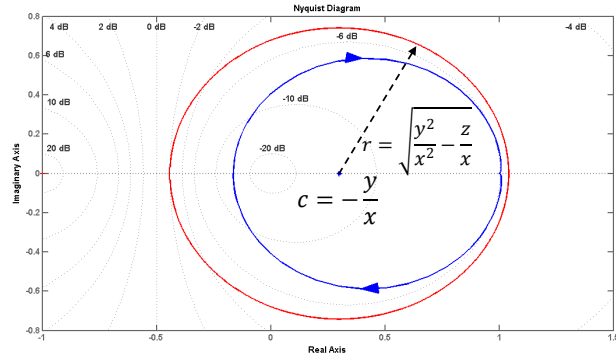


Figure 2: Typical Nyquist plot of a transfer function T_s

$P > 0$, can be expressed as:

$$\left(M_{11} + \frac{y}{z}I\right)^T P \left(M_{11} + \frac{y}{z}I\right) \leq \left(\frac{y^2}{x^2} - \frac{z}{x}\right) P \quad (16)$$

Increasing $-\frac{z}{x}$ strengthens the constraint (16). Thus constraint (15) and constraint (16) cannot be simultaneously relaxed. The following optimization problem should therefore be solved: “find x, y, z such that the condition (16) is satisfied and that it maximizes the circle radius r for a given center c ”.

To solve this problem, without loss of generality, one redundant variable can be eliminated by setting for example $z = 1$ ¹. By introducing in condition (16) the variables r and c , the

¹This is equivalent to dividing both inequalities (14) and (16) by $z > 0$ and performing the change of variables: $\tilde{x} = \frac{x}{z} \leq 0$, $\tilde{y} = \frac{y}{z}$.

optimization problem can be formulated as:

$$\min_{\chi, P} \quad \chi = \frac{1}{r^2} \tag{17}$$

$$\begin{aligned} \text{such that} \quad P &> 0 \\ M_{11}^T P M_{11} &\leq \underbrace{\frac{1}{r^2}}_{\chi} (cM_{11} - I)^T P (cM_{11} - I) \end{aligned}$$

For a given c , the minimization problem of χ (and thus the maximization of r) with the decision variables $P \in \mathbb{R}^{N_s \times N_s}$ and $\chi \in \mathbb{R}^+$, is a generalized eigenvalue problem involving Linear Matrix Inequality constraints, that is a standard (quasi convex) LMI optimization problem [37]. If c is a decision variable, the optimization problem (17) is no longer a generalized eigenvalue minimization problem. However, based on the spectrum of the matrix M_{11} , c can be chosen in order to ensure the feasibility of condition (16): with $\lambda_i^{M_{11}}$ is the i -th eigenvalue of M_{11}

$$|c| = \left| \frac{z}{x} \right| \leq \frac{1}{2 \max_i (|\operatorname{Re}(\lambda_i^{M_{11}})|)}. \tag{18}$$

Then, using a linear search on c in the interval defined by (18), the best c and r can be found in the sense of relaxing the T_s transfer function circle constraint (15) and thus the condition (ii) of Corollary 1. Finally, $\{x, y, z = 1\}$ is obtained from $\{r, c\}$ by a simple transformation. Let us now generalize this result to the global performance case.

2.5 Theorem 1 conditions (Performance)

As in the previous subsection, condition (i) of Theorem 1 is equivalently expressed in terms of the center c and the radius r : there exists $P = P^T > 0$ such that

$$\begin{bmatrix} I \\ M \end{bmatrix}^T \left[\begin{array}{cc|cc} -P & 0 & cP & 0 \\ 0 & -\eta^2 I & 0 & 0 \\ \hline cP & 0 & (r^2 - c^2)P & 0 \\ 0 & 0 & 0 & I \end{array} \right] \begin{bmatrix} I \\ M \end{bmatrix} \leq 0 \quad (19)$$

Condition (19) can be factorized as:

$$\begin{bmatrix} I \\ M \end{bmatrix}^T \left[\begin{array}{cc|cc} I & 0 & -cI & 0 \\ 0 & I & 0 & 0 \\ \hline 0 & 0 & I & 0 \\ 0 & 0 & 0 & I \end{array} \right]^T \left[\begin{array}{cc|cc} -P & 0 & 0 & 0 \\ 0 & -\eta^2 I & 0 & 0 \\ \hline 0 & 0 & r^2 P & 0 \\ 0 & 0 & 0 & I \end{array} \right] \left[\begin{array}{cc|cc} I & 0 & -cI & 0 \\ 0 & I & 0 & 0 \\ \hline 0 & 0 & I & 0 \\ 0 & 0 & 0 & I \end{array} \right] \begin{bmatrix} I \\ M \end{bmatrix} \leq 0$$

With the change of variables $\hat{P} = r^2 P$ and $\beta = \eta^2 r^2$, the following optimization problem is obtained:

$$\min_{\chi, \hat{P}} \quad \chi = \frac{1}{r^2} \quad (20)$$

$$\text{such that} \quad \begin{array}{c} \hat{P} \\ > \\ 0 \end{array} \quad \Phi^T \left[\begin{array}{cc|cc} 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ \hline 0 & 0 & \hat{P} & 0 \\ 0 & 0 & 0 & I \end{array} \right] \Phi \leq \underbrace{\frac{1}{r^2}}_{\chi} \Phi^T \left[\begin{array}{cc|cc} \hat{P} & 0 & 0 & 0 \\ 0 & \beta I & 0 & 0 \\ \hline 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right] \Phi$$

$$\text{where } \Phi = \left[\begin{array}{cc|cc} I & 0 & -cI & 0 \\ 0 & I & 0 & 0 \\ \hline 0 & 0 & I & 0 \\ 0 & 0 & 0 & I \end{array} \right] \begin{bmatrix} I \\ \hline M \end{bmatrix}.$$

The optimization problem (20) with a given center c and ratio β is a generalized eigenvalue problem involving Linear Matrix Inequality constraints.

Remark 4. The parameter $\beta = \eta^2/\chi$ is a tuning parameter for the condition (20). For a given β , the minimization of χ using (20) maximizes the radius r of the circle and thus relaxes the constraint (15). Furthermore, the upper bound η on the H_∞ -norm of the global performance transfer function *i.e.* $\|(I_{N_s} \otimes T_s) \star M\|_\infty$ is minimized. If the optimization problem (20) has no solution or the computed radius $r = 1/\sqrt{\chi}$ is too tight constraint on

(15), β should be increased to relax the condition of the optimization problem (20). The initial value of $\beta = r^2\eta^2$ can be computed for the control law K which only ensures the global stability, *i.e.* the conditions of Corollary 1 for x, y, z found by solving the optimization problem (17).

Remark 5. The result of the optimization problem (20) is both 1) a property of $\{x, y, z\}$ -dissipativity to be ensured by the controller K *i.e.* condition (ii) of Theorem 1 and 2) an upper bound on the global transfer function H_∞ -norm *i.e.* the performance objective. This upper bound is actually found before the computation of the control law. It *a priori* ensures a bound on the achieved global performance level if the controller ensures the local dissipativity constraint *i.e.* condition (ii) of Theorem 1.

3 Local control for global performance

When applying Theorem 1 or Corollary 1, if condition (i) is satisfied and in order to solve Problem 1, it is necessary to find K such that $T_s = G \star K$ is strictly $\{-Z, -Y^T, -X\}$ -dissipative.

As explained in the discussion page 13, this problem reduces to compute K such that $\|\hat{T}\|_\infty < 1$ where \hat{T} is defined by (7), which is a standard H_∞ problem.

As discussed in the previous sections, in the Microelectronics design of PLL networks, as T_s represents a PLL, the design of K such that $T_s = G \star K$ satisfies some specifications, corresponds to the local design step. A strong interest of Theorem 1 or Corollary 1 is to reveal how to design the controller K of the local system $T_s = G \star K$ in order to *a priori* ensure the stability of the global system $(I_{N_s} \otimes T_s) \star M$. Theorem 1 also guarantees the

global system performance defined by (2) and thus present a solution to Problem 1. The question is to investigate how to apply this result in order to directly enforce typical design specifications on the global system at the local design step.

A popular approach to specify the performance of a (closed-loop) system is to define frequency dependent upper bound on the magnitude of the frequency response of several (closed-loop) Single Input, Single Output (SISO) transfer functions, see e.g. [19,44,45]. The computation of the controller such that the performance is satisfied can be expressed as an optimization problem involving an inequality on the weighted H_∞ norm of the closed loop system. Such a performance specification is particularly interesting in Microelectronics [46–49]. As a consequence, the computation of the filter K of a PLL can be expressed as a standard H_∞ control problem: compute K such that

$$\|W_{lpo}G_{lp} \star KW_{lpi}\|_\infty < 1 \quad (21)$$

where $G_{lp} \star K$ is a matrix of transfer functions which includes the closed-loop transfer functions associated to the performance and where $W_{lpo} = \mathbf{diag}_i(w_{lpo}^i)$ and $W_{lpi} = \mathbf{diag}_j(w_{lpi}^j)$ are diagonal matrices of weighting functions. The inequality (21) ensures that the magnitude of the frequency response of the transfer function between the input j and the output i is below $\frac{1}{|w_{lpo}^i(j\omega)||w_{lpi}^j(j\omega)|}$ for all frequency ω .

In the following theorem, we reveal how to enforce, with a good choice of G_{lp} , a given frequency dependent upper bound on the magnitude of the frequency response of the global system transfer function *i.e.* on $|(I_{N_s} \otimes T_s(j\omega)) \star M|$, by computing a controller K such that (21). To do so, a new global weighting function $W = (I_{N_w} \otimes T_s) \star M_W$ will be used.

Please notice that this transfer function is chosen as the global system (1) to be in the form of interconnection of N_w subsystems T_s and static matrix M_W . The reason of such choice will be discussed after theorem. For the sake of clearness, the result is presented in the case of one SISO global transfer function, *i.e.* $n_z = n_w = 1$: nevertheless, due to the MIMO H_∞ norm constraint properties [19], the result can be applied in the case of several ones *i.e.* for MIMO global transfer functions.

Theorem 3. *Let be $\eta > 0$, a stable interconnection M , 3 real matrices $X = X^T \in \mathbb{R}^{m_s \times m_s} \geq 0$, $Y \in \mathbb{R}^{m_s \times n_s}$, $Z = Z^T \in \mathbb{R}^{n_s \times n_s}$, G and G_{lp} which define the local subsystem T_s and a SISO local performance (*lp*) transfer function T_{lp} such that:*

- $G \star K = T_s$
- $G_{lp} \star K = T_{lp}$

and a SISO stable interconnected system W defined by $(I_{N_w} \otimes T_s) \star M_W$ such that

$$|W(j\omega)|^{-1} \approx |T_{lp}(j\omega)|.$$

If there exists

- (i) a positive definite matrix $P \in \mathbb{R}^{(N_s+N_w) \times (N_s+N_w)}$ such that \widetilde{M} is $\{\text{diag}(P \otimes X, -\eta^2 I), \text{diag}(P \otimes Y, 0), \text{diag}(P \otimes Z, I)\}$ –dissipative where \widetilde{M} is defined by:

$$(I_{N_s+N_w} \otimes T_s) \star \widetilde{M} = (I_{N_w} \otimes T_s) \star M_W \times (I_{N_s} \otimes T_s) \star M \quad (22)$$

- (ii) a controller K such that

$$\left\| (-X)^{\frac{1}{2}} (G + X^{-1}Y) \star K (Z - Y^T X^{-1}Y)^{-\frac{1}{2}} \right\|_{\infty} < 1 \quad (23)$$

$$\|W_{lpo} G_{lp} \star K W_{lpi}\|_{\infty} < 1$$

then the controller K ensures that the global system $(I_{N_s} \otimes T_s) \star M$ is stable and

$$|((I_{N_s} \otimes T_s) \star M)(j\omega)|_{dB} \leq -|W_{lpo}(j\omega)W_{lpi}(j\omega)|_{dB} + 20 \log_{10}(\eta), \quad \forall \omega \quad (24)$$

Proof. This theorem is proved by applying Theorem 1 to the (augmented) interconnected system $(I_{N_s+N_w} \otimes T_s) \star \widetilde{M}$ where \widetilde{M} defined by (22).

Let us assume that both conditions of this theorem are satisfied. Condition **(i)** is exactly condition **(i)** of Theorem 1. The first condition in (23) implies that there exist a controller K such that

$$\|\hat{T}\|_{\infty} < 1 \quad (25)$$

where \hat{T} is defined by (7). From the discussion on page 13, this condition is equivalent to condition **(ii)** of Theorem 1. Then, by applying Theorem 1, the augmented system $(I_{N_s+N_w} \otimes T_s) \star \widetilde{M}$ is stable and

$$\left\| (I_{N_s+N_w} \otimes T_s) \star \widetilde{M} \right\|_{\infty} \leq \eta.$$

As W and $(I_{N_s} \otimes T_s) \star M$ are SISO, the last inequality is equivalent to:

$$|((I_{N_s} \otimes T_s) \star M)(j\omega)| \leq \frac{\eta}{|W(j\omega)|}, \quad \forall \omega.$$

On the other hand, the second condition in (23) implies that (see the definition of G_{lp}):

$$|T_{lp}(j\omega)| < \frac{1}{|W_{lpo}(j\omega)W_{lpi}(j\omega)|} \quad (26)$$

Since $|W(j\omega)|^{-1} \approx |T_{lp}(j\omega)|$, we obtain:

$$|((I_{N_s} \otimes T_s) \star M)(j\omega)| \leq \frac{\eta}{|W_{lpo}(j\omega)W_{lpi}(j\omega)|}, \quad \forall \omega.$$

Inequality (24) is obtained by expressing the previous equation in dB, which concludes the proof. ■

Remark 6. The stable interconnected system W can be interpreted as the global weighting function imposing a global performance frequency constraint as in the standard H_∞ approach:

$$\|W((I_{N_s} \otimes T_s) \star M)\|_\infty \leq \eta$$

Traditionally, in the H_∞ framework, the system to be constrained is represented as an *interconnection of integrators i.e.* a state-space representation. It is also the case for the weighting functions as well and then for the augmented system. The existing efficient algorithms of the standard H_∞ problem resolution (see *e.g.* [19, 37]) are based on this important fact.

Nevertheless, in our problem, the global system $(I_{N_s} \otimes T_s) \star M$ is naturally represented as an interconnection of the *subsystems dynamics* T_s (see (1) and Fig. 1). For this reason, the weighting function W , similarly to the standard H_∞ framework, has to be represented as the interconnection of N_w subsystems T_s .

Another requirement on the weighting function W is that its magnitude should be equal to the magnitude of the inverse of a local performance transfer function, at least for some frequency ranges of interest. This implies that the local performance transfer function T_{lp} should be chosen as well as an interconnection of T_s . This original idea allows to introduce

a *natural* and *strong connection* between the local T_{lp} and global $(I_N \otimes T_s) \star M$ performance transfer functions so that their difference can be easily expressed and minimized.

Remark 7. The proof of Theorem 3 reveals that the inequality (24) is deduced from the inequality:

$$|((I_N \otimes T_s) \star M)(j\omega)|_{dB} \leq |T_{lp}(j\omega)|_{dB} + 20 \log_{10}(\eta), \quad \forall \omega \quad (27)$$

The inequality (27) presents the relation between the shape of the frequency response of the global system $(I_N \otimes T_s) \star M$, that is the *global performance*, and the shape of the frequency response of the stand-alone subsystem T_{lp} that is the *local performance*.

Discussion In Theorem 3, the first inequality of (23) ensures the strict $\{-Z, -Y^T, -X\}$ dissipativity of T_s which is necessary to the application of Theorem 1. The second inequality ensures the local subsystem performance *i.e.* the performance of an independent stand-alone subsystem T_{lp} which is expressed as the upper frequency bound (26). Therefore, since the stable interconnected system W is defined as the inverse of T_{lp} expressed as an interconnection of T_s , its series combination with the global system (22) defines an augmented system describing a relation between the global and local performance transfer functions. Ensuring a bound η on this relation allows to ensure the global system performance (24) by enforcing the local subsystem performance (26).

Theorem 3 allows to compute the local controller K that ensures the global system performance expressed in the form of a frequency upper bound (24). This computation is performed in some how relative fashion *i.e.* the global system performance is rather ensured in the relative then the absolute way (see Remark 7). We now discuss how it is possible

to transform the global system performance problem into a standard local H_∞ controller design problem that implies (23).

Assume that the condition (i) of the Theorem 3 is satisfied. The computation of the controller such that both inequalities in (23) is actually a multi-objective H_∞ optimization problem. It is usually transformed into a standard one objective optimization problem by defining an augmented plant \tilde{G} such that, for example:

$$\begin{aligned} \bullet \quad & \begin{bmatrix} I & 0 \end{bmatrix} \tilde{G} \star K \begin{bmatrix} I \\ 0 \end{bmatrix} = G \star K = T_s \\ \bullet \quad & \begin{bmatrix} 0 & I \end{bmatrix} \tilde{G} \star K \begin{bmatrix} 0 \\ I \end{bmatrix} = G_{lp} \star K = T_{lp} \end{aligned}$$

The standard H_∞ problem can now be defined: find the controller K such that:

$$\left\| \mathbf{diag} \left((-X)^{\frac{1}{2}}, W_{lpo} \right) \left(\left(\tilde{G} + \mathbf{diag}(X^{-1}Y, 0, 0) \right) \star K \right) \mathbf{diag} \left((Z - Y^T X^{-1} Y)^{-\frac{1}{2}}, W_{lpi} \right) \right\|_\infty < 1. \quad (28)$$

Due to the H_∞ norm properties, the controller K such that the inequality (28) is satisfied is such that the inequalities (23) are satisfied. Depending on the problem under consideration, it could be more interesting to replace the H_∞ problem defined by (28) by another H_∞ problem :

$$\left\| W^o \hat{G} \star K W^i \right\|_\infty < 1 \quad (29)$$

where the choice of the plant \hat{G} and the weighting functions W^o and W^i is different from the choice presented in (28) but where the inequality (29) implies the inequalities (23). The motivation of this idea is illustrated on the numerical example presented in the next section.

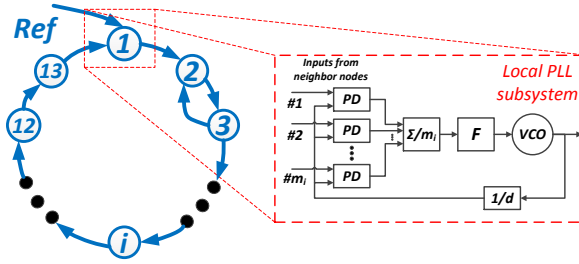


Figure 3: The active clock distribution network

4 PLL Network design

In this section, the method presented in Section 3 is applied to the design of an active clock distribution network, that is, a network of N_s Phase-Locked Loops (PLL), (see Fig. 3). A Phase-Locked Loop is a feedback system that generates a periodic output signal which has to be synchronized (in frequency and/or in phase) with an external periodic signal. Each PLL of the network is composed of a Voltage Controlled Oscillator (VCO) generating a local periodic signal which is fed back through a frequency divider ($1/d$) to the Phase Detectors (PD). The purpose of each PD is to measure the phase difference between the VCO signal and the m_i external input signals, up to a scaling factor, see Fig. 3. From the average of the m_i differences (block Σ/m_i), the Filter (F) computes the command signal which is the input of the VCO and should modify the frequency of its output. In a network of PLLs, the PD s between adjacent PLLs are usually factorized.

The Microelectronics design problem is to design an active clock distribution network (an example is presented Fig.3) that achieves the frequency and phase synchronization of the N_s VCO outputs to an external reference signal, denoted Ref , within a specified time and

with a suitable damping. Since the main parameters of a VCO and a PD as well as the network topology are deduced from the technology process, the chip structure and the clock generation requirements (clock frequency, number of PLLs with their localization, VCO and PD resolution, *etc.*), the design problem can be expressed as the design of the filter F such that the active clock network achieves the specifications.

In the example detailed below, we focus on a network of $N_s = 13$ PLLs with a circular topology and one external reference signal connected to the PLL 1, see Fig.3. Such an example is challenging since the traditional design procedure fails to achieve the design of a satisfactory network, see [50]. This point is emphasized in the next subsection with the application of the traditional design procedure. The application of our approach is investigated in the subsection 4.2 and the subsection 4.3.

4.1 Traditional design process

The traditional design process consists in: *(i)* design a filter F such that the stand-alone PLL has a good behavior and *(ii)* evaluation of the behavior of the active clock distribution network with this PLL. The design of a stand-alone PLL filter is based on a model where the periodic signals are represented by their phase [47, 49, 51]. In this domain, the synchronization of the VCO signal with a periodic (sinusoidal) external signal is expressed as tracking the phase of the VCO signal given by a ramp whose slope is the reference oscillator frequency. In addition, the phase domain PLL model, which is non-linear, can be

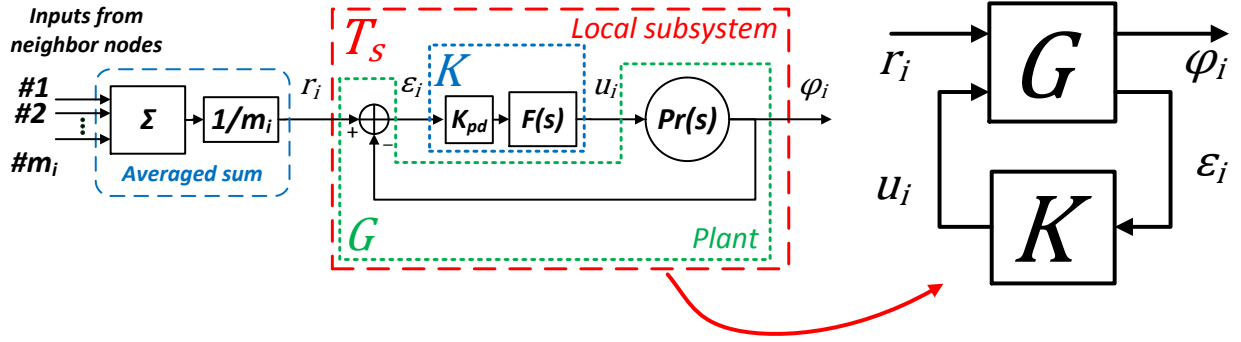


Figure 4: Phase domain model of the PLL subsystem

linearized around an operating point². Based on this assumption and the PLL representation illustrated in Fig.3, the (linearized phase domain) model of one PLL is presented in Fig.4 (left part), where K_{pd} denotes the phase detector linear gain and $Pr(s)$ the VCO transfer function defined with K_{VCO} the gain of the VCO by

$$Pr(s) = \frac{K_{VCO}/d}{s}.$$

The numerical values are presented in Tab. 1 where the VCO central frequencies represent constant perturbations at the VCO inputs that have to be rejected to ensure the synchronization.

Using the traditional frequency domain design methods, the following filter F is obtained with a satisfying behavior (PLL synchronization with periodical external signal at its input, response time less than $6 \mu sec$, infinite gain margin, phase margin of 79 degrees, *etc.*):

$$F(s) = 0.44 \frac{s + 464.2}{s}.$$

²One major non-linear issue of such PLL network known as mode-locking states will not be addressed in this paper but can be circumvented independently as in [2, 3, 5].

Table 1: Numerical values of the considered clock distribution network

Parameter	Numerical value
Reference frequency	$f_{ref} = 50 \text{ kHz}$
Frequency divider factor	$d = 4$
VCO central frequencies	random initialized around $d \cdot f_{ref} \pm 25\%$
VCO gain	$K_{VCO} = 160.12 \text{ Hz/c.u.}$
PD gain	$K_{PD} = 21.22 \text{ e.u./rad}$
Number of inputs	$m_i \in \{1, 2\}$ depending on PLL position
Dimensions	$N_s = 13, n_s = 1, m_s = 1$

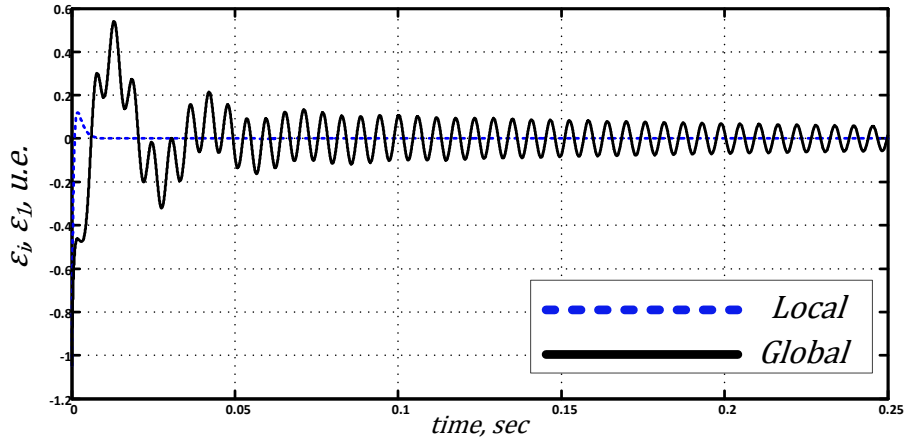


Figure 5: Synchronization errors of a stand-alone PLL (dashed blue line) and of PLL 1 of the global network (full black line)

Unfortunately, the obtained active clock distribution network does not behave as well as one stand-alone PLL, see Fig.5. In this figure, the time-domain simulation reveals that, in contrast with the synchronization error ε_i of the stand-alone PLL (blue line), the synchronization error ε_1 of the PLL 1 of the network is not well-damped (considerable overshoot and parasite oscillations) and slowly decays to zero (around 100 times). This different behavior can be observed in the frequency domain. Let $S_{loc} = T_{r_i \rightarrow \varepsilon_i}$ be the transfer function between the reference input r_i and the synchronization error $\varepsilon_i = r_i - \varphi_i$ of the stand-alone PLL, see Fig. 4:

$$S_{loc}(s) = \frac{1}{1 + F(s)K_{PD}Pr(s)}.$$

Let $S_{glob} = T_{Ref \rightarrow \varepsilon_1}$ be the transfer function between the reference input Ref of the network and the PLL 1 synchronization error defined as the difference between the reference signal and the PLL 1 VCO output: $\varepsilon_1 = Ref - \varphi_1$. Fig.6 illustrates the magnitude of the frequency response of S_{loc} (dashed blue line) and of S_{glob} (full blue line). In both cases,

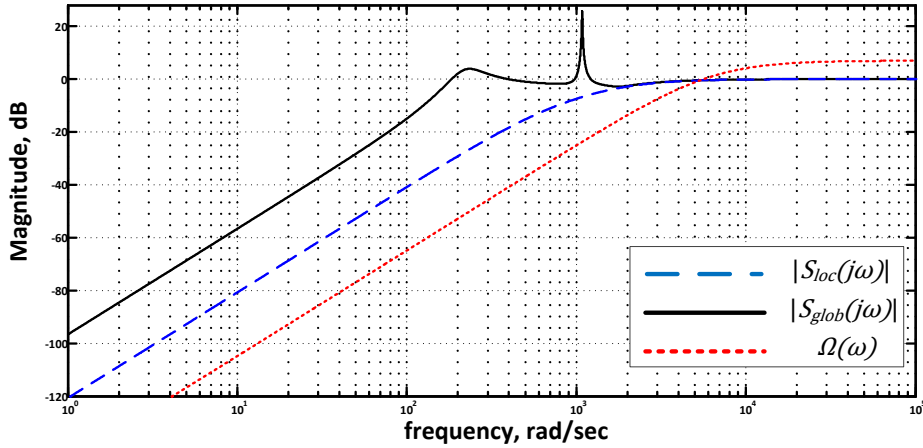


Figure 6: Magnitude of the transfer functions S_{loc} (dashed blue line) and S_{glob} (full black line), as well as the considered frequency constraint (red dotted line) ensuring the global performance

the synchronization (or ramp tracking) is ensured by the slope of +40 dB/dec in the low frequency (LF) range. The time response is reduced by increasing the cut-off frequency and decreasing the transfer function magnitude in the LF range. The overshoot is limited by reducing the maximal transfer function gain (see [19, 52] for more details). Therefore, in order to improve the behavior of the PLL network (reduced time response and well-dampedness), the magnitude of the frequency response of the transfer function S_{glob} has to respect the frequency constraint represented by the red dotted line in Fig.6. This frequency constraint enforces a more important cut-off frequency and magnitude attenuation in LF range while maintaining the peak low in the middle frequency range. It can be represented as the magnitude of the frequency response $\Omega(\omega)$ of a transfer function as defined below:

$$\Omega(s) = |W_{gl}(s)|^{-1} / (W_{gl}(s))^{-1} = \frac{2.23 (s + 0.56)^2}{(s + 6269)^2}. \quad (30)$$

In the next subsection, we reveal how to design F such that the magnitude of the frequency response of the transfer function S_{glob} satisfies this frequency constraint.

4.2 Proposed design process

The purpose is to compute F in order to enforce the synchronization of PLL 1 of the network with a specified behavior *i.e.* the frequency response of the global transfer function S_{glob} respects the frequency constraint enforced by (30). To this purpose, the approach presented section 3 (Theorem 3) is iteratively applied by modifying X , Y , Z and the weighting functions W_{lpi} and W_{lpo} until the global specifications are satisfied. The synchronization of all N_s PLLs or more complex synchronization specifications [50] can be explicitly investigated by this approach. Nevertheless, for the sake of clarity, these cases are not presented in this example.

Notations First, let us redefine the problem in Theorem 3 notations. In the example under consideration, the transfer function $S_{glob} = T_{Ref \rightarrow \varepsilon_1}(s)$ can be represented as an

interconnection M of N_s subsystems $T_s = G \star K, (I_N \otimes T_s(s)) \star M$, with

$$M = \left[\begin{array}{cccccc|ccc} 0 & 0 & 0 & 0 & \cdots & 0 & \frac{1}{2} & \frac{1}{2} \\ \frac{1}{2} & 0 & \frac{1}{2} & 0 & \cdots & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & \cdots & 0 & 0 & 0 \\ \vdots & \ddots & \ddots & \ddots & \ddots & \vdots & \vdots & \vdots \\ \vdots & \ddots & \ddots & \ddots & \ddots & \vdots & \vdots & \vdots \\ 0 & \ddots & \ddots & \ddots & \ddots & 0 & 0 & 0 \\ 0 & \cdots & \cdots & \cdots & 0 & 1 & 0 & 0 \\ \hline -1 & 0 & \cdots & \cdots & \cdots & \cdots & 0 & 1 \end{array} \right]$$

$$G(s) = \begin{bmatrix} 0 & Pr(s) \\ 1 & -Pr(s) \end{bmatrix} \quad \text{and} \quad K(s) = K_{pd}F(s).$$

The local performance transfer function T_{lp} is defined as:

$$T_{lp}(s) = S_{loc}(s) = T_{r_i \rightarrow \varepsilon_i}(s) = \frac{1}{1 + K(s)Pr(s)}.$$

Thus, $T_{lp} = G_{lp} \star K$ with

$$G_{lp}(s) = \begin{bmatrix} 1 & -Pr(s) \\ 1 & -Pr(s) \end{bmatrix}.$$

Note that in our problem, T_s is defined as $T_s(s) = \frac{K(s)Pr(s)}{1 + K(s)Pr(s)}$. T_{lp} can be represented as an interconnection of T_s , since

$$T_{lp} = 1 - T_s. \quad (31)$$

In order to ensure that $|W(j\omega)|^{-1} \approx |T_{lp}(j\omega)|$ for the frequency range of interest, where W

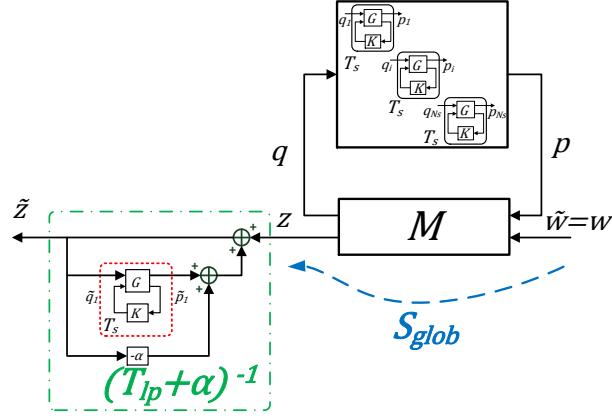


Figure 7: Augmented system $(I_{N_s+N_w} \otimes T_s) \star \widetilde{M}$

can be represented as an interconnection of subsystems T_s , we choose:

$$W = \frac{1}{T_{lp} + \alpha} = \frac{1}{1 + \alpha - T_s}$$

where $\alpha = 0.01$ is such that W is well-posed³, stable and $|W(j\omega)|^{-1} \approx |T_{lp}(j\omega)|$ in the frequency range of interest. Then, $W = T_s \star M_W$ with

$$M_W = \begin{bmatrix} (1 + \alpha)^{-1} & (1 + \alpha)^{-1} \\ (1 + \alpha)^{-1} & (1 + \alpha)^{-1} \end{bmatrix} = \begin{bmatrix} M_{11}^W & M_{12}^W \\ M_{21}^W & M_{22}^W \end{bmatrix}.$$

The system S_{glob} augmented with the weighting function W defined by $(I_{N+N_w} \otimes T_s) \star \widetilde{M} = (I_{N_w} \otimes T_s) \star M_W \times (I_N \otimes T_s) \star M$ is represented in Fig. 7 while \widetilde{M} is defined by:

$$\widetilde{M} = \begin{bmatrix} M_{11}^W & M_{12}^W M_{21} & M_{12}^W M_{22} \\ 0 & M_{11} & M_{12} \\ M_{21}^W & M_{22}^W M_{21} & M_{22}^W M_{22} \end{bmatrix}.$$

³For more details, see [53].

Local control for global performance In order to ensure the global performance, by applying Theorem 3, the conditions (i) and (ii) should be ensured for some X, Y, Z and W_{lp}^i, W_{lp}^o chosen such that the frequency constraint defined by (30) and:

$$-|W_{lpi}(j\omega)W_{lpo}(j\omega)|_{dB} + 20\log_{10}(\eta) \leq -20\log(\Omega(\omega)) \quad (32)$$

Condition (i) of Theorem 3 is satisfied by an appropriate choice of X, Y, Z which are scalars in the considered example. One possible choice is described in Subsection 2.4.

Condition (ii) (i.e. (23)) can be satisfied by solving the standard H_∞ problem (28) with

$$\tilde{G}(s) = \begin{bmatrix} 0 & 0 & Pr(s) \\ 1 & 1 & -Pr(s) \\ 1 & 1 & -Pr(s) \end{bmatrix} : \tilde{G}(s) \star K = \begin{bmatrix} T_s(s) & T_s(s) \\ T_{lp}(s) & T_{lp}(s) \end{bmatrix}.$$

Nevertheless, in order to ensure additional design specification on the PLL, the H_∞ problem (28) is replaced by another H_∞ problem (29).

Additional local performance specifications In addition to synchronization, the effects of the input/output PD/VCO noises and possible perturbations have to be attenuated/rejected at the local level. The temperature and power disturbances as well as the central frequency variations of the VCO can be modeled as a constant or a slowly varying perturbation on the VCO input. Another specification is the limitation of the controller bandwidth (reduced control energy and noise attenuation). Such specifications can be enforced for a stand-alone PLL by ensuring upper bounds on the magnitude of the frequency response of the following transfer functions:

- For synchronization, PLL damping, *VCO* output noise rejection: $T_{r_i \rightarrow \varepsilon_i}(s) = T_{lp}(s)$
- For *VCO* input noise and constant perturbation rejection: $Pr(s)T_{lp}(s)$
- For controller bandwidth limitation: $K(s)T_{lp}(s)$
- For *PD* and reference signal noise rejection: $Pr(s)K(s)T_{lp}(s) = T_s(s)$

Note that the first local performance specification is related to the transfer function T_{lp} , *i.e.* the same transfer function introduced to enforce the global performance. In order to take into account these local performance requirements, the H_∞ problem (29) is solved with

$$\widehat{G} = \begin{bmatrix} 1 & -Pr(s) & -Pr(s) \\ 0 & 0 & 1 \\ X^{-1}Y & Pr(s) & Pr(s) \\ 1 & -Pr(s) & -Pr(s) \end{bmatrix} \quad \begin{aligned} W^o &= \mathbf{diag}(W_1^o, W_2^o, (W_1^i)^{-1}, b) \\ W^i &= \mathbf{diag}(W_1^i, W_2^i) \end{aligned}$$

where $b = (-X)^{-\frac{1}{2}}(Z - Y^T X^{-1}Y)^{-\frac{1}{2}}$. Due to the properties of the H_∞ norm, the solution of problem (29) directly implies the following conditions (see the expression of \widehat{G}).

$$\|(T_s + X^{-1}Y)b\|_\infty < 1$$

that is, the first condition (23) of Theorem 3, and

$$\begin{aligned} \|W_1^o Pr T_{lp} W_2^i\|_\infty < 1 & \quad \|W_2^o K T_{lp} W_1^i\|_\infty < 1 \\ \|W_2^o Pr K T_{lp} W_2^i\|_\infty < 1 & \quad \|W_1^o T_{lp} W_1^i\|_\infty < 1 \end{aligned} \quad (33)$$

The weighting functions $W_{1(2)}^{i(o)}$ are chosen such that they imply the local performance constraint *i.e.* the second constraint in (23) of Theorem 3 and the additional local performance specifications.

It is true that the additional local performance is ensured for a stand-alone PLL and is not necessarily ensured for the PLL in the network (globally). Nevertheless, for some performance specification, global performance specification can be related to the local one. Let us focus on

- Reference noise rejection on the first PLL output: $T_{Ref \rightarrow \varphi_1}(s) = Pr(s) K(s) S_{glob}(s)$
- Power consumption of the first PLL or, equivalently, the global bandwidth limitation of the first PLL : $T_{Ref \rightarrow u_1}(s) = K(s) S_{glob}(s)$

Note that

$$\left| \frac{K(j\omega) S_{glob}(j\omega)}{K(j\omega) T_{lp}(j\omega)} \right| = \left| \frac{Pr(j\omega) K(j\omega) S_{glob}(j\omega)}{Pr(j\omega) K(j\omega) T_{lp}(j\omega)} \right| = \left| \frac{S_{glob}(j\omega)}{T_{lp}(j\omega)} \right| = \left| \frac{((I_N \otimes T_s) \star M)(j\omega)}{T_{lp}(j\omega)} \right|$$

From the proof of Theorem 3, if its conditions are satisfied then

$$\left| \frac{((I_N \otimes T_s) \star M)(j\omega)}{T_{lp}(j\omega)} \right| \leq \eta$$

Therefore

$$|T_{Ref \rightarrow u_1}(j\omega)|_{dB} \leq -|K(j\omega) T_{lp}(j\omega)|_{dB} + 20 \log_{10}(\eta)$$

and

$$|T_{Ref \rightarrow \varphi_1}(j\omega)|_{dB} \leq -|Pr(j\omega) K(j\omega) T_{lp}(j\omega)|_{dB} + 20 \log_{10}(\eta)$$

that is, for these specifications, the global performance specification is constrained by the local one.

4.3 Computation results

The numerical results for the proposed design approach are now presented. On the iterations on the matrices X , Y , Z and the weighting functions, we detail the numerical results for the last iteration, except for some graphical representation which can be present some intermediate results.

We first compute a filter F such that the global system is stable by applying Corollary 1 where condition **(ii)** is replaced by the resolution of the H_∞ problem defined by the first inequality of (23). In order to obtain the matrices X , Y and Z such that condition **(i)** of Corollary 1 is satisfied, the optimization problem (17) is solved. The circle center $c = 0.1$ satisfies condition (18) since $\frac{1}{2\max_i(|\operatorname{Re}(\lambda_i^{M_{11}})|)} = 0.5233$. Then the resolution of the optimization problem (17) gives $r = 0.947$, $X = -1.129$, $Y = 0.113$, $Z = 1$. For these values, the resolution of the H_∞ problem defined by the first inequality of (23) gives a filter F such that condition **(ii)** of Corollary 1 is satisfied:

$$F(s) = \frac{K(s)}{K_{pd}} = \frac{0.64(s + 205.9)}{s}.$$

The global system stability is confirmed by the maximum real part of its poles: -75.9. Fig.8 reveals that with this filter, neither the local system (blue dashed line) or the global system (blue full line) satisfies the performance specification (red dotted line).

We now focus on the design of a filter F such that the global system performance specification is satisfied by applying Theorem 3. Let us choose $\beta = r^2 \cdot \eta_0^2 = 316$ with $\eta_0 \approx 19.65$, that is 26 dB which is an upper bound on the difference between $|S_{glo}(j\omega)|$ and $|S_{loc}(j\omega)|$ for the previously computed filter F , see Fig.8. By solving the optimization problem (20),

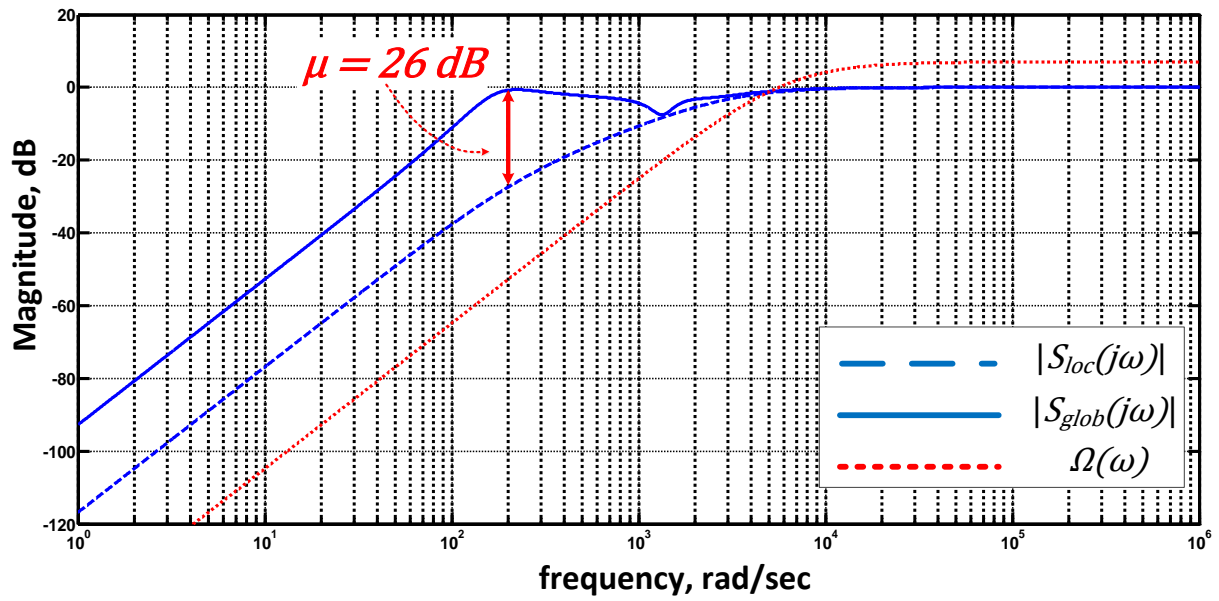


Figure 8: The magnitude of the local performance transfer function S_{loc} (blue dashed line), of the global performance transfer function S_{glob} (blue full line) and of the corresponding global performance frequency constraint (red dotted line)

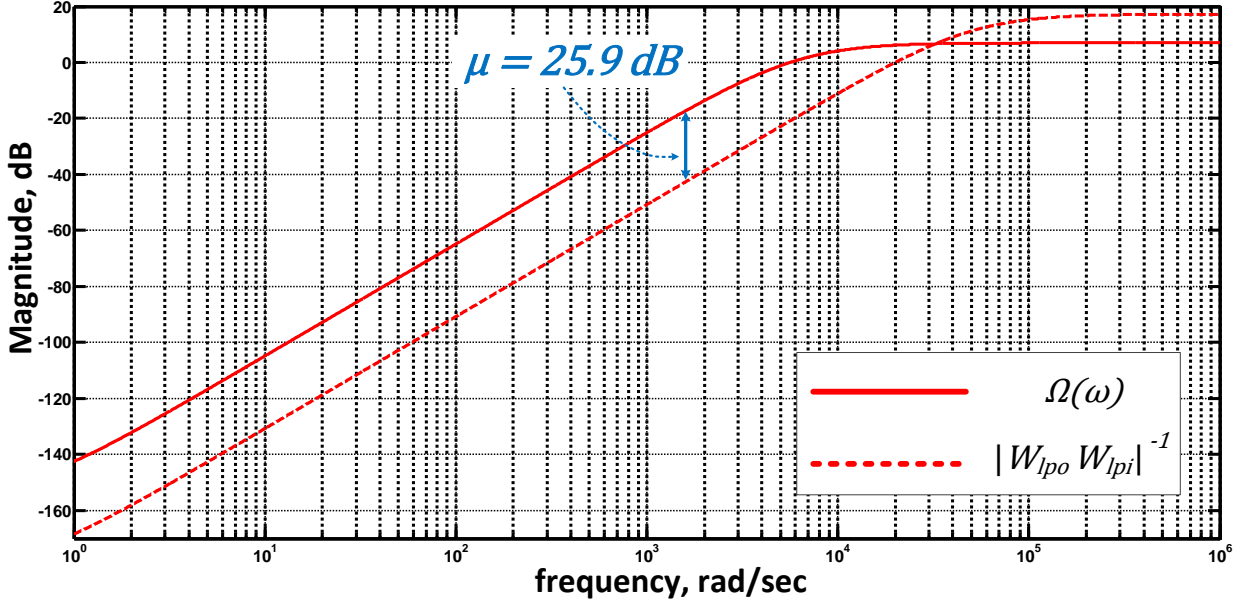


Figure 9: Chosen local (red dashed line) and global (red full line) frequency constraints we obtain: $r = 0.905$, $X = -1.236$, $Y = 0.124$, $Z = 1$, $\eta \approx 19.63$, and that is 25.9 dB.

The weighting functions W_{lpi} , W_{lpo} are chosen such that (32) is satisfied:

$$(W_{lpo}(s)W_{lpi}(s))^{-1} = \frac{7.28 (s + 0.56)^2}{(s + 50310)^2} \quad (34)$$

see Fig.9. Note that the difference between two constraints in logarithmic scale is equal to 25.9 dB for the whole frequency range except the high frequencies. Nevertheless, since in the high frequency range, the magnitude of the global transfer function S_{glob} is closed to 0 dB, see [50], this issue is not critical. The additional local weighting functions $W_{1(2)}^o$ and $W_{1(2)}^i$ are chosen as:

$$\begin{aligned} W_1^i(s) &= 9.22 \cdot 10^{-7} & W_2^i(s) &= 6.34 \cdot 10^{-6}; \\ W_1^o(s) &= \frac{2 \cdot 10^5 (s + 1.23 \cdot 10^6)}{s + 273.9} & W_2^o(s) &= 1010. \end{aligned}$$

Since $S_{loc} = G_{lp} \star K$ is the first input, first output transfer function of $\widehat{G} \star K$, a first choice of

W_1^o and W_1^i can be such that $|W_1^o W_1^i| \approx |W_{lpo} W_{lpi}|$. In this case, from equation (34), $W_1^o W_1^i$ has to be a transfer function of order at least 2, which is not the case. The main reason is that it is interesting to limit the order of the weighting functions, in order to obtain a filter of the smallest order. For this purpose, note that, from (33), $\|W_1^o Pr T_{lp} W_2^i\|_\infty < 1$. Then, $\|W_{lpo} G_{lp} \star K W_{lpi}\|_\infty < 1$ can be ensured by choosing W_1^o and W_2^i such that $W_1^o Pr W_2^i \approx W_{lpo} W_{lpi}$. As $Pr(s) = \frac{K_{VCO}/d}{s}$, W_1^i , W_2^i and W_2^o can be chosen constant and W_1^o a transfer function of order 1.

The resolution of the standard H_∞ problem (29) leads to the filter:

$$F(s) = \frac{49.7(s + 1483)}{s}.$$

The global stability, the local performance specification (26), the additional local performance specification (33) and the global performance (24) are ensured with this filter. The local performance transfer functions with the corresponding constraints are represented in Fig.10.

The evolution of the magnitude of global performance transfer function S_{glob} (dashed blue lines) along the different iterations is illustrated in the Fig.11. The cut-off frequency and Low Frequency (LF) range gain change until the global performance specification is reached. The result of the last iteration is represented by the full red line.

The results of the last iteration are presented in details in the Fig.12. The magnitude of the local performance transfer function S_{loc} (blue dashed line) satisfies the local frequency constraint $|W_{lpo}(j\omega) W_{lpi}(j\omega)|^{-1}$ (red dashed line). Furthermore the magnitude of the global performance transfer function S_{glob} (blue full line) satisfies the local frequency constraint

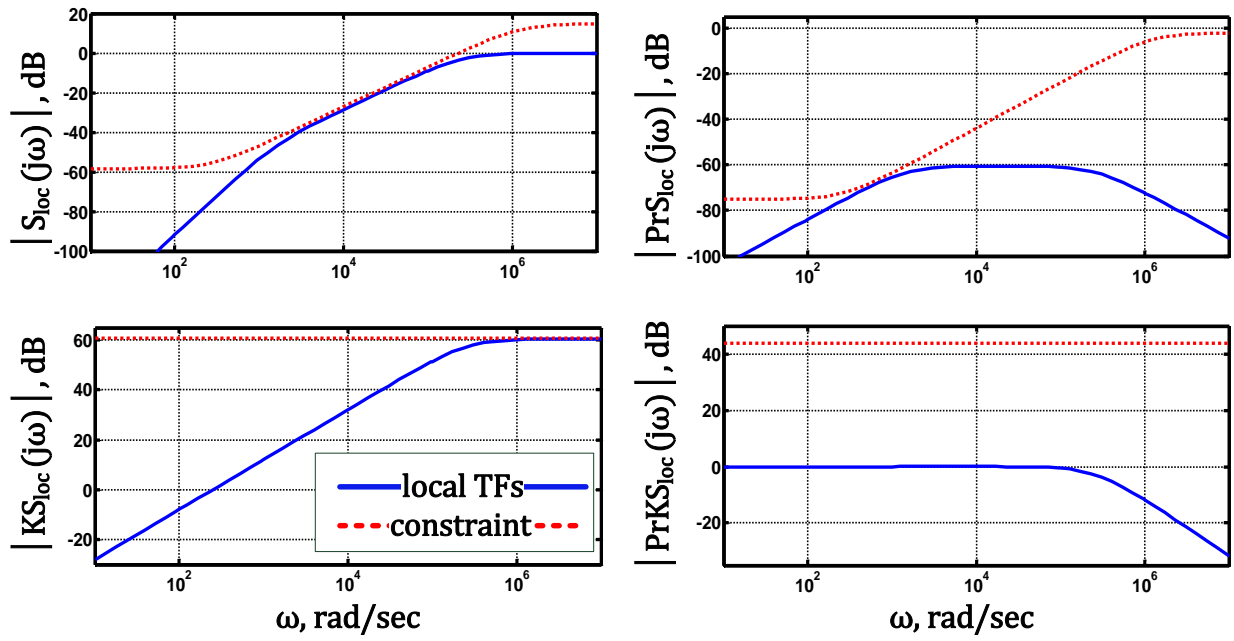


Figure 10: Additional local performance transfer function magnitudes (full blue lines) and corresponding constraints $\left| W_{1(2)}^o(j\omega) W_{1(2)}^i(j\omega) \right|^{-1}$ (dotted red lines)

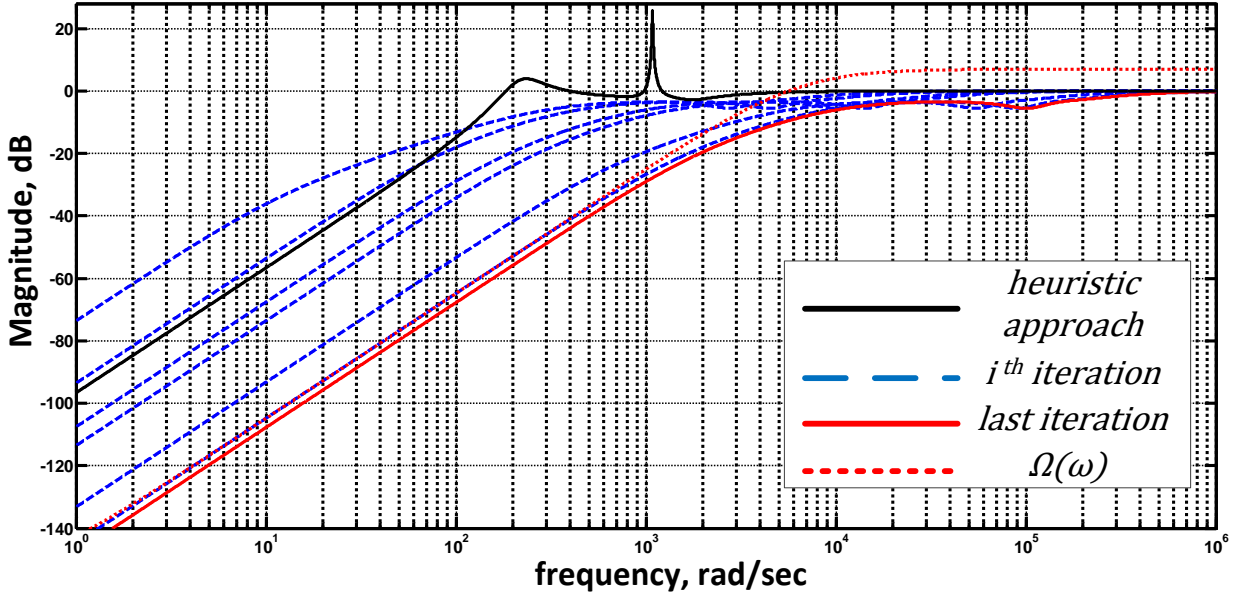


Figure 11: Obtained global performance transfer function magnitudes as a result of the heuristic design approach (black full line), of intermediate iteration (blue dashed lines) and last iteration (red full line) of the proposed design approach, the global frequency constraint (red dotted line)

$|S_{loc}(j\omega)|_{dB} + 20 \log(\eta)$ (green dashed-dotted line). Since the constraint $|S_{loc}(j\omega)| + 20 \log(\eta)$ is lower than the global performance frequency constraint $\Omega(j\omega)$ (full red line), the global performance specification is satisfied.

The benefit of the proposed design process with respect to the traditional one is illustrated by Fig.13 where the magnitudes of the frequency response of global (full lines) and local (dashed lines) transfer functions are represented for the traditional process (black lines) and the proposed process (blue lines). In Fig.14 is represented the tracking error in the time domain with both approaches. We observe that the time response and the damping are dramatically improved.

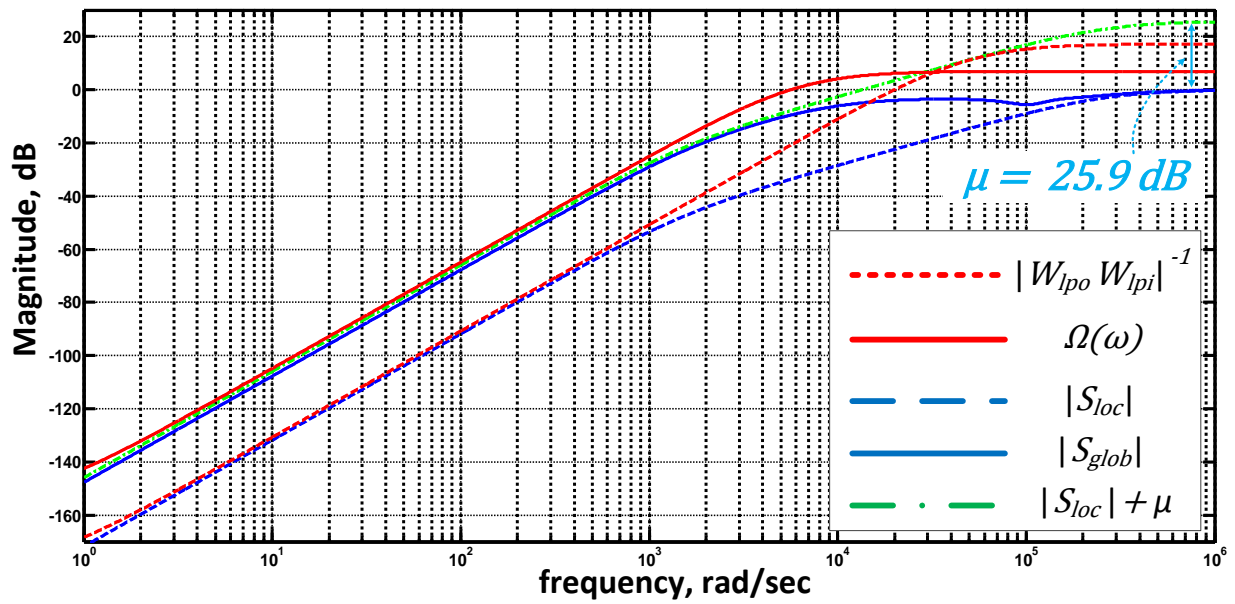


Figure 12: Magnitude of the local performance transfer function S_{loc} (blue dashed line) and of the global performance transfer function S_{glob} (blue full line), local frequency constraint $|W_{lpo}(j\omega)W_{lpi}(j\omega)|^{-1}$ (red dashed line), frequency constraint imposed by $|S_{loc}| + \mu$ (green dashed-dotted line) and the global performance frequency constraint (full red line) $\Omega(\omega)$

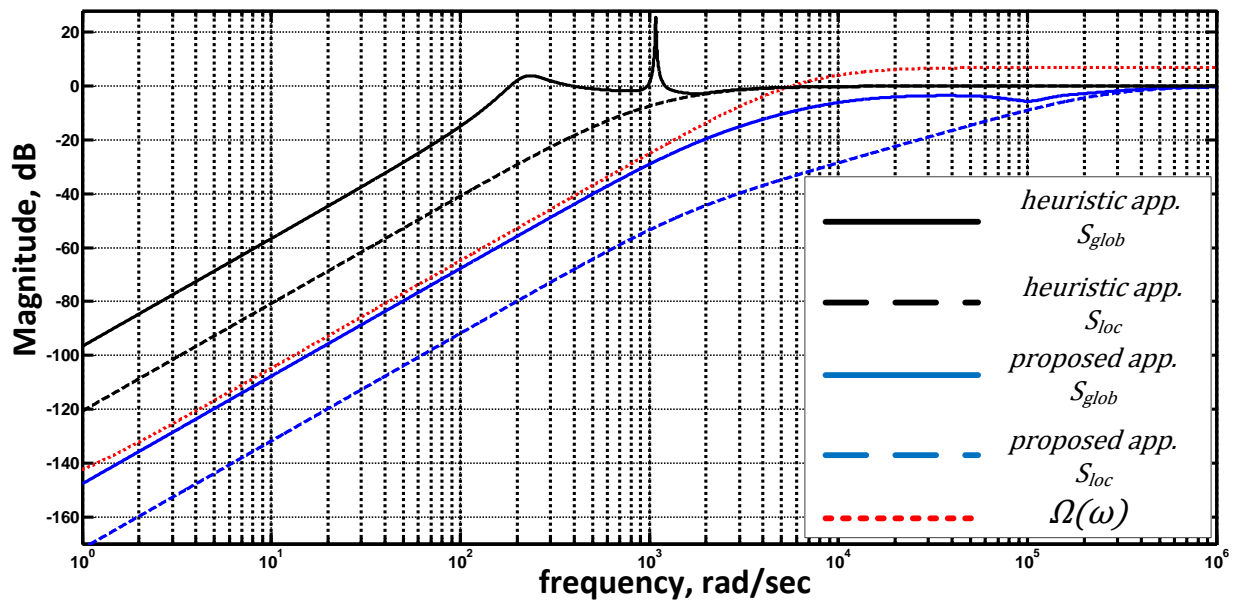


Figure 13: Magnitudes of designed global (full lines) and local (dashed lines) transfer functions for both heuristic (black lines) and proposed relative performance design (blue lines) approaches with corresponding global performance frequency constraint $\Omega(\omega)$ (dotted red line).

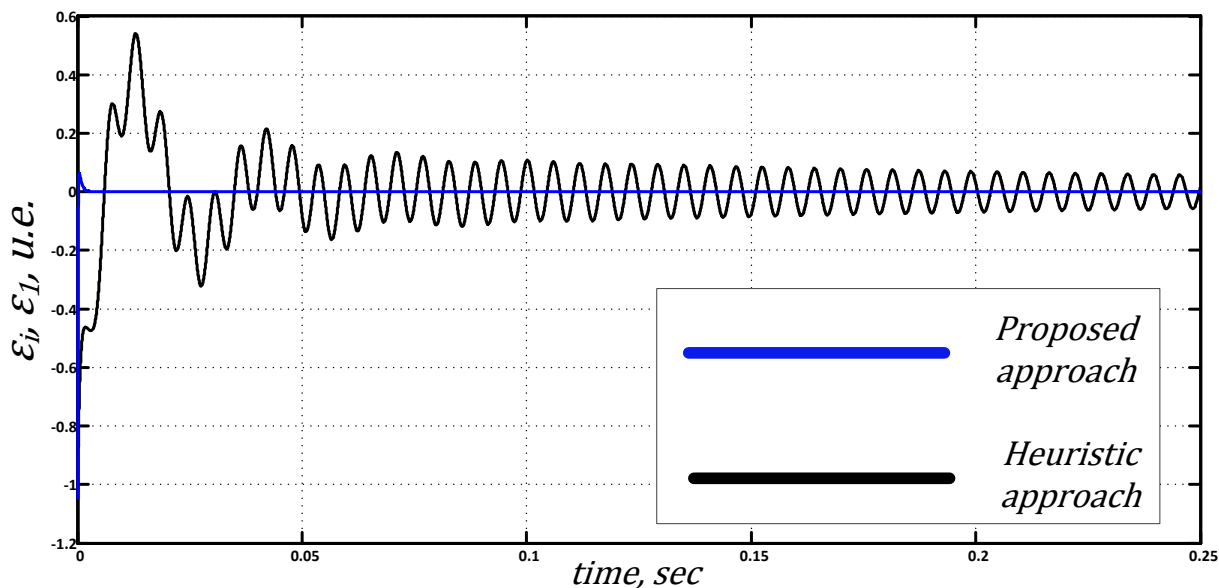


Figure 14: Temporal simulations results for both heuristic (black line) and proposed (blue line) design approaches

5 Conclusion

A new control law design method for large scale LTI systems defined as the given interconnection of identical subsystems is presented in this paper. The design method enforces a general global performance specification by designing a local controller such that the (stand-alone) closed loop subsystem satisfies performance specifications. This approach is motivated by improving the traditional design process of PLL network, rather than proposing a totally different one in order to ensure the acceptability of the method by the designers. The application to the design of a PLL network reveals that in addition to a priori ensure global stability and performance, a high level of performance is achieved with respect to the traditional design. The proposed design procedure has some common features with the

traditional iterative, “trial and error” approaches to decentralized control [6]. Thanks to the a priori guarantees of global stability and performance, it is a nice alternative to these traditional approaches.

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