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# Extended-Kalman-Filter-like observers for continuous time systems with discrete time measurements 

Vincent Andrieu *

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#### Abstract

In this short note is studied the observer introduced in [1] and [5]. The relationship between the Lipschitz constant and the measurement stepsize is exhibited. For second order system, we evaluate the authorized measurement stepsize.


## 1 Introduction

We consider a continuous time system of the form:

$$
\begin{equation*}
\dot{x}=f(x, u), x=\left(x_{1}, \ldots, x_{n}\right) \tag{1}
\end{equation*}
$$

with the state $x$ is in $\mathbb{R}^{n}$ and the input $u$ is in $U \subseteq \mathbb{R}$. The solutions of this system are denoted $\operatorname{as}$ D $x(t)$. The state of this system is accessible via a discrete time measure

$$
\begin{equation*}
y_{k}=C x\left(t_{k}\right), \tag{2}
\end{equation*}
$$

where $\left(t_{k}\right)_{k \in \mathbb{N}}$ is a sequence of positive real number defined as:

$$
t_{k+1}=t_{k}+\delta
$$

where $\delta$ is a positive real number.

[^0]The problem under consideration is an observer problem. How can we estimate the state of the system knowing only the measurements.

This problem has been addressed in [1] (see also [3] and the related filtering problem in (目).

Following what has been done for continuous time measurement in [2], we consider the case in which the system can be written (possibly after a change of coordinates) in the form:

$$
\begin{equation*}
\dot{x}=A x+\phi(x, u) \tag{3}
\end{equation*}
$$

with $A x=\left(x_{2}, \ldots, x_{n-1}, 0\right)$,

$$
y_{k}=C x\left(t_{k}\right)=x_{1}\left(t_{k}\right)
$$

and where the function $\phi: \mathbb{R}^{n} \times U \rightarrow \mathbb{R}^{n}$ satisfies an upper triangular globally Lipschitz condition, i.e. is such that for $i$ in $\{1, \ldots, n\}$ we have:

$$
\begin{equation*}
\left|\phi_{i}(x+e, u)-\phi_{i}(x, u)\right| \leq c_{L} \sum_{j=1}^{i}\left|e_{j}\right|, \forall(x, e) \in \mathbb{R}^{n} \times \mathbb{R}^{n}, u \in U \tag{4}
\end{equation*}
$$

where $c_{L}$ is a positive real number named the Lipschitz constant of the nonlinear system.
Inspired by [1] and [5], an estimate of the state can be given as any piecewise continuous function $\hat{x}: \mathbb{R}_{+} \rightarrow \mathbb{R}^{n}$ solution of the following continuous-discrete system ${ }^{[]}$

$$
\left\{\begin{array}{l}
\dot{\hat{x}}=A \hat{x}+\phi(\hat{x}, u)  \tag{5}\\
\dot{S}=-\theta S-A^{\prime} S-S A \\
\hat{x}\left(t_{k}\right)=\hat{x}\left(t_{k}^{-}\right)-\delta S\left(t_{k}\right)^{-1} C^{\prime}\left(C \hat{x}\left(t_{k}^{-}\right)-y_{k}\right) \\
S\left(t_{k}\right)=S\left(t_{k}^{-}\right)+\delta C^{\prime} C
\end{array}, t \in[k \delta,(k+1) \delta)\right.
$$

with $\theta$ a positive real number.
Note that compared to what has been done in [1], the Matrix updated law is a Lyapunov like one instead of a a Riccati like one. This one has been used in [5] since it allows to study analytically its limit and to give a better approximation on the bound involved in the highgain approach. Compare to [5] the only difference is the fact that the gain matrix depends on $S$ which is time varying.

## 2 Main Theorem

Inspired by the result of [6] , the following theorem can be obtain.

[^1]Theorem 1 There exists a positive real number $\zeta$ (which depends only on the dimension of the system), such that, for all $\delta<\frac{\zeta}{c_{L}}$ there exist two positive real numbers $\theta_{1}<\theta_{2}$ such that for any $\theta$ in $\left[\theta_{1}, \theta_{2}\right]$, the estimate $\hat{x}$ converges asymptotically to the state of the system (3).

Proof : Let $E=\Theta^{-1}(\hat{x}-x)$ where $\Theta=\operatorname{diag}\left\{1, \ldots, \theta^{n-1}\right\}$ with $\theta$ a positive real number larger then 1 to be specified. The scaled error satisfies for $t$ in $[k \delta,(k+1) \delta)$ :

$$
\dot{E}=\Theta^{-1} A[\hat{x}-x]+\Theta^{-1}[\phi(\hat{x}, u)-\phi(x, u)]
$$

and since $\Theta^{-1} A=\theta A \Theta^{-1}$, it yields for $t$ in $[k \delta,(k+1) \delta)$ :

$$
\dot{E}=\theta A E+\Delta_{\theta} \phi(x, E),
$$

with $\Delta_{\theta} \phi(x, E)=\Theta^{-1}[\phi(\hat{x}, u)-\phi(x, u)]$ such that $\left|\Delta_{\theta} \phi_{i}(x, E)\right| \leq c_{L} \sum_{j=1}^{i}\left|E_{j}\right|$ since $\theta>1$. Moreover, the scaled error satisfies:

$$
\begin{aligned}
E\left(t_{k}\right) & =\Theta^{-1}\left[\hat{x}\left(t_{k}^{-}\right)-\delta S\left(t_{k}\right)^{-1} C^{\prime}\left(C \hat{x}\left(t_{k}^{-}\right)-y_{k}\right)-x\left(t_{k}\right)\right] \\
& =\left(I-\delta \tilde{S}\left(t_{k}\right)^{-1} C^{\prime} C\right) E\left(t_{k}^{-}\right)
\end{aligned}
$$

where $\tilde{S}(\cdot)=\Theta S(\cdot) \Theta$. Note that $\tilde{S}$ satisfies:

$$
\begin{equation*}
\tilde{S}\left(t_{k}\right)=\tilde{S}\left(t_{k}^{-}\right)+\delta C^{\prime} C \tag{6}
\end{equation*}
$$

and for $t$ in $\left[t_{k},(k+1) \delta\right)$ :

$$
\begin{align*}
\dot{\tilde{S}} & =-\theta \tilde{S}-\Theta A^{\prime} \Theta^{-1} \tilde{S}-\tilde{S} \Theta^{-1} A \Theta \\
& =-\theta \tilde{S}-\theta A^{\prime} \tilde{S}-\theta \tilde{S} A \tag{7}
\end{align*}
$$

Note that the sequence $\left(s_{k}\right)_{k \in \mathbb{N}}$ such that $s_{k}=\theta \tilde{S}\left(t_{k}\right)$ satisfies

$$
s_{k+1}=\exp (-\rho) \exp \left(-A^{\prime} \rho\right) s_{k} \exp (-A \rho)+\rho C^{\prime} C
$$

where $\rho=\theta \delta$. Which gives,
$s_{k}=\exp (-k \rho) \exp \left(-A^{\prime} k \rho\right) s_{0} \exp (-A k \rho)+\sum_{\ell=0}^{k-1} \exp (-\ell \rho) \exp \left(-A^{\prime} \ell \rho\right) \rho C^{\prime} C \exp (-A \ell \rho)$,
Following what has been done in [5], the following Lemma can be shown (its proof is given in Section (3).

Lemma 1 There exist two continuous function $\alpha_{1}$ and $\alpha_{2}$ such that for all $\rho>0$, the matrix series converges to a limit denoted $s_{\infty}(\rho)$, formally defined as:

$$
s_{\infty}=\sum_{\ell=0}^{+\infty} \exp (-\ell \rho) \exp \left(-A^{\prime} \ell \rho\right) \rho C^{\prime} C \exp (-A \ell \rho)
$$

such that

$$
\begin{equation*}
\alpha_{1}(\rho) I \leq s_{\infty} \leq \alpha_{2}(\rho) I, \tag{8}
\end{equation*}
$$

where $\alpha_{1}$ and $\alpha_{2}$ are two strictly positive continuous functions.
The proof is divided in three steps:
Part 1: With this lemma in hand, we will first show that for all $\rho>0$, there exists $t_{0}$ such that for all $t>t_{0}$, the matrix $\tilde{S}$ satisfies:

$$
\begin{equation*}
\gamma_{1}(\rho) I<\theta \tilde{S}(t)<\gamma_{2}(\rho) I \tag{9}
\end{equation*}
$$

where $\gamma_{1}$ and $\gamma_{2}$ are two continuous positive function defined as,

$$
\begin{equation*}
\gamma_{1}(\rho)=\exp (-\rho) \alpha_{1}(\rho) c_{3}^{\operatorname{Int}(\rho)} c_{1}, \gamma_{2}(\rho)=\alpha_{2}(\rho) c_{2} c_{4}^{\operatorname{Int}(\rho)} \tag{10}
\end{equation*}
$$

where $\operatorname{Int}(\cdot)$ denotes the integer part of a positive real number, $c_{1}$ and $c_{2}$ are two positive real numbers and $\left(c_{3}, c_{4}\right)$ are two positive real numbers such that $c_{3}<1$ and $c_{4}>1$.

Indeed, for $k$ in $\mathbb{N}$ and $s$ in $[0, \delta), \tilde{S}(k+s)$ is solution of (7), hence,

$$
\theta \tilde{S}(k+s)=\exp (-\theta s) \exp \left(-A^{\prime} \theta s\right) s_{k} \exp (-A \theta s)
$$

With Lemma, for $k$ sufficiently large, the matrix $s_{k}$ is definite positive, it yields that for all $v$ in $\mathbb{R}^{n}$,

$$
v^{\prime} \theta \tilde{S}(k+s) v=\exp (-\theta s)\left|s_{k}^{\frac{1}{2}} \exp (-A \theta s) v\right|^{2}
$$

Hence, we get for all $v$ in $\mathbb{R}^{n}$ :

$$
v^{\prime} \theta \tilde{S}(k+s) v \geq \exp (-\rho)\left(\lambda_{\min }\left\{s_{k}^{\frac{1}{2}}\right\}\right)^{2}|\exp (-A \theta s) v|^{2}
$$

Note that

$$
\begin{aligned}
|\exp (-A \theta s) v|^{2} & =|\exp (-A \theta s) v|^{2} \\
& =|\exp (-A \operatorname{Int}(\theta s)) \exp (-A[\theta s-\operatorname{Int}(\theta s)]) v|^{2} \\
& \geq\left(\lambda_{\min }\left\{\exp (-A)^{\prime} \exp (-A)\right\}\right)^{\operatorname{Int}(\theta s)}|\exp (-A[\theta s-\operatorname{Int}(\theta s)]) v|^{2} \\
& \geq\left(\lambda_{\min }\left\{\exp (-A)^{\prime} \exp (-A)\right\}\right)^{\operatorname{Int}(\theta s)} c_{1}|v|^{2}
\end{aligned}
$$

where $c_{1}=\min _{r \leq 1}\left\{\lambda_{\min }\left\{\exp (-A r)^{\prime} \exp (-A r)\right\}\right\}$. Furthermore, note that $\operatorname{Det}\left\{\exp (-A)^{\prime} \exp (-A)\right\}=$ 1 , hence $c_{3}:=\lambda_{\min }\left\{\exp (-A)^{\prime} \exp (-A)\right\}<1$. Consequently,

$$
|\exp (-A \theta s) v|^{2} \geq c_{3}^{\operatorname{Int}(\rho)} c_{1}
$$

And since, $\left(\lambda_{\min }\left\{s_{k}^{\frac{1}{2}}\right\}\right)^{2} \geq \lambda_{\min }\left\{s_{k}\right\}$, it yields for all $k$ sufficiently large and all $v$ in $\mathbb{R}^{n}$

$$
v^{\prime} \theta \tilde{S}(k+s) v \geq \exp (-\rho) \lambda_{\min }\left\{s_{k}\right\} c_{3}^{\operatorname{Int}(\rho)} c_{1}|v|^{2}
$$

In a same way,

$$
v^{\prime} \theta \tilde{S}(k+s) v \leq\left(\lambda_{\max }\left\{s_{k}^{\frac{1}{2}}\right\}\right)^{2}|\exp (-A \theta s) v|^{2}
$$

and,

$$
|\exp (-A \theta s) v|^{2} \leq\left(\lambda_{\max }\left\{\exp (-A)^{\prime} \exp (-A)\right\}\right)^{\operatorname{Int}(\theta s)} c_{2}|v|^{2}
$$

where $c_{2}=\max _{r \leq 1} \lambda_{\max }\left\{\exp (-A r)^{\prime} \exp (-A r)\right\}$. Hence we get

$$
v^{\prime} \theta \tilde{S}(k+s) v \leq \lambda_{\max }\left\{s_{k}\right\} c_{2}\left(\lambda_{\max }\left\{\exp (-A)^{\prime} \exp (-A)\right\}\right)^{\operatorname{Int}(\theta s)}|v|^{2}
$$

Due to the fact that $c_{4}:=\lambda_{\max }\left\{\exp (-A)^{\prime} \exp (-A)\right\}>1$, we get

$$
v^{\prime} \theta \tilde{S}(k+s) v \leq \lambda_{\max }\left\{s_{k}\right\} c_{2} c_{4}^{\operatorname{Int}(\rho)}|v|^{2}
$$

which, with Lemma 2 implies that equation (9) holds and finishes the first part of the proof. Part 2: Let now $V(E)=E^{\prime} \tilde{S} E$. We will now show that it is strictly decreasing providing $\theta$ and $\rho$ satisfy an inequality constraint depending on $c_{L}$ the Lipschitz constant of the nonlinear system. The function $V$ satisfies along the trajectories of the system for $t$ in $[k \delta,(k+1) \delta)$ :

$$
\begin{align*}
\dot{V}(E) & =E^{\prime}\left[\theta A^{\prime} \tilde{S}(t)+\theta \tilde{S}(t) A+\dot{S}\right] E+2 E^{\prime} \tilde{S} \Delta_{\theta} \phi(x, E)  \tag{11}\\
& =-\theta E^{\prime} \tilde{S}(t) E+2 E^{\prime} \tilde{S}(t) \Delta_{\theta} \phi(x, E) \tag{12}
\end{align*}
$$

With Schwarz inequality, we have for all $t>0$

$$
\begin{equation*}
E^{\prime} \tilde{S}(t) \Delta_{\theta} \phi(x, E) \leq|E|\left|\tilde{S}(t) \Delta_{\theta} \phi(x, E)\right| \leq \lambda_{\max }\{\tilde{S}(t)\}|E|\left|\Delta_{\theta} \phi(x, E)\right| \tag{13}
\end{equation*}
$$

and we have ${ }^{\text {p }}$

[^2]\[

$$
\begin{equation*}
\left|\Delta_{\theta} \phi(x, E)\right|^{2}=\sum_{i=1}^{n}\left|\Delta_{\theta} \phi_{i}(x, E)\right|^{2} \tag{14}
\end{equation*}
$$

\]

$$
\leq \sum_{i=1}^{n} c_{L}^{2}\left(\sum_{j=1}^{i}\left|E_{j}\right|\right)^{2} \leq n c_{L}^{2}\left(\sum_{j=1}^{n}\left|E_{j}\right|\right)^{2} \leq n^{2} c_{L}^{2}|E|^{2}
$$

Consequently, we get:

$$
\begin{equation*}
\dot{V}(E) \leq\left[\lambda_{\max }\{\tilde{S}(t)\} 2 n c_{L}-\theta \lambda_{\min }\{\tilde{S}(t)\}\right]|E|^{2} \tag{15}
\end{equation*}
$$

Note that, due to equation (9), we get that if we can select $\theta$ and $\delta$ such that

$$
\begin{equation*}
\theta>\frac{2 \gamma_{2}(\rho) n c_{L}}{\gamma_{1}(\rho)} \tag{16}
\end{equation*}
$$

with $\rho=\delta \theta$ we would get

$$
\lambda_{\max }\{\tilde{S}(t)\} 2 n c_{L}-\theta \lambda_{\min }\{\tilde{S}(t)\}<0,
$$

for all $t$ sufficiently large. Hence we get $\dot{V}(E)<0$ for all $t$ in $[k \delta,(k+1) \delta)$, with $k$ sufficiently large. Moreover, we have

$$
\begin{aligned}
V(E)\left(t_{k}\right) & =E\left(t_{k}^{-}\right)^{\prime}\left(I-\delta \tilde{S}^{-1} C^{\prime} C\right)^{\prime} \tilde{S}\left(t_{k}\right)\left(I-\delta \tilde{S}^{-1} C^{\prime} C\right) E\left(t_{k}^{-}\right) \\
& =E\left(t_{k}^{-}\right)^{\prime}\left[\tilde{S}\left(t_{k}\right)-2 \delta C^{\prime} C+\delta^{2} C^{\prime} C \tilde{S}\left(t_{k}\right)^{-1} C^{\prime} C\right] E\left(t_{k}^{-}\right) \\
& =E\left(t_{k}^{-}\right)^{\prime}\left[\tilde{S}\left(t_{k}^{-}\right)-\delta C^{\prime} C+\delta^{2} C^{\prime} C\left(\tilde{S}\left(t_{k}^{-}\right)+\delta C^{\prime} C\right)^{-1} C^{\prime} C\right] E\left(t_{k}^{-}\right) \\
& =V(E)\left(t_{k}^{-}\right)-E\left(t_{k}^{-}\right)^{\prime} C^{\prime} p C E\left(t_{k}^{-}\right)
\end{aligned}
$$

where,

$$
p=\delta-\delta^{2} C\left(\tilde{S}\left(t_{k}^{-}\right)+\delta C^{\prime} C\right)^{-1} C^{\prime}
$$

Following (4] (see also [1]), note that if we note $q=\left[C \tilde{S}\left(t_{k}^{-}\right)^{-1} C^{\prime}+\frac{1}{\delta}\right]$, we have,

$$
\begin{aligned}
p q & =\left[\delta-\delta^{2} C \tilde{S}\left(t_{k}^{-}\right)^{-1}\left(I+\delta C^{\prime} C \tilde{S}\left(t_{k}^{-}\right)^{-1}\right)^{-1} C^{\prime}\right]\left[C \tilde{S}\left(t_{k}^{-}\right)^{-1} C^{\prime}+\frac{1}{\delta}\right] \\
& =1
\end{aligned}
$$

Hence, we get :

$$
V(E)\left(t_{k}\right)=V(E)\left(t_{k}^{-}\right)-E\left(t_{k}^{-}\right)^{\prime} C^{\prime} q^{-1} C E\left(t_{k}^{-}\right)
$$

with $q>0$. Consequently the function $V$ is decreasing along the trajectories of the system for $t$ sufficiently large. Its eigenvalues being lower and upper bounded as shown in the first part of the proof, we get convergence of the estimation error to the origin.
Part 3: We now show that the constraint we have on $\theta$ and $\rho$ can be given in terms of an upper bound on $\delta$ depending on $c_{L}$.

To obtain estimation, we have to find $\delta$ and $\theta$ such that (16) is satisfied and $\theta>1$. Note that (16) is equivalent with :

$$
\delta<\frac{\varphi(\rho)}{c_{L}}
$$

with $\varphi(\rho)=\frac{\rho \gamma_{1}(\rho)}{2 n \gamma_{2}(\rho)}$ where $\gamma_{1}$ and $\gamma_{2}$ are defined in (10). Hence, we get:

$$
\begin{equation*}
\varphi(\rho)=\rho \exp (-\rho) \frac{\alpha_{1}(\rho)}{2 n \alpha_{2}(\rho)} \frac{c_{1}}{c_{2}}\left(\frac{c_{3}}{c_{4}}\right)^{\operatorname{Int}(\rho)} \tag{17}
\end{equation*}
$$

Note that since $\alpha_{1}(\rho) \leq \alpha_{2}(\rho)$ and $c_{3}<1$ and $c_{4}>1$ we get:

$$
\lim _{\rho \rightarrow 0} \varphi(\rho)=0, \lim _{\rho \rightarrow+\infty} \varphi(\rho)=0
$$

Hence we can define the positive real number $\kappa$ as :

$$
\kappa=\max _{\rho>0}\{\varphi(\rho)\},
$$

This function being continuous, we get that for all $c_{L} \delta<\kappa$, there exists $\tilde{\rho}_{1}(\delta)$ and $\tilde{\rho}_{2}(\delta)$ such that

$$
\varphi(\rho)>c_{L} \delta, \forall \rho \in\left(\tilde{\rho}_{1}(\delta), \tilde{\rho}_{2}(\delta)\right)
$$

and,

$$
\varphi\left(\tilde{\rho}_{2}(\delta)\right)=\varphi\left(\tilde{\rho}_{1}(\delta)\right)=c_{L} \delta .
$$

The function $\varphi$ being definite positive, we get that $\lim _{\delta \rightarrow 0} \tilde{\rho}_{2}(\delta)=+\infty$. This implies that there exists $0<c_{L} \zeta \leq \kappa$ such that for all $0<\delta<\zeta$ there exist $\rho_{1}(\delta)$ and $\rho_{2}(\delta)$ such that we have

$$
\delta<\rho_{1}(\delta)<\rho_{2}(\delta), \varphi(\rho)>c_{L} \delta, \forall \rho \in\left[\rho_{1}(\delta), \rho_{2}(\delta)\right]
$$

Hence, this implies

$$
\frac{\varphi(\rho)}{c_{L}}>\delta, \forall \theta \in\left[\theta_{1}, \theta_{2}\right], 1<\theta_{1}<\theta_{2}
$$

where $\theta_{1}=\frac{\rho_{1}(\delta)}{\delta}$ and $\theta_{2}=\frac{\rho_{2}(\delta)}{\delta}$ which concludes the proof.

## 3 Proof of Lemma 2]

We want to show that the sequence

$$
s_{k}=\sum_{\ell=0}^{k} \exp (-\ell \rho) \exp \left(-A^{\prime} \ell \rho\right) \rho C^{\prime} C \exp (-A \ell \rho), k \in \mathbb{N}
$$

converges to a limit as $k$ goes to infinity denoted $s_{\infty}$ and such that

$$
\begin{equation*}
\alpha_{1}(\rho) I \leq s_{\infty} \leq \alpha_{2}(\rho) I \tag{18}
\end{equation*}
$$

First, note that $A$ being nilpotent, it yieldst

$$
C \exp (-A \ell \rho)=\sum_{j=0}^{n-1} C(-A \ell \rho)^{j}=r_{\ell \rho} S_{n}
$$

where $S_{n}=\operatorname{diag}\left(1, \ldots, \frac{1}{(n-1)!}\right)$ and, $r_{\ell \rho}$ is a vector defined as :

$$
r_{l}=\left(1,(-\rho \ell), \ldots,(-\rho \ell)^{n-1}\right)
$$

Consequently, we get

$$
s_{k}=\rho S_{n} M_{k}(\rho) S_{n},
$$

where $M_{k}$ is the symetric positive (at least) semi-definite matrix defined as

$$
M_{k}(\rho)=\left(\sum_{\ell=0}^{k} \exp (-\ell \rho) r_{\ell \rho}^{\prime} r_{\ell \rho}\right) .
$$

Hence to get the result it is sufficient to work on the sequence $M_{k}$ since we have for all $k$ in $\mathbb{N}$ where $M_{k}$ is definite positive:

$$
v^{\prime} s_{k} v=\rho\left|M_{k}^{\frac{1}{2}}(\rho) S_{n} v\right|^{2}
$$

${ }^{4}$ In fact we have

$$
C \exp (-A \ell \rho)=r_{\rho \ell} S_{n} \mathcal{O}
$$

where $\mathcal{O}$ is the observability matrix associated to the couple $(A, C)$, i.e.,

$$
\mathcal{O}=\left[\begin{array}{c}
C \\
C A \\
\vdots \\
C A^{n-1}
\end{array}\right]=I
$$

It yields,

$$
\rho\left(\lambda_{\min }\left\{M_{k}^{\frac{1}{2}}(\rho)\right\} \lambda_{\min }\left\{S_{n}\right\}\right)^{2} I \leq s_{k} \leq \rho\left(\lambda_{\max }\left\{M_{k}^{\frac{1}{2}}(\rho)\right\} \lambda_{\max }\left\{S_{n}\right\}\right)^{2} I
$$

or,

$$
\rho \lambda_{\min }\left\{M_{k}(\rho)\right\}\left(\lambda_{\min }\left\{S_{n}\right\}\right)^{2} I \leq s_{k} \leq \rho \lambda_{\max }\left\{M_{k}(\rho)\right\} \lambda_{\max }\left(\left\{S_{n}\right\}\right)^{2} I
$$

Consequently

$$
\frac{\rho}{(n-1)!} \lambda_{\min }\left\{M_{k}(\rho)\right\} I \leq s_{k} \leq \rho \lambda_{\max }\left\{M_{k}(\rho)\right\} I
$$

So, provided its limit exists, (18) is satisfied with:

$$
\begin{equation*}
\alpha_{1}(\rho)=\frac{\rho}{(n-1)!} \lambda_{\min }\left\{M_{\infty}(\rho)\right\}, \alpha_{2}(\rho)=\rho \lambda_{\max }\left\{M_{\infty}(\rho)\right\} \tag{19}
\end{equation*}
$$

where,

$$
M_{\infty}(\rho)=\lim _{k \rightarrow+\infty} M_{k}(\rho)
$$

1. We first show that the for all $k$ sufficiently large, the matrix $M_{k}$ is definite positive. First note that:

$$
M_{k}(\rho) \geq \exp (-(n-1) \rho)\left(\sum_{\ell=0}^{n-1} r_{\ell \rho}^{\prime} r_{\ell \rho}\right), \forall k \geq n
$$

Note that for all $v$ in $\mathbb{R}^{n}$,

$$
v^{\prime}\left(\sum_{\ell=0}^{n-1} r_{\ell \rho}^{\prime} r_{\ell \rho}\right) v=\sum_{\ell=0}^{n-1}\left|r_{\ell \rho} v\right|^{2}=|R(\rho) v|^{2} \geq \lambda_{\min }\left\{R(\rho)^{\prime} R(\rho)\right\}|v|^{2},
$$

where $R$ is a Vandermonde matrix $R=\left(r_{0}, \ldots, r_{n-1}\right)^{\prime}$. Hence, we get

$$
M_{k}(\rho) \geq \lambda_{\min }\left\{R(\rho)^{\prime} R(\rho)\right\} \exp (-(n-1) \rho) I, \forall k \geq n
$$

which shows that $M_{k}$ is definite positive for all $k \geq n$.
2. We now show that $M_{\infty}$ exists and is a continuous matrix function of $\rho$. Note that

$$
\left(M_{k}(\rho)\right)_{i, j}=\sum_{\ell=0}^{k} \exp (-\ell \rho)(-\ell \rho)^{i+j-2}=(-\rho)^{i+j-2} \frac{\partial^{i+j-2} \pi_{k}}{\partial \rho^{i+j-2}}(\rho),
$$

[^3]where $\pi_{k}(\phi)=\sum_{\ell=0}^{k} \exp (-\ell \rho)$. Note that we have
$$
\pi_{\infty}(\rho)=\lim _{k \rightarrow+\infty} \pi_{k}(\rho)=\frac{1}{1-\exp (-\rho)}
$$

Consequently,

$$
\left(M_{\infty}(\rho)\right)_{i, j}=(-\rho)^{i+j-2} \frac{\partial^{i+j-2} \pi_{\infty}}{\partial \rho^{i+j-2}}(\rho),
$$

Note that this function is continuous $\|$ for all $\rho>0$.

## 4 Numerical evaluation of $\zeta$

In this paragraph, we try to give the value of $\zeta$ in the simple case where the dimension of the system is $n=2$.
Computation of $M_{\infty}, \alpha_{1}$ and $\alpha_{2}$ : We have,

$$
M_{\infty}(\rho)=\left(\begin{array}{cc}
\frac{1}{1-\exp (-\rho)} & \frac{-\rho \exp (-\rho)}{(1-\exp (-\rho))^{2}} \\
\frac{-\rho \exp (-\rho)}{(1-\exp (-\rho))^{2}} & \frac{\rho^{2}[\exp (-\rho)+\exp (-2 \rho)]}{(1-\exp (-\rho))^{3}}
\end{array}\right)
$$

Note that when $n=2$, we have,

$$
\alpha_{1}(\rho)=\lambda_{\min }\left\{M_{\infty}(\rho)\right\}, \alpha_{2}(\rho)=\lambda_{\max }\left\{M_{\infty}(\rho)\right\}
$$

Employing WolframAlpha website, we are able to give explicitly this eigenvalues as a function of $\rho$.
Computation of $c_{1}, c_{2}, c_{3}$ and $c_{4}$ : We have also,

$$
\exp \left(-A^{\prime} r\right) \exp (A r)=\left(\begin{array}{cc}
1 & -r \\
-r & 1+r^{2}
\end{array}\right)
$$

$$
{ }^{6} \text { Moreover, it satisfies: } \quad \lim _{\rho \rightarrow 0}\left[\rho\left(M_{\infty}(\rho)\right)_{i, j}\right]=(-1)^{(i+j-2)}(i+j-2)!,
$$

Hence, we have:

$$
\lim _{\rho \rightarrow 0} \alpha_{1}(\rho)>0
$$

consequently,

$$
\lambda_{\min }\left\{\exp \left(-A^{\prime} r\right) \exp (A r)\right\}=\frac{2+r^{2}-r \sqrt{r^{2}+4}}{2},
$$

and,

$$
\lambda_{\max }\left\{\exp \left(-A^{\prime} r\right) \exp (A r)\right\}=\frac{2+r^{2}+r \sqrt{r^{2}+4}}{2}
$$

Note that $r \mapsto \lambda_{\min }\left\{\exp \left(-A^{\prime} r\right) \exp (A r)\right\}$ is a strictly decreasing function on $\mathbb{R}_{+}$and that $r \mapsto \lambda_{\max }\left\{\exp \left(-A^{\prime} r\right) \exp (A r)\right\}$ is a strictly increasing function on $\mathbb{R}_{+}$. Consequently, we obtain

$$
c_{1}=c_{2}=\frac{3-\sqrt{5}}{2}, c_{3}=c_{4}=\frac{3+\sqrt{5}}{2}
$$

Note that with these data, the function $\varphi$ defined in (17) can be computed threw matlab. Its maximal value is reached for

$$
\kappa=\max _{\rho>0}\{\varphi(\rho)\}=0.0022, \rho_{\text {opt }}=\operatorname{Argmax}_{\rho>0}\{\varphi(\rho)\}=0.5327>\kappa .
$$

Consequently, in this case, $\zeta=0.0022$.
Better approximation of $\gamma_{1}$ and $\gamma_{2}$ : In the particular case of $n=2$ a better approximation can be given. Indeed, we have

$$
\min _{s<\delta}\left\{\lambda_{\min }\left\{\exp (-A \theta s)^{\prime} \exp (-A \theta s)\right\}\right\}=\frac{2+\rho^{2}-\sqrt{\left(2+\rho^{2}\right)^{2}-4}}{2}, \rho=\theta \delta .
$$

Hence, we can take:

$$
\gamma_{1}(\rho)=\exp (-\rho) \alpha_{1}(\rho) \frac{2+\rho^{2}-\sqrt{\left(2+\rho^{2}\right)^{2}-4}}{2}, \gamma_{2}(\rho)=\alpha_{2}(\rho) \frac{2+\rho^{2}+\sqrt{\left(2+\rho^{2}\right)^{2}-4}}{2}
$$

The function $\varphi(\rho)=\frac{\rho \gamma_{1}(\rho)}{2 n \gamma_{2}(\rho)}$ can be computed threw matlab. Its maximal value is reached for

$$
\kappa=\max _{\rho>0}\{\varphi(\rho)\}=0.0148, \rho_{\text {opt }}=\operatorname{Argmax}_{\rho>0}\{\varphi(\rho)\}=0.5313>\kappa .
$$

Consequently, in this case, $\zeta=0.0148$. This implies that given the Lipschitz constant of the system $c_{L}$, the observer which maximizes the time between two measures can be tuned with

$$
\delta=\frac{0.0148}{c_{L}} \quad, \quad \theta=35.9 c_{L}
$$

## 5 Conclusion

In this short note, the continuous discrete observer presented in [5] has been studied. The maximal time between two measurement has been exhibited as a function of the Lipschitz constant. Even for small dimension, the obtained value seems to be small.

## References

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    ${ }^{1}$ Solution should be written $x\left(x_{0}, t, u(\cdot)\right)$ but to simplify the presentation we prefer this notation.

[^1]:    ${ }^{2}$ For a time function $x(\cdot)$ the notation $x\left(t_{k}^{-}\right)$means (when it exists) $\lim _{t \rightarrow t_{k}, t<t_{k}} x(t)$

[^2]:    ${ }^{3}$ Here we have used the inequality:

    $$
    \left(a_{1}+\ldots a_{n}\right)^{2} \leq n\left(a_{1}^{2}+\ldots+a_{n}^{2}\right)
    $$

[^3]:    ${ }^{5}$ This implies that it is a full rank matrix.

