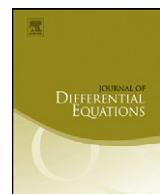




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Q1 Exponential Dichotomy on the real line: SVD and QR methods [☆]
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ABSTRACT

In this work we show when and how techniques based on the Singular Value Decomposition (SVD) and the QR decomposition of a fundamental matrix solution can be used to infer if a system enjoys—or not—exponential dichotomy on the whole real line.

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1. Introduction

Exponential dichotomy is at the heart of the fundamental perturbation results for linear systems of Coppel and Palmer ([10,11] and [24–26,28]), of the spectral theory of Sacker and Sell [31,32], of the geometric theory of Fenichel [19], of shadowing results [29,30], of perturbation results for invariant manifolds [18], of the fundamental perturbation results for connecting orbits of Beyn and Sandstede ([6–8] and [34]), and it has proven also a formidable ally to justify and gain insight into the behavior of various algorithmic approaches for solving boundary value problems, for approximating invariant surfaces and for computing traveling waves, among other uses (see [3,9,14,21]). Indeed, the property

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1 of exponential dichotomy is so **ubiquitous** and relevant that one would expect that there are several
 2 practical techniques to ascertain whether or not a given system enjoys exponential dichotomy on
 3 the real line. However, as far as we know, there are no reliable techniques allowing us to do that
 4 in general. The cases in which one is able to ascertain exponential dichotomy for a given system
 5 are perturbative in nature: slow varying systems (close to constant or periodic systems), roughness
 6 results (that is, systems close to systems having exponential dichotomy), L_1 perturbation results; e.g.,
 7 see [11], [8], [26] and [33].

8 Analysis and implementation of methods for approximating Lyapunov exponents of differential
 9 equations, and more generally also Exponential Dichotomy spectrum, henceforth Σ_L and Σ_{ED} respec-
 10 tively, *on the half-line* $t \geq 0$ (or $t \leq 0$), are rather well established. Our goal in this work is to lay down
 11 the theoretical ground work for the development of methods to ascertain if a system has exponential
 12 dichotomy (henceforth ED) *on the entire real line*. Our techniques will rest on the same methodolo-
 13 gies which have proven valuable to approximate spectra on the half-line. As a matter of fact, we will
 14 use the same assumptions needed in that context, ultimately some measure of integral separation in
 15 the system. As a result, our criteria to ascertain ED will not be of perturbation nature (for a system
 16 close to one which has or not have ED, such a constant coefficient problem), but will be resting on
 17 structural (and generic) properties of the underlying system.

18 We begin by reviewing the concepts of spectra on \mathbb{R}^+ (or \mathbb{R}^-) and lay down notation for later
 19 use.

21 1.1. Spectra of dynamical systems

22 Consider the n -dimensional dynamical system

$$25 \quad \dot{y} = f(y), \quad y(0) = y_0, \quad (1)$$

27 with solution $y(t, y_0)$. Consider the linearized problem

$$30 \quad \dot{x} = Df(y(t, y_0))x, \quad (2)$$

32 or simply

$$35 \quad \dot{x} = A(t)x, \quad (3)$$

36 where we suppose that A is continuous and bounded. With $X(\cdot)$, we will indicate a fundamental
 37 matrix solution of the system. The principal matrix solution will be written as $\phi(\cdot)$, so that $\dot{\phi} = A(t)\phi$,
 38 $\phi(0) = I$.

40 Spectra are defined for the linear problem (3), and in general they will depend on the solution
 41 trajectory of (1), that is on the initial condition y_0 . [There are important situations when this depen-
 42 dependency on y_0 can be removed, in a measure theoretical sense: [2,4,5,22,23].]

43 Two spectra associated to (3) are of interest to us: The Lyapunov spectrum, Σ_L , and the Exponen-
 44 tial Dichotomy (or Sacker-Sell) spectrum, Σ_{ED} .

45 The Lyapunov spectrum: Σ_L . This is defined in terms of upper and lower Lyapunov exponents (LEs for
 46 short). Given a fundamental matrix X for (3), define μ_j , $j = 1, \dots, n$, as

$$49 \quad \mu_j = \limsup_{t \rightarrow \infty} \frac{1}{t} \log \|X(t)e_j\|, \quad (4)$$

51 where the e_j 's are the standard unit vectors. When the sum of the numbers μ_j is minimized as we
 52 vary over all possible ICs (initial conditions) $X(0)$, the numbers are called (upper) Lyapunov exponents
 53 of the system, and the ICs are said to form a *normal basis*. We will write λ_j^s , $j = 1, \dots, n$, for the

1 ordered upper LEs of (3). By working with the adjoint system, $\dot{z} = -A^T(t)z$, one analogously defines
 2 its upper LEs which are called lower LEs of the original system (3), call them λ_j^i , $j = 1, \dots, n$, which
 3 again are considered ordered. The Lyapunov spectral intervals can now be defined:
 4

$$\Sigma_L := \bigcup_{j=1}^n [\lambda_j^i, \lambda_j^s]. \quad (5)$$

10 In case in which $\lambda_j^i = \lambda_j^s = \lambda_j$, for all $j = 1, \dots, n$, the system is called *regular*.
 11

13 *The Exponential Dichotomy, or Sacker–Sell, spectrum:* Σ_{ED} . This is defined in terms of exponential
 14 dichotomy (see [11]). Recall that (3) has Exponential Dichotomy on the half-line (i.e., $t, s \geq 0$ below) if
 15 there exist constants $K \geq 1$, $\alpha > 0$, and a projection P such that
 16

$$\begin{aligned} \|X(t)PX^{-1}(s)\| &\leq K e^{-\alpha(t-s)}, \quad t \geq s, \\ \|X(t)(I-P)X^{-1}(s)\| &\leq K e^{\alpha(t-s)}, \quad t \leq s, \end{aligned} \quad (6)$$

22 where X is a fundamental matrix solution. Σ_{ED} is defined to be the set of values $\lambda \in \mathbb{R}$ for which the
 23 shifted systems $\dot{y} = [A(t) - \lambda I]y$ do not have exponential dichotomy. For some $k \leq n$, Σ_{ED} is given by
 24 (see [31]) a collection of disjoint subintervals
 25

$$\Sigma_{ED} := [a_1, b_1] \cup \dots \cup [a_k, b_k]. \quad (7)$$

30 We make the simple, but fundamental, observation that (3) has ED on the half-line $t \geq 0$, iff $0 \notin \Sigma_{ED}$.
 31

32 **Remark 1.** By virtue of the roughness theorem for exponential dichotomies (see [10]), Σ_{ED} is stable,
 33 that is it is continuous with respect to perturbation in the coefficients $A(\cdot)$ of (3). Stability theory
 34 for Σ_L is more complicated (and restrictive), and one needs that the system enjoys the property of
 35 integral separation in some form (see [1,17]).
 36

38 *Directions: Growth subspaces.* Associated to the spectral intervals, both Σ_L and Σ_{ED} , there is very im-
 39 portant geometric information on the subspaces of solutions (i.e., of initial conditions) which achieve
 40 a certain asymptotic exponential growth; e.g., whose asymptotic growth is in the intervals of Σ_{ED} .
 41 As it turns out, this geometric information will be key for establishing ED on the real line, but it is
 42 typically neglected in algorithmic studies on approximation of spectra of dynamical systems.
 43

44 Let us define the forward stable subspace \mathcal{S}^+ to be the set of initial conditions leading to decreasing
 45 solutions in forward time:
 46

$$\mathcal{S}^+ = \{x \in \mathbb{R}^n : \|\Phi(t)x\| \rightarrow 0 \text{ as } t \rightarrow +\infty\}. \quad (8)$$

49 Let us also define the backward stable subspace \mathcal{S}^- (this is called unstable subspace by Sacker and
 50 Sell [31]) be the set of initial conditions leading to decreasing solutions in backward time:
 51

$$\mathcal{S}^- = \{x \in \mathbb{R}^n : \|\Phi(t)x\| \rightarrow 0 \text{ as } t \rightarrow -\infty\}. \quad (9)$$

1 *ED on the real line.* The problem (3) enjoys ED on the entire real line, when (6) holds with $t, s \in \mathbb{R}$.
 2 The following characterization of ED on the entire real line (see [10,11]) is the one which we will
 3 adopt in order to device techniques to verify ED:

4 5 System (3) has ED in \mathbb{R} if and only if it has ED in \mathbb{R}^+ and \mathbb{R}^- , and $\mathbb{R}^n = \mathcal{S}^+ \oplus \mathcal{S}^-$.

6 7 As a consequence, to check if system (3) has ED in \mathbb{R} , we will check that:

8 9 (ED-1) It has ED in \mathbb{R}^+ and \mathbb{R}^- .

10 11 (ED-2) The stable and the unstable subspaces of (8) and (9) are complementary. [That is, they form a
 12 basis for \mathbb{R}^n .]

13 14 [Notice that if there is ED in \mathbb{R}^+ and \mathbb{R}^- , then convergence to 0 as $t \rightarrow \pm\infty$ in (8) and (9) is expo-
 15 nentially fast.]

16 17 We are now ready to look at the two techniques which we propose to use to detect ED on the
 18 real line. As it turns out, these are the same techniques which have been used to approximate Σ_L and
 19 Σ_{ED} on the half-line. The first technique is based on the SVD of a fundamental matrix X : $X = U \Sigma V^T$.
 20 The second technique, which rests on the most popular approach to approximate Σ_L and Σ_{ED} on
 21 the half-line, is based on the QR factorization of a (normal) fundamental matrix solution X , that is
 $X = Q R$.

22 23 **Notation.** For a positive, non-vanishing function f , we will use the shorthand notation $\chi^s(f)$, $\chi^i(f)$
 24 and $\chi(f)$ to mean respectively $\limsup_{t \rightarrow +\infty} \frac{1}{t} \log(f)$, $\liminf_{t \rightarrow +\infty} \frac{1}{t} \log(f)$ and $\lim_{t \rightarrow \infty} \frac{1}{t} \log|f(t)|$.
 25 Similar notation will be used for the limit as $t \rightarrow -\infty$.

2. SVD method

26 27 In order to approximate Σ_L and Σ_{ED} on the half-line, SVD techniques use the information emerg-
 28 ing from the smooth SVD (if this is possible) of a fundamental solution X . That is, for $t \geq 0$, one seeks
 29 the decomposition $X(t) = U(t) \Sigma(t) V^T(t)$, with all factors being as smooth as X , $\Sigma = \text{diag}(\sigma_1, \dots, \sigma_n)$,
 30 and U and V orthogonal. To find this SVD, one can integrate the differential equations governing the
 31 evolution of the U , V , and Σ factors in the SVD of X . In [12], see also [16], we have shown that if the
 32 system (3) has stable and distinct Lyapunov exponents (a generic situation), then the decomposition
 33 $X = U \Sigma V^T$ with smooth factors is valid, differential equations for U and V can be given, and the
 34 evolution of the singular values is given by

$$37 \quad 38 \quad 39 \quad \dot{\sigma}_j = C_{jj}(t)\sigma_j, \quad j = 1, \dots, n, \quad C(t) := U^T(t)A(t)U(t). \quad (10)$$

40 41 Moreover, we have shown how, given any fundamental matrix solution X (i.e., any initial condition
 42 $X(0)$ for it), both Σ_L and Σ_{ED} can be recovered from knowledge of the diagonal of C . Finally, we have
 43 also shown, and this will be important soon, that—for systems with stable and distinct exponents—the
 44 orthogonal function V in the SVD converges exponentially fast to a constant matrix \bar{V} and that the
 45 columns of $X(0)\bar{V}$ span the growth subspaces associated to (Σ_L) and (Σ_{ED}) . It must be stressed that,
 46 with SVD techniques, we are able to obtain Σ_L and Σ_{ED} without the need to require the fundamental
 47 matrix solution X to be normal, any initial condition $X(0)$ will do (as long as the system has stable
 48 and distinct Lyapunov exponents). It comes natural then, to use the principal matrix solution Φ to
 49 characterize the desired quantities. Clearly in this case the growth subspaces (which are those in (8)
 50 and (9)) are given by the columns of \bar{V} . We refer to [13] for computational details on SVD techniques.

51 52 Perhaps surprisingly, we already have all the necessary ingredients in order to ascertain ED on \mathbb{R}
 53 by using the SVD method. In fact, it is sufficient to proceed as follows (see (ED-1)–(ED-2)):

54 55 (a) Relatively to (3) in forward and backward time, apply SVD techniques to the principal matrix
 56 solution Φ , to verify ED on the two half-lines \mathbb{R}^+ and \mathbb{R}^- . This requires verifying that $0 \notin \Sigma_{ED}$

1 for both forward and backward problems. If $0 \in \Sigma_{ED}$ for either forward or backward problem,
 2 then there is no ED. Otherwise, there is ED on the two half-lines. Also, recall that we expect (for
 3 systems with stable and distinct Lyapunov exponents) that the factors V of the SVDs **converge**
 4 (exponentially fast) to constant orthogonal matrices, in forward and backward time, call these \bar{V}^+
 5 and \bar{V}^- respectively.

6 (b) Take the columns of \bar{V}^+ and \bar{V}^- relative to the stable modes (those leading to intervals of Σ_{ED}
 7 to the left, respectively right, of 0): \bar{V}_s^+, \bar{V}_s^- . The key observation here (see [12]) is that \bar{V}_s^+ spans
 8 \mathcal{S}^+ and \bar{V}_s^- spans \mathcal{S}^- .

9 (c) Verify if $[\bar{V}_s^- \bar{V}_s^+]$ is a basis for \mathbb{R}^n . If it is, then the system has ED on \mathbb{R} , otherwise it does not.

10 To sum up, SVD techniques based on (a), (b) and (c) above can be used to verify ED on \mathbb{R} . But,
 11 our analytical results justifying this approach require the assumption of distinct and stable Lyapunov
 12 exponents. Although the assumption of stable Lyapunov exponents is natural, the need to have distinct
 13 Lyapunov exponents is violated in cases of practical interest (such as when the coefficient matrix A
 14 assumes a constant limiting value \bar{A} , and \bar{A} has complex conjugate eigenvalues). In our practical
 15 experience, the SVD method seems to work reliably also in these cases, but the lack of theoretical
 16 justification in the case of stable and not distinct LEs is bothersome, and we cannot guarantee that this
 17 technique will not encounter difficulties in the case of stable, but not distinct, Lyapunov exponents
 18 (either forward or backward in time). To avoid these difficulties, we then turn our attention to the
 19 next technique.

21 3. QR method

22 QR techniques are based on the (unique) decomposition of a *normal* fundamental matrix solution
 23 $X = QR$, where Q is orthogonal and R is upper triangular with positive diagonal. In practice, one
 24 only finds Q and (the logarithm of the) diagonal of R . For example, in a continuous realization of the
 25 QR method, we integrate differential equations for Q , so that $R = Q^T X$ would satisfy the triangular
 26 system $\dot{R} = B(t)R$ with $B(t) := Q^T A Q - Q^T \dot{Q}$. As it is well understood, see [5,15,16] for details
 27 on \mathbb{R}^+ , one can approximate Σ_L and Σ_{ED} by using only the diagonal of B ; in other words, only Q is
 28 needed and R is really never formed, since $\text{diag}(B) = \text{diag}(Q^T A Q)$.

29 So, to check (ED-1), one can use QR techniques to find Σ_{ED} for both forward and backward problems
 30 and verify that there is ED on the two half-lines. The issue is how to verify (ED-2), that is how
 31 to obtain information on the growth subspaces by the QR technique. It is our goal in this chapter to
 32 show that—under certain assumptions—one can obtain the needed directional information from the
 33 rescaled function $Y = R^{-1} \text{diag}(R)$. We will prove that Y “converges” to a matrix \bar{Y} for both forward
 34 and backward problems, and that these limiting values can be used to verify (ED-2).

35 **Remark 2.** As we previously observed, when using the QR technique we need to require that the
 36 fundamental matrix solution X be normal. But, we actually will need more: We will need that the
 37 initial condition $X(0)$ be chosen in such a way that conditions (12) and (13) below be verified in
 38 forward time, with the similar conditions (14)–(15) to be verified in backward time. In general, this
 39 request will imply that the initial conditions for the forward and backward integration will not be the
 40 same. An easy example of this is the diagonal system $\dot{x} = Ax$, with $A = \text{diag}(1, -2)$. Obviously, this
 41 system enjoys Exponential Dichotomy, and the principal matrix solution Φ is a normal fundamental
 42 matrix solution for this problem. However, by using Φ , condition (12) is verified, but (14) is not. If for
 43 the backward time integration we choose $X(0) = [e_2, e_1]$, then we will get $Q = [e_2, e_1]$, and then for
 44 $t \leq 0$ $B = \text{diag}(-2, 1)$ and (14) is verified. In other words, with respect to the notation of (8) and (9),
 45 we have $\mathcal{S}^+ = e_2$ and $\mathcal{S}^- = e_1$.

46 With the above remark in mind, our results can be summarized as follows: Given suitable initial
 47 conditions, $X^+(0)$ and $X^-(0)$, for the forward and backward problems respectively, the corresponding
 48 Y^+ and Y^- will “converge” (for $t \rightarrow +\infty$ and $t \rightarrow -\infty$, respectively) and the required growth sub-
 49 spaces can be extracted from these limiting Y ’s, for problems with *stable* Lyapunov exponents, though
 50 not necessarily distinct. [To be precise, for the case of **non-distinct** Lyapunov exponents, we show

1 convergence of Y to subspaces associated to the equal exponents, though we will not have that each
 2 column of Y converges.] We will also show that convergence is taking place exponentially fast. Once
 3 these theoretical results will be at hand, the basic technique to verify (ED-2) will be similar to what
 4 we did for the SVD:

5
 6 Use the columns of $X^+(0)\bar{Y}^+$ and $X^-(0)\bar{Y}^-$ relative to the forward and backward stable modes, $X^+(0)\bar{Y}_s^+$
 7 and $X^-(0)\bar{Y}_s^-$, and verify if $[X^-(0)\bar{Y}_s^- X^+(0)\bar{Y}_s^+]$ is a basis for \mathbb{R}^n .

8
 9 **Remark 3.** In our analysis below, we will focus only on the case $t \geq 0$, since, after rewriting assump-
 10 tions (12) and (13) in backward time, see (14) and (15), the results for $t \leq 0$ are identical.

11
 12 Assume now that a normal fundamental matrix solution $X(t)$ has been chosen (that is, the initial
 13 conditions $X(0)$ have been chosen) so that for its unique QR factorization $X(t) = Q(t)R(t)$, with Q
 14 orthogonal and R with positive diagonal entries, we have

$$\dot{R} = B(t)R,$$

16
 17 and the Lyapunov exponents can be computed from the diagonal of B (that is, of R) and are stable.
 18 Our purpose in this chapter is to show that—under these mild conditions—certain subspaces associ-
 19 ated to R converge and can be used to detect exponential dichotomy on the real line. The result is
 20 important, since we will only need to have stable Lyapunov exponents, though not necessarily stable
 21 and distinct, as it was the case for the SVD.

22 So, we will consider the system

$$\dot{x} = B(t)x, \quad (11)$$

23
 24 with B continuous, bounded and upper triangular. In what follows, we will assume that the diagonal
 25 elements of B satisfy one of the following assumptions:

26
 27 (i) B_{ii} and B_{jj} are integrally separated, i.e. there exist $a, d > 0$ such that

$$\int_s^t (B_{ii}(\tau) - B_{jj}(\tau)) d\tau \geq a(t-s) - d, \quad t \geq s \geq 0, \quad i < j, \quad (12)$$

28
 29 (ii) B_{ii} and B_{jj} are not integrally separated, but $\forall \epsilon > 0$ there exists $M_{ij}(\epsilon) > 0$ such that

$$\left| \int_s^t (B_{ii}(\tau) - B_{jj}(\tau)) d\tau \right| \leq M_{ij}(\epsilon) + \epsilon(t-s), \quad t \geq s \geq 0, \quad i < j. \quad (13)$$

30
 31 **Remark 4.** For the sake of completeness, we remark that for the backward problem $t \leq 0$, one should
 32 assume to have taken an initial condition $X(0)$ so that for the QR factorization $X(t) = Q(t)R(t)$, $t \leq 0$,
 33 one obtains a triangular system $\dot{R} = B(t)R$ satisfying the following analogs of (12) and (13):

34
 35 (i) B_{ii} and B_{jj} are integrally separated, i.e.

$$\int_s^t (B_{ii}(\tau) - B_{jj}(\tau)) d\tau \geq -a(t-s) - d, \quad t \leq s \leq 0, \quad i < j, \quad (14)$$

1 (ii) B_{ii} and B_{jj} are not integrally separated, but $\forall \epsilon > 0$ there exists $M_{ij}(\epsilon) > 0$ such that

$$2 \quad \left| \int_s^t (B_{ii}(\tau) - B_{jj}(\tau)) d\tau \right| \leq M_{ij}(\epsilon) - \epsilon(t-s), \quad t \leq s \leq 0, \quad i < j. \quad (15)$$

3 Considering again only the case of $t \geq 0$, assume moreover that the B_{ii} 's are arranged in p , $p \leq n$,
4 subsets as follows. Let n_1, \dots, n_p be such that $\sum_{i=1}^n n_i = n$, $m_i = \sum_{k=1}^i n_k$ and denote with $B_{jj}^{(i)}$ the
5 $(m_{i-1} + j)$ -th diagonal element of B . The set $\{B_{jj}^{(i)}, j = 1, \dots, n_i\}$ will be the i -th subset. Elements
6 belonging to the same subset satisfy assumption (13) while elements belonging to different subsets
7 satisfy (12) as follows

$$8 \quad \int_s^t (B_{jj}^{(i)}(\tau) - B_{kk}^{(i+1)}(\tau)) d\tau \geq a(t-s) - d, \quad j = 1, \dots, n_i, \quad k = 1, \dots, n_{i+1}.$$

9 In what follows we will use the notation B_{jj} where there is no ambiguity, otherwise we will use the
10 subset-notation $B_{jj}^{(i)}$. **The same** criteria will apply also to other quantities.

11 In [17], it was shown that the assumptions (12) and (13) guarantee that the Lyapunov spectrum
12 can be computed from the diagonal of B and that it is stable. In fact, in both cases (12) and (13), for
13 all $j = 1, \dots, n$, the LEs are given by

$$14 \quad \limsup_{t \rightarrow +\infty} \frac{1}{t} \log \int_0^t B_{jj}(\tau) d\tau = \chi^s(R_{jj}) = \lambda_j^s, \quad \liminf_{t \rightarrow +\infty} \frac{1}{t} \log \int_0^t B_{jj}(\tau) d\tau = \chi^i(R_{jj}) = \lambda_j^i.$$

15 **Remark 5.** Denote with R the principal matrix solution of (11) and let

$$16 \quad W_j = \left\{ w \in \mathbb{R}^n \mid \limsup_{t \rightarrow +\infty} \frac{1}{t} \log \|R(t)w\| \leq \lambda_j^s \right\},$$

17 i.e. W_j is the subspace of initial conditions leading to upper LEs lesser or equal to λ_j^s . (The W_j 's are
18 called lineals in the classical theory of LEs.) Then conditions (12) and (13) guarantee that W_j is also
19 the space of initial conditions leading to lower LEs lesser or equal to λ_j^i . Indeed, by [17, Theorem 4.2]
20 there exists a Lyapunov transformation T which reduces system (11) into block diagonal form. The
21 lineals for the new block system are the subspaces $\{e_{m_{j-1}+1}, \dots, e_{m_j}, \dots, e_n\}$ so that the lineals for
22 system (11) are the spaces $W_j = \text{span}\{T_{m_{j-1}+1}(0), \dots, T_{m_j}(0), \dots, T_n(0)\}$, where with e_j and T_j we
23 denoted respectively the j -th canonical vector of \mathbb{R}^n and the j -th column of T . By [17, Theorem 5.1],
24 W_j is also the set of initial conditions leading to lower LEs lesser than λ_j^i .

25 Now, let $Y = R^{-1} \text{diag}(R)$, so that Y satisfies the following ODE

$$26 \quad \dot{Y} = -YC(t), \quad (16)$$

27 where C is strictly upper triangular with entries

$$28 \quad C_{ij}(t) = B_{ij}(t) \frac{R_{jj}(t)}{R_{ii}(t)} = B_{ij}(t) e^{\int_0^t B_{jj}(\tau) - B_{ii}(\tau) d\tau}, \quad j > i. \quad (17)$$

29 Observe that Y is upper triangular with 1's on the diagonal.

1 **Notation.** We found it useful to adopt two different notations for the columns of Y . The standard Y_j
 2 for the j -th column, or the one inherited from the B_{jj} 's: $Y_j^{(i)}(t)$ for the $(m_{i-1} + j)$ -th column of $Y(t)$.
 3

4 **Lemma 6.** *The elements of C satisfy the following*

5 (i) $\chi^s(C_{ij}) < -a$, if B_{ii} and B_{jj} are integrally separated;
 6 (ii) for all $\epsilon > 0$, $\chi^{s,i}(B_{ij}) - \epsilon \leq \chi^{s,i}(C_{ij}) \leq \epsilon$ if B_{ii} and B_{jj} satisfy (13).

7 **Proof.** Let $M = \sup_{t \geq 0} \|B(t)\|$. If B_{ii} and B_{jj} belong to different subsets, say the k -th and the l -th
 8 respectively, $k < l$, then $|C_{ij}(t)| \leq (\sup_{t \geq 0} \|B(t)\|) e^{-a(l-k)t+d} = M e^{-a(l-k)t+d}$. It follows easily that
 9 $\chi^s(C_{ij}) < -(l-k)a < -a$.

10 If instead B_{ii} and B_{jj} belong to the same subset then $|C_{ij}(t)| \leq M e^{M_{ij}(\epsilon)+\epsilon t}$ and $|C_{ij}(t)| \geq$
 11 $|B_{ij}(t)| e^{-M_{ij}(\epsilon)-\epsilon t}$ for arbitrary ϵ . It follows that

$$\chi^{s,i}(B_{ij}) - \epsilon \leq \chi^{s,i}(C_{ij}) \leq \epsilon. \quad \square$$

12 **Corollary 7.** *Let Y be the matrix in (16). If the system is integrally separated, i.e. if the B_{ii} 's are integrally
 13 separated for all $i = 1, \dots, n$, then $Y \rightarrow \bar{Y}$, with \bar{Y} constant, upper triangular matrix with 1's on the diagonal.*

14 **Proof.** By Lemma 6 and integral separation of the B_{ii} 's it follows that $\chi^s(C_{ij}) < 0$ for $i \neq j$, while
 15 $C_{ii}(t) = 0$, for all t and for all $i = 1, \dots, n$. Then $C \rightarrow_{t \rightarrow \infty} 0$ exponentially fast so that $Y \rightarrow_{t \rightarrow \infty} \bar{Y}$. \square

16 **Lemma 8.** *Let Y be the matrix in (16). Then for all $\epsilon > 0$ sufficiently small, and any $j = 1, \dots, n_i$, $i = 1, \dots, p$,
 17 we have*

$$\chi^s(Y_j^{(i)}) \leq (j-1)\epsilon \leq (n_i - 1)\epsilon.$$

18 **Proof.** Eq. (16) can be solved explicitly and the elements of Y are given by the following expression

$$Y_{ij}(t) = Y_{ij}(0) - \int_0^t \sum_{k=i}^{j-1} Y_{ik}(\tau) C_{kj}(\tau) d\tau, \quad j > i,$$

$$Y_{ii}(t) = 1, \quad \forall t \geq 0. \quad (18)$$

19 The $Y_j^{(i)}$ column of Y satisfies the following

$$Y_j^{(i)}(t) = Y_j^{(i)}(0) - \sum_{k=1}^{m_{i-1}} \int_0^t C_{k,m_{i-1}+j}(t_1) Y_k(t_1) dt_1 - \sum_{k=m_{i-1}+1}^{m_{i-1}+j-1} \int_0^t C_{k,m_{i-1}+j}(t_1) Y_k(t_1) dt_1. \quad (19)$$

20 Then

$$\|Y_j^{(i)}(t)\| \leq \|Y_j^{(i)}(0)\| + \sum_{k=1}^{m_{i-1}} \int_0^t \|C_{k,m_{i-1}+j}(t_1) Y_k(t_1)\| dt_1 + \sum_{k=m_{i-1}+1}^{m_{i-1}+j-1} \int_0^t \|C_{k,m_{i-1}+j}(t_1) Y_k(t_1)\| dt_1. \quad (20)$$

Take first $i = 1$. We show through inductive reasoning that $\chi^s(Y_j^{(1)}) \leq (j-1)\epsilon$. Clearly $\chi^i(Y_1^{(1)}) = \chi^s(Y_1^{(1)}) = 0$. Notice that for $i = 1$, the first sum in (19) and in (20) is zero. As far as the second sum is concerned, we have $\chi^s(C_{kj}Y_k) \leq \chi^s(C_{kj}) + \chi^s(Y_k) \leq \epsilon + (k-1)\epsilon$ for $k = 1, \dots, j-1$. Then $\chi^s(\int C_{kj}Y_k) \leq k\epsilon$. It follows that $\chi^s(Y_j^{(1)}) \leq \max_{k=1, \dots, j-1}(\chi^s(\int C_{kj}Y_k)) \leq (j-1)\epsilon$. The proof for $i > 1$ is analogous and it follows upon noticing that the first sum goes to zero exponentially fast since for $k = 1, \dots, m_{i-1}$ we have $\chi^s(C_{k, m_{i-1}+j}Y_k) \leq \chi^s(C_{k, m_{i-1}+j}) + \chi^s(Y_k) \leq -a + \max_{l=1, \dots, i-1}(n_l - 1)\epsilon < 0$, where in the last inequality we picked $\epsilon < \max_{l=1, \dots, i-1} \frac{a}{(n_l - 1)}$. \square

Remark 9. Lemma 8 implies $\chi^s(Y_1^{(i)}) = \chi(Y_1^{(i)}) = 0$ for all $i = 1, \dots, p$.

In the two limiting cases of Lemma 8 we have:

- $p = n$ so that $n_i = 1$ for all $i = 1, \dots, n$ and $\chi^i(Y_j) = \chi^s(Y_j) = 0$, $j = 1, \dots, n$ (this follows from Corollary 7 as well);
- $p = 1$ so that $n_1 = n$ and $0 \leq \chi^i(Y_j) \leq \chi^s(Y_j) \leq (j-1)\epsilon$, $j = 1, \dots, n$.

Next, let $S_i(t) = \text{span}(Y_1^{(i)}(t), \dots, Y_{n_i}^{(i)}(t))$, $i = 1, \dots, p$.

Theorem 10. The spaces S_i converge for $t \rightarrow +\infty$: $\lim_{t \rightarrow +\infty} S_i(t) = \bar{S}_i$, $i = 1, \dots, p$.

Proof. Clearly the first n_1 columns of Y always span the same space so that $S_1(t) = \bar{S}_1 = \text{span}\{e_1, \dots, e_{n_1}\}$ for all t . Consider now, for all t , $S_2(t) = \text{span}\{Y_1^{(2)}(t), \dots, Y_{n_2}^{(2)}(t)\}$. Let $T > 0$ and fix $0 < \tau \leq 1$. We will prove the statement for S_2 , the proof for the other subspaces being analogous. Using (19) we have

$$Y_1^{(2)}(T + \tau) - Y_1^{(2)}(T) = - \sum_{k=1}^{n_1} \int_T^{T+\tau} C_{k, n_1+1}(t_1) Y_k^{(1)}(t_1) dt_1. \quad (21)$$

Lemmas 6 and 8 imply $\chi^s(C_{k, n_1+1}Y_k^{(1)}) \leq \chi^s(C_{k, n_1+1}) + \chi^s(Y_k^{(1)}) < -a + (n_1 - 1)\epsilon < 0$, if we choose $\epsilon < \frac{a}{(n_1 - 1)}$. It follows that the integrals in (21) tend to zero for $T \rightarrow +\infty$, uniformly in τ . Hence

$$\lim_{T \rightarrow \infty} (Y_1^{(2)}(T + \tau) - Y_1^{(2)}(T)) = 0,$$

and $Y_1^{(2)}(t)$ tends to a constant vector: $\bar{Y}_1^{(2)}$. We do two more steps. First we show that for $T \rightarrow \infty$, $\text{span}(\bar{Y}_1^{(2)}, Y_2^{(2)}(T), Y_2^{(2)}(T + \tau))$ is a **two-dimensional** subspace of \mathbb{R}^n . We have

$$\begin{aligned} Y_2^{(2)}(T + \tau) - Y_2^{(2)}(T) &= - \sum_{k=1}^{n_1} \int_T^{T+\tau} C_{k, n_1+2}(t_1) Y_k^{(1)}(t_1) dt_1 - \int_T^{T+\tau} C_{n_1+1, n_1+2}(t_1) Y_1^{(2)}(t_1) dt_1. \end{aligned} \quad (22)$$

When $T \rightarrow +\infty$ the first sum in the **right-hand** side goes to zero. Moreover, $Y_1^{(2)} \rightarrow \bar{Y}_1^{(2)}$, i.e. for all $\delta > 0$, there exists $T_\delta > 0$ such that for all $t > T_\delta$, and for $j = 1, \dots, n_1 + 1$,

$$\bar{Y}_{1j}^{(2)} - \delta \leq Y_{1j}^{(2)}(t) \leq \bar{Y}_{1j}^{(2)} + \delta.$$

1 It follows that for $j = 1, \dots, n_1 + 1$,

$$\lim_{T \rightarrow \infty} \int_T^{T+\tau} C_{n_1+1, n_1+2}(t_1) Y_{1j}^{(2)}(t_1) dt_1 = \lim_{T \rightarrow \infty} \int_T^{T+\tau} C_{n_1+1, n_1+2}(t_1) dt_1 \bar{Y}_{1j}^{(2)},$$

8 and therefore $\lim_{T \rightarrow \infty} (Y_2^{(2)}(T + \tau) - Y_2^{(2)}(T)) \in \text{span}(\bar{Y}_1^{(2)})$, since $Y_1^{(2)}(t)$ converges to $\bar{Y}_1^{(2)}$ exponentially fast.

10 Given the triangular structure of Y , this implies that these vectors belong to the same **two-dimensional** 11 subspace. Denote with $\bar{Y}_1^{(2)}$ and $\bar{Y}_2^{(2)}$ a basis for this subspace. For the next step we 12 have

$$\begin{aligned} Y_3^{(2)}(T + \tau) - Y_3^{(2)}(T) &= - \sum_{k=1}^{n_1} \int_T^{T+\tau} C_{k, n_1+3}(t_1) Y_k^{(1)}(t_1) dt_1 - \int_T^{T+\tau} C_{n_1+1, n_1+3}(t_1) Y_1^{(2)}(t_1) dt_1 \\ &\quad - \int_T^{T+\tau} C_{n_1+2, n_1+3}(t_1) Y_2^{(2)}(t_1) dt_1. \end{aligned} \quad (23)$$

23 As before, when $T \rightarrow \infty$, the first integral goes to 0. Moreover, reasoning as above, we obtain that

$$\lim_{T \rightarrow \infty} (Y_3^{(2)}(T + \tau) - Y_3^{(2)}(T)) \in \text{span}(\bar{Y}_1^{(2)}, \bar{Y}_2^{(2)}),$$

28 so that again, given the triangular structure of Y , these vectors belong to the same **three-dimensional** 29 subspace. The proof for the other $Y_j^{(i)}$ is analogous. \square

31 **Remark 11.** From the proof of Theorem 10, we notice that the (vector) functions $Y_1^{(i)}$ converge, as 32 $t \rightarrow \infty$, for all $i = 1, \dots, p$. Instead, the remaining $Y_j^{(i)}$ ($j \geq 2$) do not necessarily converge, though 33 they keep belonging to the same subspaces.

36 Our next goal is to show that by taking initial conditions in a direction identified by the subspaces 37 \bar{S}_i we converge to solutions having growth given by the i -th Lyapunov exponents. The following 38 **lemma** will be useful.

40 **Lemma 12.** Let $w \neq 0$ be a constant vector in \bar{S}_i . Write w in the basis $Y_1(t), \dots, Y_n(t)$, as 41 $w = \sum_{j=1}^n c_j(t) Y_j(t) = \sum_{k=1}^p \sum_{j=1}^{n_p} c_j^{(k)}(t) Y_j^{(k)}(t)$. Let $c^{(i)} = [c_1^{(i)}, \dots, c_{n_i}^{(i)}]^T$. Then

$$\chi^s(c^{(i)}) = \chi^i(c^{(i)}) = 0.$$

46 **Proof.** By Theorem 10, $c_j^{(k)} \rightarrow 0$ for $k \neq i$ and $\sum_{j=1}^{n_i} c_j^{(i)}(t) Y_j^{(i)}(t) \rightarrow w$. Assume by contradiction that 47 $\chi^s(c^{(i)}) = \alpha > 0$. Notice $\chi^s(c^{(i)}) \leq \max_{j=1, \dots, n_i} \chi^s(c_j^{(i)})$. Adopt for w the same block notation we 48 adopted for c . If $\chi^s(c_{n_i}^{(i)}) = \alpha$ then $c_{n_i}^{(i)}(t) Y_{n_i n_i}(t) = c_{n_i}^{(i)}(t) \rightarrow w_{n_i}^{(i)}$ and this is a constant quantity. So 49 it must be $\chi^s(c_{n_i}^{(i)}) = 0$. Similarly $c_{n_{i-1}}^{(i)}(t) + c_{n_i}^{(i)}(t) Y_{n_i, n_{i-1}}^{(i)}(t) \rightarrow w_{n_{i-1}}^{(i)}$ and this implies $\chi^s(c_{n_{i-1}}^{(i)}) \leq 0$. 50 Reasoning in this way, it must be $\chi^s(c_j^{(i)}) = 0$ for all $j = 1, \dots, n_1$ and this proves $\chi^s(c^{(i)}) = 0$. We 51 still need to show $\chi^i(c^{(i)}) = 0$. Clearly $\chi^i(c^{(i)}) \leq 0$. Assume by contradiction that $\chi^i(c^{(i)}) = -\alpha < 0$. 52 Then there exists an increasing sequence $(t_k)_{k \in \mathbb{N}}$, converging to ∞ , such that

$$\lim_{t_k \rightarrow +\infty} \frac{c_1^{(i)}(t_k)^2 + \cdots + c_{n_i}^{(i)}(t_k)^2}{e^{(-\alpha+\epsilon)t_k}} = 0,$$

for all $\epsilon > 0$. This means that $\lim_{t_k \rightarrow +\infty} \frac{c_j^{(i)}(t_k)^2}{e^{(-\alpha+\epsilon)t_k}} = 0$, for all $j = 1, \dots, n_i$ so that $\chi^i(c_j^{(i)}) \leq -\alpha$. Moreover, we have that $\limsup_{t_k \rightarrow +\infty} \frac{1}{t_k} \log(|c_j^{(i)}(t_k)| \|Y_j^{(i)}(t_k)\|) < 0$ for all $j = 1, \dots, n_i$, if we choose $\epsilon = \frac{\alpha}{n}$ in Lemma 8. But then, along this sequence, we would have

$$\chi(w) = \chi^s(w) = \limsup_{t_k \rightarrow +\infty} \frac{1}{t_k} \log \left(\left\| \sum_{j=1}^{n_i} c_j^{(i)}(t_k) Y_j^{(i)}(t_k) \right\| \right) < 0,$$

and this is a contradiction since w is a nonzero constant vector, and thus $\chi(w) = 0$. \square

Let $Z(t) = Y(t)^{-1}$. Then $Z(t)$ satisfies the following ODE

$$\dot{Z} = ZC, \quad (24)$$

where C is the matrix whose elements are defined in (17). The elements of Z are given by

$$Z_{ij}(t) = Z_{ij}(0) + \sum_{k=i+1}^j \int_0^t C_{ik}(s) Z_{kj}(s) ds, \quad i < j,$$

$$Z_{ii}(t) = 1, \quad \forall t \geq 0.$$

Denote with $z_i(t)^T$ the i -th row of Z so that $Z = \begin{pmatrix} z_1(t)^T \\ \vdots \\ z_n(t)^T \end{pmatrix}$. Lemma 8 applies to the elements of Z as well so that

$$\chi^s(z_j^{(i)}) \leq (j-1)\epsilon \leq (n_i-1)\epsilon. \quad (25)$$

Remark 13. If the system is integrally separated then $\chi^s(z_i) = \chi(z_i) = 0$, for all $i = 1, \dots, n$.

Remark 14. Eq. (25) implies $\chi^s(z_1^{(i)}) = \chi(z_1^{(i)}) = 0$, for $i = 1, \dots, p$.

Denote with $z_j^{(i)}(t)$ the $(m_i + j)$ -th row vector of $Z(t)$. Then, being Z the inverse of Y , we have that $z_i^T Y_i = 1$ so z_i is aligned with Y_i and perpendicular to all other Y_j 's. Then $\text{span}(z_1^{(i)}(t), \dots, z_{n_i}^{(i)}(t)) = \text{span}(Y_1^{(i)}(t), \dots, Y_{n_i}^{(i)}(t)) \rightarrow \bar{S}_i$.

Theorem 15. For all $w \neq 0$, $w \in \bar{S}_j$, and all $j = 1, \dots, p$:

$$\limsup_{t \rightarrow +\infty} \frac{1}{t} \log \|R(t)w\| = \lambda_j^s, \quad \liminf_{t \rightarrow +\infty} \frac{1}{t} \log \|R(t)w\| = \lambda_j^i,$$

with $R(t)$ fundamental matrix solution of (11).

1 **Proof.** We will first prove that $\chi^s(R(t)w) \geq \lambda_j^s$ (similarly, $\chi^i(R(t)w) \geq \lambda_j^i$). Then the statement will
2 follow from some geometrical considerations.

3 Rewrite w in the basis $\{Y_1(t), \dots, Y_n(t)\}$ as
4

$$5 w = \sum_{i=1}^n c_i(t) Y_i(t) = \sum_{k=1}^p \sum_{i=1}^{n_p} c_i^{(k)}(t) Y_i^{(k)}(t) = w_1 + \dots + w_p, \quad w_k = \sum_{i=1}^{n_k} c_i^{(k)} Y_i^{(k)}(t).$$

10 Notice that, by Lemma 12, we have $\chi(c_i^{(j)}) = 0$, and also $c_i^{(k)}(t) \rightarrow_{t \rightarrow \infty} 0$, $k \neq j$.
11

12 We now look at the characteristic exponent of $R(t)w$. We have
13

$$14 \|R(t)w\| = \|\text{diag}(R(t))Z(t)w\| = \begin{vmatrix} R_{11}(t)z_1(t)^T w \\ \vdots \\ R_{nn}(t)z_n(t)^T w \end{vmatrix} = \begin{vmatrix} R_{11}(t)c_1(t) \\ \vdots \\ R_{nn}(t)c_n(t) \end{vmatrix} \\ 15 \geq \begin{vmatrix} R_{11}^{(j)}(t)c_1^{(j)}(t) \\ \vdots \\ R_{n_j n_j}^{(j)}(t)c_{n_j}^{(j)}(t) \end{vmatrix} \geq R_{n_j n_j}^{(j)}(t) \begin{vmatrix} c_1^{(j)}(t) \\ \vdots \\ c_{n_j}^{(j)}(t) \end{vmatrix}. \quad (26)$$

23 This implies $\chi^s(R(t)w) \geq \lambda_j^s$, and $\chi^i(R(t)w) \geq \lambda_j^i$.
24

25 For $w \in \bar{S}_1$ we clearly have
26

$$27 \limsup_{t \rightarrow +\infty} \frac{1}{t} \log \|R(t)w\| = \lambda_1^s \quad \text{and} \quad \liminf_{t \rightarrow +\infty} \frac{1}{t} \log \|R(t)w\| = \lambda_1^i.$$

30 Next, denote with W_2 the space of initial conditions leading to exponential growth less than or
31 equal to λ_2^s , i.e. $W_2 = \{v \in \mathbb{R}^n \text{ s.t. } \limsup_{t \rightarrow +\infty} \frac{1}{t} \log \|R(t)v\| \leq \lambda_2^s\}$. Then $\dim(W_2) = n - n_1$ while
32 $\dim(\bar{S}_1 \oplus \bar{S}_2) = n_1 + n_2$ so that $\dim((\bar{S}_1 \oplus \bar{S}_2) \cap W_2) = n_2$. Let $w \in \bar{S}_1 \oplus \bar{S}_2$. If $w_1 \neq 0$ then by simply
33 following the same reasoning as before we obtain $\chi(c^{(1)}) = 0$ and $\limsup_{t \rightarrow +\infty} \frac{1}{t} \log \|R(t)w\| = \lambda_1^s$.
34 But then this leaves the only possibility that $\bar{S}_2 \subset W_2$ and $\limsup_{t \rightarrow +\infty} \frac{1}{t} \log \|R(t)w\| = \lambda_2^s$ for $w \in \bar{S}_2$.
35 The proof for the lower exponent is analogous. Notice that the results for the lower LEs follow from
36 Remark 5 as well. \square
37

38 3.1. QR convergence speed

41 Our next goal is to show that there is exponential convergence to the subspaces \bar{S}_i , $i = 1, \dots, p$.
42 We will do this following the approach used in [12] for the SVD, with appropriate modifications to
43 account for non-orthogonal factors and of dimension greater than 1.

44 Consider the unique smooth QL decomposition of Y : $Y(t) = U(t)L(t)$ for all t , with $U \in \mathbb{R}^{n \times n}$
45 orthogonal and $L \in \mathbb{R}^{n \times n}$ lower triangular with $L_{ii} > 0$.

46 Notice that $\text{span}\{y_1, \dots, y_n\} = \text{span}\{u_1, \dots, u_n\}$ for all $i = 1, \dots, n$.

47 Let $M(t) = L(t)^{-1}$ and notice that $Z(t) = Y(t)^{-1} = M(t)U^T(t)$, so that $M(t) = Z(t)U(t)$ and UM^T
48 is the unique QR decomposition of Z^T with positive diagonal elements.

50 **Lemma 16.** Use for the diagonal elements of M the same block notation as for B . Then
51

$$52 0 \leq \chi^i(M_{jj}^{(i)}) \leq \chi^s(M_{jj}^{(i)}) < (j-1)\epsilon \leq (n_i - 1)\epsilon, \quad \chi^s(M_{ij}) \leq \max_{k=1, \dots, p} (n_k - 1)\epsilon.$$

1 **Proof.** Obviously $M_{jj}^{(i)}(t) \neq 0$. Moreover $|M_{jj}^{(i)}(t)| = |z_j^{(i)}(t)^T u_j(t)| \leq \|z_j^{(i)}(t)\|$, where with u_j we de-
 2 note the j -th column of U . By (25) it follows $\chi^s(M_{jj}^{(i)}) \leq (j-1)\epsilon \leq (n_i-1)\epsilon$. Similarly for $M_{ij}(t) =$
 3 $z_i(t)^T u_j(t)$. \square

4
 5 **Remark 17.** If system (11) is integrally separated, then Lemma 16 can be restated as follows:
 6 $\chi^s(M_{ij}) = \chi(M_{ij}) = 0$, for all $i, j = 1, \dots, n$.

7
 8 **Remark 18.** Lemma 16 implies in particular that $\chi^s(M_{n_i+1,j}) = \chi(M_{n_i+1,j}) = 0$, for all $i = 1, \dots, p$,
 9 and $j = 1, \dots, n_i + 1$.

10
 11 **Remark 19.** Assumptions (12) and (13) imply that there exists $T > 0$ so that $R_{jj}^{(i)}(t) > R_{kk}^{(i+1)}(t)$, for
 12 $t > T$, $i = 1, \dots, p-1$, $j = 1, \dots, n_i$, and $k = 1, \dots, n_{i-1}$.

13
 14 Define $U_i(t) = \text{span}(u_1^{(i)}(t), \dots, u_{n_i}^{(i)}(t))$ for $i = 1, \dots, p$. Theorem 10 insures $S_i(t) = \text{span}(y_1^{(i)}(t), \dots,$
 15 $y_{n_i}^{(i)}(t)) \rightarrow \bar{S}_i$. Then clearly $U_p(t) = S_p(t) \rightarrow \bar{S}_p = \bar{U}_p$. Moreover $U_{p-1}(t) \oplus U_p(t) = S_{p-1}(t) \oplus S_p(t) \rightarrow$
 16 $\bar{S}_{p-1} \oplus \bar{S}_p$ and from the uniqueness of the orthogonal complement it follows $U_{p-1}(t) \rightarrow \bar{U}_{p-1}$. Reasoning in
 17 the same way, we can prove that $U_i \rightarrow \bar{U}_i$ for $i = p-2, \dots, 1$. To determine speed of
 18 convergence of U_i to \bar{U}_i , define $W_i(t) = U_i(t) \oplus \dots \oplus U_p(t)$, $\bar{W}_i = \bar{U}_i \oplus \dots \oplus \bar{U}_p$. Let $e = (n_1, \dots, n_p)$
 19 and $F_e(n)$ be the space of flags in \mathbb{R}^n of type e , i.e., the set of all filtrations $\mathcal{V} = (V_i)_{i=1}^p$ such that
 20 $\mathbb{R}^n = V_1 \supset \dots \supset V_p$. Given a filtration $\mathcal{V}^{(j)} = (V_i^{(j)})_{i=1}^p$, define the sets $K_i^{(j)}$ such that $K_p^{(j)} = V_p^{(j)}$,
 21 $V_i^{(j)} = K_i^{(j)} \oplus V_{i+1}^{(j)}$, and $K_i^{(j)} \perp V_{i+1}^{(j)}$, for $i = p-1, \dots, 1$. The following quantity defines a metric in
 22 $F_e(n)$ (see [20])

$$d(\mathcal{V}^{(1)}, \mathcal{V}^{(2)}) = \max_{\substack{i \neq j \\ \|x\| = \|y\| = 1 \\ x \in K_i^{(1)}, y \in K_j^{(2)}}} |(x, y)|^{\Delta/|\lambda_i^s - \lambda_j^s|}, \quad (27)$$

23 with $\Delta = \min_{i \neq j} |\lambda_i^s - \lambda_j^s|/(n-1)$. Let $P_i^{(j)}$ be the orthogonal projection into $K_i^{(j)}$. Then as in [12]

$$d(\mathcal{V}^{(1)}, \mathcal{V}^{(2)}) = \max_{i \neq j} \|P_i^{(1)} P_j^{(2)}\|^{\Delta/|\lambda_j^s - \lambda_i^s|}. \quad (28)$$

24 Notice that, for all t , $\mathcal{W}(t) = (W_i(t))_{i=1}^p \in F_e(n)$, and that for this filtration, $U_i(t) = K_i$, $i = 1, \dots, p$,
 25 where the K_i 's are the subspaces defined above. In the same way $\bar{\mathcal{W}} = (\bar{W}_i)_{i=1}^p \in F_e(n)$ and $\bar{U}_i = K_i$.
 26 Let $P_i(t)$ be the orthogonal projection into $U_i(t)$. Then $P_i(t) \rightarrow \bar{P}_i$, with \bar{P}_i orthogonal projection
 27 into \bar{U}_i . Our aim is to prove that the convergence of $P_i(t)$ to \bar{P}_i is exponential. To do so, we will first
 28 prove Theorem 20, and the sought result will follow from (30) together with (28).

29
 30 **Theorem 20.** Assume the diagonal elements of B satisfy either (12) or (13). Let $0 \leq \tau < 1$ be fixed and define
 31 $\beta_{ij}(t) = \|P_i(t + \tau) P_j(t)\|$, $i, j = 1, \dots, p$, where

$$P_i(t) = [u_1^{(i)}(t), \dots, u_{n_i}^{(i)}(t)] [u_1^{(i)}(t), \dots, u_{n_i}^{(i)}(t)]^T \quad (29)$$

32 is the orthogonal projection onto $U_i(t)$. Then, for arbitrary $\epsilon_0 > 0$ sufficiently small we have

$$\begin{aligned} \chi^s(\beta_{ij}) &\leq \min(\lambda_j^s - \lambda_i^i, -a(j-i)) + \epsilon_0, \quad j > i, \\ \chi^s(\beta_{ij}) &\leq \min(\lambda_i^s - \lambda_j^i, -a(i-j)) + \epsilon_0, \quad i > j. \end{aligned} \quad (30)$$

1 **Proof.** To prove (30) it suffices to show that for all $x \in U_j(t)$,

$$2 \quad 3 \quad 4 \quad \chi^s(\|P_i(t + \tau)P_j(t)x\|) \leq \min(\lambda_j^s - \lambda_i^s, -a(j - i)) + \epsilon_0.$$

5 Let us consider the case of $j > i$.

6 Take $x \in U_j(t)$, $x = \sum_{k=1}^{n_j} c_k^{(j)}(t)u_k^{(j)}(t)$. Let $c^{(j)}(t) = [0, \dots, 0, c_1^{(j)}(t), \dots, c_{n_j}^{(j)}(t), 0, \dots, 0]^T$ and
7 $D(t) = \text{diag}(R(t))$. Notice that the vector $D(t)M(t)c^{(j)}(t)$ has first m_{j-1} components equal to zero.
8 Then by Remark 19, for $t > T$ we have

$$9 \quad 10 \quad 11 \quad \|R(t)x\| = \left\| D(t)M(t)U^T(t) \sum_{k=1}^{n_j} c_k^{(j)}(t)u_k^{(j)}(t) \right\| = \|D(t)M(t)c^{(j)}(t)\| \leq \tilde{R}_{jj}(t) \|M(t)\| \|x\|, \quad (31)$$

15 with $\tilde{R}_{jj}(t) = \max_{k=1, \dots, n_j} R_{kk}^{(j)}$.

16 Rewrite x in the basis $u_1(t + \tau), \dots, u_n(t + \tau)$, as $x = \sum_{k=1}^p \sum_{l=1}^{n_k} b_l^{(k)}(t)u_l^{(k)}(t + \tau)$. In particu-
17 lar let $y = \sum_{l=1}^{n_i} b_l^{(i)}(t)u_l^{(i)}(t + \tau)$ be the orthogonal projection of x onto $U_i(t + \tau)$ so that $\|y\| =$
18 $\|P_i(t + \tau)P_j(t)x\|$. Let $N^{-1} = \sup_{t \geq 0} \|R(t + \tau, t)\|$, with $R(t + \tau, t)$ being the solution at time $t + \tau$ of
19 $\dot{R} = B(t)R$, $R(t, t) = I$. Then $R(t) = R(t + \tau, t)^{-1}R(t + \tau)$ and

$$20 \quad 21 \quad 22 \quad \|R(t)x\| \geq N \|R(t + \tau)x\| = N \left\| D(t + \tau)M(t + \tau) \begin{bmatrix} b_1(t + \tau) \\ \vdots \\ b_n(t + \tau) \end{bmatrix} \right\|. \quad (32)$$

23 Take first $i = 1, l = 1$. We want to estimate the exponential behavior of $b_l^{(i)} = b_1^{(1)}$. Then (31) and
24 (32) together with M being lower triangular imply

$$25 \quad 26 \quad 27 \quad |M_{11}(t)b_1^{(1)}(t)| \leq \frac{\tilde{R}_{jj}(t)}{R_{11}(t)} \|M(t)\| \|x\|,$$

28 and this together with (13) and Remark 18 imply

$$29 \quad 30 \quad 31 \quad \chi^s(b_1^{(1)}) = \chi^s(M_{11}b_1^{(1)}) \leq \min(\lambda_j^s - \lambda_1^i, -a(j - 1)).$$

32 Consider now $i = 1, l = 2$. Then as above

$$33 \quad 34 \quad 35 \quad |M_{21}(t)b_1^{(1)}(t) + M_{22}(t)b_2^{(1)}(t)| \leq \frac{\tilde{R}_{jj}(t)}{R_{22}(t)} \|M(t)\| \|x\|.$$

36 $\chi^s(|M_{21}b_1^{(1)} + M_{22}b_2^{(1)}|) \leq \max(\chi^s(M_{21}b_1^{(1)}), \chi^s(M_{22}b_2^{(1)}))$ and it is equal to the maximum if it is
37 unique. So we have the following possibilities:

38 (i) $\chi^s(M_{22}b_2^{(1)}) > \chi^s(M_{21}b_1^{(1)})$,
39 (ii) $\chi^s(M_{22}b_2^{(1)}) < \chi^s(M_{21}b_1^{(1)})$,
40 (iii) $\chi^s(M_{22}b_2^{(1)}) = \chi^s(M_{21}b_1^{(1)})$.

41 **Cases** (i) and (ii) both imply

$$42 \quad 43 \quad 44 \quad \chi^s(b_2^{(1)}) \leq \min(\lambda_j^s - \lambda_1^i, -a(j - 1)).$$

1 For case (iii) we get
 2

$$\begin{aligned} \chi^s(b_2^{(1)}) &\leq \chi^i(M_{22}) + \chi^s(b_2^{(1)}) \leq \chi^s(M_{22}b_2^{(1)}) \\ &\leq \chi^s(M_{21}) + \chi^s(b_1^{(1)}) < \epsilon + \min(\lambda_j^s - \lambda_1^i, -a(j-1)), \end{aligned}$$

7 where the last inequality follows from Lemma 16. The proof for $i = 1$ and $l = 3$ is analogous. In this
 8 case we have again three different possibilities:
 9

- 10 (i) $\chi^s(M_{33}b_3^{(1)}) > \chi^s(M_{31}b_1^{(1)})$, $\chi^s(M_{32}b_2^{(1)})$,
- 11 (ii) $\chi^s(M_{33}b_3^{(1)}) < \chi^s(M_{31}b_1^{(1)})$, or $\chi^s(M_{33}b_3^{(1)}) < \chi^s(M_{32}b_2^{(1)})$,
- 12 (iii) $\chi^s(M_{33}b_3^{(1)}) = \chi^s(M_{3j}b_j^{(1)}) > \chi^s(M_{3k}b_k^{(1)})$, with $j = 1, 2, k = 2, 1$.

15 Again cases (i) and (ii) are easy. For case (iii) (when $\chi^s(M_{33}b_3^{(1)})$ is not the only maximum), we
 16 examine the equality with $j = 2, k = 1$, which gives rise to the bound $\chi^s(M_{33}b_3^{(1)}) = \chi^s(M_{32}b_2^{(1)}) \leq$
 17 $\chi^s(M_{32}) + \chi^s(b_2^{(1)}) < 2\epsilon + \min(\lambda_j^s - \lambda_1^i, -a(j-1))$, where the 2ϵ term comes from the estimate in
 18 Lemma 16 together with the bound for $\chi^s(b_2^{(1)})$. The proof for $i = 1$ and $l = 4, \dots, n_1$ is analogous,
 19 and we will get the following estimate: $\chi^s(b_l^{(1)}) < (l-1)\epsilon + \min(\lambda_j^s - \lambda_1^i, -a(j-1))$.
 20

21 Finally, we have $\chi^s(\|y\|) = \frac{1}{2}\chi^s(\sum_{k=1}^{n_1}(b_k^{(1)})^2) \leq \max_{k=1, \dots, n_1} \chi^s(b_k^{(1)}) \leq \min(\lambda_j^s - \lambda_1^i, -a(j-1)) +$
 22 $(n_1-1)\epsilon$ and (30) is proved for $i = 1$ and $j > 1$. Here $\epsilon_0 = (n_1-1)\epsilon$. Notice that if we choose
 23 $\epsilon < \frac{a}{(n_1-1)}$, then $\min(\lambda_j^s - \lambda_1^i, -a(j-1)) + (n_1-1)\epsilon \leq \min(\lambda_j^s - \lambda_2^i, -a(j-2))$. This follows upon
 24 noticing that $\min(\lambda_j^s - \lambda_1^i, -a(j-1)) + a \leq \min(\lambda_j^s - \lambda_2^i, -a(j-2))$.
 25

26 Take now $i = 2, l = 1$. Then (32) and (31) imply
 27

$$\tilde{R}_{jj}(t) \|M(t)\| \|x\| \geq \|R(t)x\| \geq R_{11}^{(2)}(t) \left| \sum_{k=1}^{n_1} M_{n_1+1,k} b_k^{(1)}(t) + M_{n_1+1,n_1+1} b_1^{(2)}(t) \right|,$$

32 so that
 33

$$\chi^s \left(\sum_{k=1}^{n_1} M_{n_1+1,k} b_k^{(1)} + M_{n_1+1,n_1+1} b_1^{(2)} \right) \leq \min(\lambda_j^s - \lambda_2^i, -a(j-2)).$$

38 And again we have three possibilities:
 39

- 40 (i) $\chi^s(M_{n_1+1,n_1+1} b_1^{(2)}) > \chi^s(\sum_{k=1}^{n_1} M_{n_1+1,k} b_k^{(1)})$,
- 41 (ii) $\chi^s(M_{n_1+1,n_1+1} b_1^{(2)}) < \chi^s(\sum_{k=1}^{n_1} M_{n_1+1,k} b_k^{(1)})$,
- 42 (iii) $\chi^s(M_{n_1+1,n_1+1} b_1^{(2)}) = \chi^s(\sum_{k=1}^{n_1} M_{n_1+1,k} b_k^{(1)})$.

44 **Cases** (i) and (ii) imply $\chi^s(b_1^{(2)}) \leq \min(\lambda_j^s - \lambda_2^i, -a(j-2))$. Case (iii) implies $\chi^s(b_1^{(2)}) \leq$
 45 $\max_{k=1, \dots, n_1} \chi^s(M_{n_1+1,k} b_k^{(1)}) \leq \min(-a(j-1), \lambda_j^s - \lambda_1^i) + (n_1-1)\epsilon \leq \min(\lambda_j^s - \lambda_2^i, -a(j-2))$.
 46

47 For $i = 2, l = 2, \dots, n_2$ the proof is analogous to the one for $i = 1$. Reasoning in the same way we
 48 will obtain $\chi^s(b_l^{(2)}) \leq (n_2-1)\epsilon + \min(\lambda_j^s - \lambda_2^i, -a(j-2))$.
 49

50 The statement of the theorem follows by choosing $\epsilon_0 = (n_1-1)\epsilon < a$, for $i = 1, \dots, j-1$.
 51

52 The proof for $i > j$ is analogous. \square

53 **Corollary 21.** Assume system (11) to be integrally separated and let $\beta_{ij}(t) = \|P_i(t+\tau)P_j(t)\|$, where $P_i(t)$ is
 54 the orthogonal projection onto $U_i(t)$ defined in (29). Then

$$\begin{aligned} \chi^s(\beta_{ij}) &\leq \min(\lambda_j^s - \lambda_i^i, -a(j-i)), \quad j > i, \\ \chi^s(\beta_{ij}) &\leq \min(\lambda_i^s - \lambda_j^i, -a(i-j)), \quad i > j. \end{aligned}$$

Finally, we are ready to estimate the rate of convergence of the orthogonal projections $P_i(t)$ to \bar{P}_i , $i = 1, \dots, p$. Let $\alpha_{ij}(t) = \|P_i(t)\bar{P}_j\|$, $i, j = 1, \dots, p$.

Theorem 22. *Let the diagonal elements of B satisfy assumption (13) or (12). Then for all $i, j = 1, \dots, p$,*

$$\chi^s(\alpha_{ij}) \leq A |\lambda_i^s - \lambda_j^s|, \quad i \neq j, \quad \chi^s(1 - \alpha_{ii}) \leq \max_{j \neq i} \chi^s(\alpha_{ij}), \quad (33)$$

where $A = \max_{k \neq l} (\chi^s(\beta_{kl}) / |\lambda_k^s - \lambda_l^s|)$.

Proof. The proof for $i \neq j$ is much the same as that of [12, Theorem 5.4]. We omit the details here, and we just point out that the result follows from (30) together with (28).

In what follows, we prove the result for $i = j$. Since $P_1(t) + P_2(t) + \dots + P_p(t) = I$, with I the identity matrix in $\mathbb{R}^{n \times n}$, we can rewrite $\bar{P}_i = \sum_{j=1}^p P_j(t)\bar{P}_i$. Then $1 = \|\bar{P}_i\| \leq \sum_{j=1}^p \|P_j(t)\bar{P}_i\|$ so that $(1 - \|\bar{P}_i P_i(t)\|) \leq \sum_{j \neq i} \|P_j(t)\bar{P}_i\|$ and this rewrites as $(1 - \alpha_{ii}(t)) \leq \sum_{j \neq i} \alpha_{ij}(t)$. It follows that $\chi^s(1 - \alpha_{ii}) \leq \chi^s(\sum_{j \neq i} \alpha_{ij}) \leq \max_{j \neq i} \chi^s(\alpha_{ij})$. \square

Remark 23. The importance of exponential convergence to the relevant subspaces cannot be stressed enough. For example, in a numerical implementation of methods to ascertain ED, it implies that one can truncate the real line to a finite (large) interval, ascertain ED on this finite interval, and then invoke perturbation results to conclude that the original problem had ED on the infinite interval. It is also important to stress that the assumptions we have needed in order to infer exponential convergence of the subspaces are *structural assumptions* on the system, namely some form of integral separation on the two half-lines (a generic assumption, see [24]), or more precisely we assumed that the system has stable Lyapunov exponents in forward and backward time. This is in contrast to existing results to infer ED, which require strong assumptions on the type of system at hand, e.g. slow varying systems, or systems for which roughness and L_1 perturbation results can be applied; again, see [8,11,19,21,26,33,34].

4. Exponential dichotomy from QR on \mathbb{R}

In Sections 2 and 3, we have used the equivalence to having ED given by the conditions (ED-1) and (ED-2), to see how the forward and backward stable subspaces can be obtained from the SVD or QR of appropriate fundamental matrix solutions. In this section we consider an alternative strategy, namely we give sufficient conditions that guarantee that the Sacker-Sell spectrum (on the whole line) is bounded away from the origin.

Consider the linear inhomogeneous problem

$$\dot{y} = A(t)y + g(t), \quad (34)$$

where A and g are uniformly bounded and A is piecewise continuous. Recall (see [11, p. 67]) that the homogeneous system (that is (3)) has exponential dichotomy on \mathbb{R} , that (6) holds with $t, s \in \mathbb{R}$, if and only if (34) has a unique bounded solution for any bounded and continuous function g .

First of all, we envision performing a global time dependent orthogonal change of variables ($x(t) = Q^T(t)y(t)$ and $f(t) = Q^T(t)g(t)$) which brings the system in upper triangular form B . We write this system in block form

$$\dot{x} = B(t)x + f(t), \quad B(t) = \begin{pmatrix} B_{11}(t) & B_{12}(t) \\ 0 & B_{22}(t) \end{pmatrix}. \quad (35)$$

Below, let $\|x\| = \sup_t \|x(t)\|_2$, $\|f\| = \sup_t \|f(t)\|_2$, and $\|A\| = \sup_t \|A(t)\|_F$.

1 Next, we make the following assumptions on a fundamental matrix solution R associated to B :
 2 $\dot{R} = B(t)R$ (these are effectively assumptions on the existence of a transformation Q leading to a
 3 triangular problem for which these assumptions are satisfied). There exist constants $\alpha_1 > 0$, $\alpha_2 > 0$
 4 and $K_1 \geq 1$, $K_2 \geq 1$ such that

$$6 \quad \|R_{11}(t)R_{11}^{-1}(s)\| \leq K_1 e^{-\alpha_1(s-t)}, \quad s \geq t, \quad (36)$$

$$7 \quad \|R_{22}(t)R_{22}^{-1}(s)\| \leq K_2 e^{-\alpha_2(t-s)}, \quad t \geq s. \quad (37)$$

10 **Theorem 24.** If (36) and (37) hold, then (11) has ED on \mathbb{R} ; that is, (6) holds with $t, s \in \mathbb{R}$, $X(t) = R(t) =$
 11 $\begin{pmatrix} R_{11}(t) & R_{12}(t) \\ 0 & R_{22}(t) \end{pmatrix}$, with

$$14 \quad R_{12}(t) = -R_{11}(t) \int_t^\infty R_{11}^{-1}(\tau) B_{12}(\tau) R_{22}(\tau) d\tau \quad (38)$$

18 and $P = \begin{pmatrix} 0 & 0 \\ 0 & I \end{pmatrix}$. Indeed, with this R , there exists a bounded solution x of (35) such that

$$21 \quad \|x\| \leq \|f\| \cdot \left\{ \frac{K_2}{\alpha_2} \sqrt{1 + \kappa^2 \frac{K_1^2}{(\alpha_1 + \alpha_2)^2}} + \frac{K_1}{\alpha_1} \sqrt{1 + \kappa^2 \frac{K_2^2}{(\alpha_1 + \alpha_2)^2}} \right\}, \quad (39)$$

25 where $\kappa = \|B_{12}\|$.

27 **Proof.** We write by the variation of parameters formula

$$29 \quad x(t) = \int_{-\infty}^t R(t) P R^{-1}(s) f(s) ds - \int_t^\infty R(t) (I - P) R^{-1}(s) f(s) ds. \quad (40)$$

33 We employ the upper triangular fundamental matrix solution $R(t)$ (see [16, Lemma 7.4]) where $R_{12}(t)$
 34 is given by (38). Then

$$36 \quad R(t) P R^{-1}(s) = \begin{pmatrix} 0 & R_{12}(t) R_{22}^{-1}(s) \\ 0 & R_{22}(t) R_{22}^{-1}(s) \end{pmatrix},$$

$$39 \quad R(t) (I - P) R^{-1}(s) = \begin{pmatrix} R_{11}(t) R_{11}^{-1}(s) & R_{11}(t) (R^{-1}(s))_{12} \\ 0 & 0 \end{pmatrix},$$

42 where

$$44 \quad (R^{-1}(s))_{12} = \left[\int_s^\infty R_{11}^{-1}(\tau) B_{12}(\tau) R_{22}(\tau) d\tau \right] R_{22}^{-1}(s).$$

48 Because of (36) and (37), to verify (6) we need to get (exponentially decreasing) bounds for
 49 $\|R_{12}(t) R_{22}^{-1}(s)\|$ when $t \geq s$ and for $\|R_{11}(t) (R^{-1}(s))_{12}\|$ when $t \leq s$. We have

$$52 \quad \int_{-\infty}^t \|R_{22}(s) R_{22}^{-1}(s)\| ds \leq \int_{-\infty}^t K_2 e^{-\alpha_2(t-s)} ds \leq \frac{K_2}{\alpha_2}$$

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$$\int_t^\infty \|R_{11}(t)R_{11}^{-1}(s)\| ds \leq \int_t^\infty K_1 e^{-\alpha_1(s-t)} ds \leq \frac{K_1}{\alpha_1}.$$

7 Next

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$$\int_{-\infty}^t R_{12}(t)R_{22}^{-1}(s) ds = - \int_{-\infty}^t R_{11}(t) \int_t^\infty R_{11}^{-1}(\tau) B_{12}(\tau) R_{22}(\tau) d\tau R_{22}^{-1}(s) ds$$

14 and for $t \geq s$,15
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$$\|R_{12}(t)R_{22}^{-1}(s)\| \leq \kappa \int_t^\infty K_1 e^{-\alpha_1(\tau-t)} K_2 e^{-\alpha_2(\tau-s)} d\tau \leq \kappa \frac{K_1 K_2}{(\alpha_1 + \alpha_2)} e^{-\alpha_2(t-s)}. \quad (41)$$

21 Moreover,

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$$\begin{aligned} \int_{-\infty}^t \|R_{12}(t)R_{22}^{-1}(s)\| ds &\leq \kappa \frac{K_1 K_2}{(\alpha_1 + \alpha_2)} \int_{-\infty}^t e^{-\alpha_2(t-s)} ds \\ &\leq \kappa \frac{K_1 K_2}{\alpha_2(\alpha_1 + \alpha_2)}. \end{aligned} \quad (42)$$

30 Similarly,

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$$\int_t^\infty R_{11}(t)(R^{-1}(s))_{12} ds = \int_t^\infty \left[\int_s^\infty R_{11}^{-1}(\tau) B_{12}(\tau) R_{22}(\tau) d\tau \right] R_{22}^{-1}(s) ds$$

37 and for $s \geq t$,38
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$$\|R_{11}(t)R_{12}^{-1}(s)\| \leq \kappa \int_s^\infty K_1 e^{-\alpha_1(\tau-t)} K_2 e^{-\alpha_2(\tau-s)} d\tau \leq \kappa \frac{K_1 K_2}{(\alpha_1 + \alpha_2)} e^{-\alpha_1(s-t)}. \quad (43)$$

44 Moreover,

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$$\begin{aligned} \int_t^\infty \|R_{11}(t)(R^{-1}(s))_{12}\| ds &\leq \kappa \frac{K_1 K_2}{(\alpha_1 + \alpha_2)} \int_t^\infty e^{-\alpha_1(s-t)} ds \\ &\leq \kappa \frac{K_1 K_2}{\alpha_1(\alpha_1 + \alpha_2)}. \end{aligned}$$

53 Thus,

$$\int_{-\infty}^t \|R(t)PR^{-1}(s)\| ds \leq \frac{K_2}{\alpha_2} \sqrt{1 + \kappa^2 \frac{K_1^2}{(\alpha_1 + \alpha_2)^2}},$$

and

$$\int_t^\infty \|R(t)(I - P)R^{-1}(s)\| ds \leq \frac{K_1}{\alpha_1} \sqrt{1 + \kappa^2 \frac{K_2^2}{(\alpha_1 + \alpha_2)^2}}. \quad (44)$$

Now, from (41) and (43), we obtain (6) on \mathbb{R} . Moreover, using (42) and (44) in (40), we obtain (39). \square

Remark 25. If there ~~exist~~ $\alpha > 0$, $K \geq 1$ such that

$$\|R_{22}(t)R_{22}^{-1}(s)\| \cdot \|R_{11}(s)R_{11}^{-1}(t)\| \leq K e^{-\alpha(t-s)}, \quad t \geq s, \quad (45)$$

where $R_{11}(t_0) = I$ and $R_{22}(t_0) = I$ for some t_0 , then the system is exponentially separated [25]. The exponential separation condition (45) implies having integral separation between arbitrary diagonal elements of R_{11} and R_{22} . Clearly, (36) and (37) imply (45). In addition, (45) implies

$$\|R_{12}(t)\| \leq \kappa \int_t^\infty K e^{-\alpha(\tau-t)} d\tau \cdot \|R_{22}(t)\| \leq \kappa \frac{K}{\alpha} \cdot \|R_{22}(t)R_{22}^{-1}(t_0)\| \rightarrow 0, \quad \text{as } t \rightarrow +\infty \quad (46)$$

using (37). This means that we are employing a fundamental matrix solution $R(t)$ such that $R_{11}(t_0) = I$, $R_{22}(t_0) = I$, and $\lim_{t \rightarrow \infty} R_{12}(t) = 0$. If (45) holds and there ~~exist~~ $L_1 > 0$ and $\beta_1 > 0$ such that $\alpha > \beta_1$ and

$$\|R_{11}(s)R_{11}^{-1}(t)\| \geq L_1 e^{-\beta_1(t-s)}, \quad t \geq s,$$

then (37) holds. Similarly, if (45) holds and there ~~exist~~ $L_2 > 0$ and $\beta_2 > 0$ such that $\alpha > \beta_2$ and

$$\|R_{22}(t)R_{22}^{-1}(s)\| \geq L_2 e^{-\beta_2(t-s)}, \quad t \geq s,$$

then (36) holds.

Remark 26. For the triangular problem of which in Theorem 24, we can explicitly give the forward and backward stable subspaces. In fact, from the proof of Theorem 24, we have that

$$\mathcal{S}^+ = \text{span} \begin{pmatrix} R_{12}(0) \\ I_2 \end{pmatrix}, \quad \mathcal{S}^- = \text{span} \begin{pmatrix} I_1 \\ 0 \end{pmatrix},$$

where I_1 and I_2 are identity matrices of the same size as the blocks R_{11} and R_{22} , respectively, and $R_{12}(0)$ is defined from (38).

1 **5. Examples**2 In this section we elucidate the previous theoretical results with examples of **two-dimensional**
3 upper triangular systems.4 **1.** First, consider the following problem

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$$\dot{R} = B(t)R = \begin{pmatrix} \arctan(t) + \frac{t}{1+t^2} & \epsilon f(t) \\ 0 & -\frac{1}{2}(\arctan(t) + \frac{t}{1+t^2}) \end{pmatrix} R, \quad (47)$$

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11 with $\epsilon \geq 0$ and f bounded and continuous in \mathbb{R} . Notice that the diagonal elements B_{11} and B_{22} of the
12 coefficient matrix in (47) satisfy conditions (12) and (14) so that the scalar problems $\dot{R}_{ii} = B_{ii}(t)R_{ii}$,
13 for $i = 1, 2$, do not admit ED on \mathbb{R} . This implies that, for $\epsilon = 0$, the system (47) does not admit ED on
14 \mathbb{R} either. Then, we consider the case of $\epsilon > 0$ and try to establish if (47) has or not ED on \mathbb{R} . Observe
15 that (36)–(37) do not hold (regardless of ϵ) and Theorem 24 cannot be used to establish ED on \mathbb{R} .

16 The principal matrix solution Φ of (47) is given by

17
$$\begin{aligned} \Phi_{11}(t) &= e^{t \arctan(t)}; \\ \Phi_{12}(t) &= \epsilon e^{t \arctan(t)} \int_0^t (e^{-3/2s \arctan(s)} f(s)) ds; \\ \Phi_{21} &= 0, \quad \Phi_{22}(t) = e^{-\frac{t \arctan(t)}{2}}. \end{aligned}$$

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27 Hence, (ED-1) is satisfied, and in order to verify whether or not there is ED on the whole line we
28 need to check whether or not (ED-2) is satisfied. The forward and backward stable subspaces are
29 one-dimensional and can be computed explicitly. We are looking for two vectors $v = (v_1, v_2)^T$ and
30 $w = (w_1, w_2)^T$ such that

31
$$\lim_{t \rightarrow +\infty} \|\Phi(t)v\| = 0, \quad \lim_{t \rightarrow -\infty} \|\Phi(t)w\| = 0. \quad (48)$$

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35 The first of (48) rewrites as

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$$\Phi_{11}(t)v_1 + \Phi_{12}(t)v_2 \rightarrow 0, \quad \Phi_{22}(t)v_2 \rightarrow 0,$$

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40 and clearly the second requirement is verified for any v_2 . For v_1 , we need to take it so that

41
$$\Phi_{11}(t)v_1 + \Phi_{12}(t)v_2 = e^{t \arctan(t)} \left(v_1 + \epsilon \int_0^t (e^{-\frac{3}{2}s \arctan(s)} f(s)) ds v_2 \right) \rightarrow 0. \quad (49)$$

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46 So, we must choose (except for normalization of v_2):

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$$v_1 = - \lim_{t \rightarrow +\infty} \epsilon \int_0^t (e^{-\frac{3}{2}s \arctan(s)} f(s)) ds, \quad v_2 = 1.$$

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53 With this choice of v_1 , the expression in (49) goes to zero and $\mathcal{S}^+ = \text{Span}(v)$.

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1 If we repeat the same reasoning for the second limit in (48), we obtain
 2

$$3 w_1 = -\lim_{t \rightarrow -\infty} \epsilon \int_0^t (e^{-\frac{3}{2}s \arctan(s)} f(s)) ds, \quad w_2 = 1.$$

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7 Observe that $e^{-\frac{3}{2}s \arctan(s)}$ is an even function. We distinguish two cases.
 8

10 (a) If f is odd, $w_1 = v_1$ and $\mathcal{S}^+ = \mathcal{S}^-$ so that (47) does not admit ED on \mathbb{R} .
 11 (b) If f is even, $w_1 = -v_1$, $\mathcal{S}^+ \cap \mathcal{S}^- = \{0\}$ and (47) admits ED on \mathbb{R} .

13 **Remark 27.** Example (47) can be easily extended to the case of $B_{11} = -B_{22}$, and B_{11} such that
 14

$$15 \int_s^t B_{11}(\tau) d\tau \geq a(t-s), \quad t \geq s \geq 0,$$

$$16$$

$$17$$

$$18$$

$$19 \int_s^t B_{11}(\tau) d\tau \geq -a(t-s), \quad t \leq s \leq 0,$$

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$$21$$

$$22$$

23 where $a > 0$. For example, this can be verified for a function B_{11} which is odd, like the coefficient B_{11}
 24 in (47). In this class of problems, B_{11} is always leading to the unstable Lyapunov exponent (positive
 25 in forward time, negative in backward time), so that the principal matrix solution is a suitable initial
 26 condition both in forward and backward time (see Remark 2). Again, if f is even and $\epsilon \neq 0$, there is
 27 ED on \mathbb{R} , while if f is odd there is no ED for any ϵ . Notice that (36)–(37) do not hold; in fact, for
 28 this problem having or not ED depends in a subtle way on the term B_{12} .
 29

30 **2.** This is generalization of the example in Remark 2 and should be compared with the case in the last
 31 remark above. Consider $B_{11}(t) = -B_{22}(t)$, and assume that there is integral separation on the whole
 32 line so that
 33

$$34 \int_s^t (B_{11}(\tau) - B_{22}(\tau)) d\tau = 2 \int_s^t B_{11}(\tau) d\tau \geq a(t-s) - d$$

$$35$$

$$36$$

$$37$$

$$38$$

39 for $a > 0$, $d \geq 0$, and $t \geq s$. Then, the scalar problems $\dot{R}_{ii} = B_{ii}(t)R_{ii}$, $i = 1, 2$, have ED on the whole
 40 line, (36) and (37) hold and the problem has ED on \mathbb{R} for any ϵ and $f(t)$. If we want to compute
 41 the stable subspaces (forward and backward), then to obtain \mathcal{S}^+ we may use the principal matrix
 42 solution. To obtain \mathcal{S}^- , we cannot use the principal matrix solution and a different fundamental
 43 matrix must be employed, though in this example we actually have $\mathcal{S}^- = \text{Span}(e_1)$, see Remark 26.
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45 6. Conclusion

$$46$$

47 In this paper we have shown how techniques based on the SVD or QR decomposition of a funda-
 48 mental matrix solution can be used to determine whether or not a system has exponential dichotomy
 49 on the real line. An important consequence of our results is that for a large class of problems, namely
 50 those with integral separation on both lines or more generally with stable Lyapunov exponents on
 51 both half-lines, the existence of exponential dichotomy may be determined from information on finite
 52 intervals up to exponentially small perturbations. This enlarges the class of problems for which finite
 53 interval computations are sufficient to determine exponential dichotomy from those that are asymptotically
 54 constant, asymptotically periodic, or almost periodic [27]. Another important benefit of our

1 theoretical results is that they lend themselves to practical algorithmic procedures to verify whether
 2 or not a system has exponential dichotomy, as it will be reported elsewhere.
 3

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