Matrix Berlekamp-Massey

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1 Linearly recurrent matrix sequences

First thing: what are linearly recurrent sequences and minimal generators, when talking about matrix sequences? The definition is essentially the same as in the scalar case. There is a similar notion of minimal generator of a sequence (which is a matrix): it is related to the denominator in some irreducible fraction description of the generating series of the sequence.

We consider the following definition which extends the scalar case. It can be found in [3, Sec. 3], and in [7, Def. 4.2].

Definition 1.1. Let $S = (S_k)_{k \in \mathbb{Z}_{\geq 0}} \subset \mathbb{K}^{m \times n}$. A vector $\mathbf{p} = \sum_{0 \leq k \leq d} p_k X^k \in \mathbb{K}[X]^{1 \times m}$ of degree at most d is said to be a (linear recurrence) relation for S if $\sum_{k=0}^{d} p_k S_{\delta+k} = 0$ for all $\delta \geq 0$. Then, S is said to be linearly recurrent if there exists a nontrivial relation for S.

Note: hereafter we never use the word *generator* to refer to relations like those in this definition. This word *generator* will be reserved to matrices which generate modules of relations.

The set of relations for S is a $\mathbb{K}[X]$ -submodule of $\mathbb{K}[X]^{1\times m}$, which has rank m if S is linearly recurrent. This is showed in [4, Fact 1] but only for sequences having a scalar recurrence relation. Let us now link this with the property of being linearly recurrent as in the above definition.

Lemma 1.2. The sequence S is linearly recurrent if and only if there exists a polynomial $P(X) = \sum_{0 \le k \le d} p_k X^k \in \mathbb{K}[X]$ such that $\sum_{k=0}^d p_k S_{\delta+k} = 0$ for all $\delta \ge 0$.

Proof. I cannot find this result in the literature..! (but we don't care since in our case the sequence obviously admits a scalar relation)

Then, the fact that the relation module has rank m is straightforward: since the sequence is linearly recurrent, it has a scalar recurrence polynomial P(X), hence for each i the vector $[0 \cdots 0 P(X) 0 \cdots 0]$ with P(X) at position i is a relation for S.

A matrix generator for the sequence is a matrix whose rows form a generating set for the module of relations; it is said to be

• minimal if the matrix is reduced [9, 2];

- ordered weak Popov if the matrix is in weak Popov form [5] with pivots on the diagonal;
- canonical if the matrix is in Popov form [6, 2].

In this context, an important quantity related to the sequence is the determinantal degree $\deg(\det(\mathbf{P}))$, invariant for all generators $\mathbf{P} \in \mathbb{K}[X]^{m \times m}$. Hereafter, we denote it by $\Delta(\mathcal{S})$.

Lemma 1.3. Consider a matrix sequence $S = (S_k)_{k \in \mathbb{Z}_{\geq 0}} \subset \mathbb{K}^{m \times n}$ and its generating series $\mathbf{S} = \sum_{k \geq 0} S_k / X^{k+1} \in \mathbb{K}[X^{-1}]^{m \times n}$. Then, a vector $\mathbf{p} \in \mathbb{K}[X]^{1 \times m}$ is a relation for S if and only if the entries of $\mathbf{q} = \mathbf{p}\mathbf{S}$ are in $\mathbb{K}[X]$; furthermore, in this case, $\deg(\mathbf{q}) < \deg(\mathbf{p})$.

Proof. Let $\mathbf{p} = \sum_{0 \le k \le d} p_k X^k$. For $\delta \ge 0$, the coefficient of \mathbf{q} of degree $-\delta - 1 < 0$ is $\sum_{0 \le k \le d} p_k S_{k+\delta}$. Hence the equivalence, by definition of a relation. The degree comparison is clear since \mathbf{S} has only terms of (strictly) negative degree.

Corollary 1.4. A matrix sequence $S = (S_k)_{k \in \mathbb{Z}_{\geqslant 0}} \subset \mathbb{K}^{m \times n}$ is linearly recurrent if and only if its generating series $\mathbf{S} = \sum_{k \geqslant 0} S_k / X^{k+1} \in \mathbb{K}[\![X^{-1}]\!]^{m \times n}$ can be written as a matrix fraction $\mathbf{S} = \mathbf{P}^{-1}\mathbf{Q}$ where $\mathbf{P} \in \mathbb{K}[X]^{m \times m}$ is nonsingular and $\mathbf{Q} \in \mathbb{K}[X]^{m \times n}$. In this case,

- we have $rdeg(\mathbf{Q}) < rdeg(\mathbf{P})$ and $deg(det(\mathbf{P})) \geqslant \Delta(\mathcal{S})$,
- **P** is a matrix generator of S if and only if the fraction is irreducible (i.e. $\mathbf{UP} + \mathbf{VQ} = \mathbf{I}$ for some \mathbf{U}, \mathbf{V}),
- **P** is a matrix generator of S if and only if $deg(det(\mathbf{P})) = \Delta(S)$.

For more details:

- [8, Sec. 1] when the sequence is of the form $S = (\mathbf{U}\mathbf{A}^k\mathbf{V})_k$. Note that in this case the generating series can be written $\mathbf{S} = \mathbf{U}(X\mathbf{I} \mathbf{A})\mathbf{V}$. Link with so-called realizations from control theory [2]...
- [7, Chap. 4] has things related to Hankel matrices (but it is extremely detailed, including many properties which are actually about polynomial matrices and completely independent of the "linear recurrence" context)

2 Computing minimal matrix generators

Now, we focus on the following algorithmic problem: we are given a linearly recurrent sequence and we want to find a matrix generator. If we want our algorithm to run efficiently (or simply, in finite time), we cannot access infinitely many terms of the sequence. We therefore ask for an additional input, which one often has when considering a sequence coming from some application: a (finite) bound on the degree of a minimal matrix generator. Note that a bound on the determinantal degree $\Delta(\mathcal{S})$ is sufficient since any minimal matrix generator will have degree less than this; yet better bounds can be available and will imply better efficiency.

In short, we consider the following problem.

We now show how the additional information of d allows us to find a matrix generator by considering only a small chunk of the sequence, rather than all its terms.

Problem 1 - *Minimal matrix generator*

Input:

- sequence $S = (S_k)_k \subset \mathbb{K}^{m \times n}$,
- degree bound $d \in \mathbb{Z}_{\geq 0}$.

Assumptions:

- the sequence S is linearly recurrent,
- any minimal matrix generator of S has degree at most d.

Output: a minimal matrix generator for S.

The fast computation of matrix generators is usually handled via algorithms for computing minimal approximant bases [8, 7, 1]. The next result gives the main idea behind this approach. This result is similar to [7, Thm. 4.7, 4.8, 4.9, 4.10], but in some sense the reversal is on the input sequence rather than on the output matrix generator (and also this section 4.2 of [7] provides many more details related to the mechanisms and output properties in the approximant basis algorithm, which we do not consider here).

Theorem 2.1. Let $S = (S_k)_k \subset \mathbb{K}^{m \times n}$ be a linearly recurrent sequence, let $d \in \mathbb{Z}_{\geq 0}$, and let

$$\mathbf{F} = \begin{bmatrix} \sum_{0 \le k \le 2d} S_k X^{2d-k} \\ -\mathbf{I}_n \end{bmatrix} \in \mathbb{K}[X]^{(m+n) \times n}.$$

Let further $\mathbf{p} \in \mathbb{K}[X]^{1 \times m}$ with $\deg(\mathbf{p}) \leqslant d$. Then, \mathbf{p} is a relation for \mathcal{S} if and only if there exists a vector $\mathbf{r} \in \mathbb{K}[X]^{1 \times n}$ such that $\deg(\mathbf{r}) < \deg(\mathbf{p})$ and $[\mathbf{p} \ \mathbf{r}]$ is an approximant for \mathbf{F} at order 2d + 1.

As a corollary,

Proof. From Lemma 1.3, if **p** is a relation then $\mathbf{q} = \mathbf{p}\mathbf{S}$ has polynomial entries, where $\mathbf{S} = \sum_{k \geq 0} S_k X^{-k-1}$. Then, the vector $\mathbf{r} = -\mathbf{p}(\sum_{k>2d} S_k X^{2d-k})$ has polynomial entries, has degree less than $\deg(\mathbf{p})$, and satisfies $[\mathbf{p} \ \mathbf{r}]\mathbf{F} = \mathbf{q}X^{2d+1}$.

Conversely, assume that there exists some $\mathbf{r} \in \mathbb{K}[X]^{1 \times n}$ such that $\deg(\mathbf{r}) < \deg(\mathbf{p})$ and $[\mathbf{p} \ \mathbf{r}]\mathbf{F} = 0 \mod X^{2d+1}$.

hence **p** has polynomial entries as well and $[\mathbf{p} \ \mathbf{r}]\mathbf{F} = 0 \mod X^{2d+1}$. Since **pS** has polynomial entries, **r** has polynomial entries; and since all terms of $\sum_{k>2d} S_k X^{2d-k}$ have negative degree, we have $\deg(\mathbf{r}) < \deg(\mathbf{p})$.

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