

Decay properties of certain matrix power series encountered in stochastic processes

Beatrice Meini

joint work with D. Bini and V. Ramaswami

Structured Linear Algebra Problems: Analysis, Algorithms, and
Applications
Cortona, September 15–19, 2008



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- ▶ Let A_n , for $n \geq -1$, be $k \times k$ nonnegative matrices such that $\sum_{n=-1}^{\infty} A_n$ is stochastic;



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Our goal: To analyze “decay properties” of $G(w)$ (more details will follow in the next slides)



Motivation

Consider a Markov Renewal Process (MRP) of M/G/1-type with levels l_0, l_1, l_2, \dots , defined by the kernel

$$K(x) = \begin{bmatrix} \tilde{B}_0(x) & \tilde{B}_1(x) & \tilde{B}_2(x) & \tilde{B}_3(x) & \dots \\ \tilde{A}_{-1}(x) & \tilde{A}_0(x) & \tilde{A}_1(x) & \tilde{A}_2(x) & \dots \\ & \tilde{A}_{-1}(x) & \tilde{A}_0(x) & \tilde{A}_1(x) & \ddots \\ & & \tilde{A}_{-1}(x) & \tilde{A}_0(x) & \ddots \\ 0 & & & & \ddots & \ddots \end{bmatrix}, \quad x \geq 0,$$

where

$$\tilde{A}_k(x) = P\{X_{n+1} \in l_{i+k}, \tau_{n+1} - \tau_n \leq x \mid X_0, \dots, X_n, \tau_0, \dots, \tau_n, X_n \in l_i\},$$

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Motivation

Let

$$\tilde{G}(n, x) = P\{\text{in } n \text{ steps and in time } \leq x \text{ the system goes from level } \ell_1 \text{ to level } \ell_0\}$$

Question: How fast does $\tilde{G}(n, x)$ go to zero, as $n \rightarrow \infty$?

Idea: Through Laplace-Stiltjes transforms reduce the MRP to an M/G/1-type Markov chain



Motivation

The M/G/1-type Markov chain is defined by the transition probability matrix

$$\begin{bmatrix} B_0 & B_1 & B_2 & B_3 & \dots \\ A_{-1} & A_0 & A_1 & A_2 & \dots \\ & A_{-1} & A_0 & A_1 & \ddots \\ & & A_{-1} & A_0 & \ddots \\ 0 & & & \ddots & \ddots \end{bmatrix}.$$

Let $G_n = P\{\text{in } n \text{ steps we go from level } \ell_1 \text{ to level } \ell_0\}$

Question: How fast does G_n go to zero, as $n \rightarrow \infty$?

Back to $G(w)$

Recall that $G(w)$ solves

$$X = w \sum_{n=-1}^{\infty} A_n X^{n+1}.$$

Known facts:

1. $G(w)$ is analytic $|w| < 1$, convergent for $|w| = 1$
2. $G(w) = \sum_{n=0}^{\infty} w^n G_n$
3. $G_n \geq 0$ for any $n \geq 0$, and G_n is the sought probability.

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Assumptions:

- ▶ $\sum_{i=-1}^{\infty} A_i$ is irreducible;
- ▶ the convergence radius of $A(z) = \sum_{i=-1}^{\infty} z^{i+1} A_i$ is $r_a > 1$.



Idea

To apply to $G(w)$ the:

Theorem (Cauchy estimate)

Let R be the convergence radius of the series $f(w) = \sum_{j=0}^{\infty} w^j f_j$.
Then, for any $0 < r < R$ and for any $j \geq 0$ one has

$$|f_j| \leq \frac{\mu(r)}{r^j}$$

where $\mu(r) = \sup_{|w|=r} |f(w)|$.

Problem: to determine the convergence radius of $G(w)$



A scalar example: Poisson distribution

Let $\lambda > 0$ and

$$A_i = e^{-\lambda} \frac{\lambda^{i+1}}{(i+1)!}, \quad i \geq -1.$$

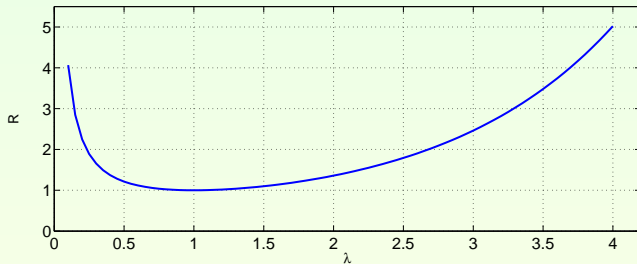
One has

$$G(w) = \sum_{n=1}^{\infty} w^n e^{-n(\lambda-1)} \lambda^{n-1}$$

The convergence radius is $R = (\lambda e^{-(\lambda-1)})^{-1}$



Convergence radius R in function of λ



How can we estimate R without knowing $G(w)$?

Scalar case

Let A_i be scalar numbers, and $A(z) = \sum_{i=-1}^{\infty} z^{i+1} A_i$. We look for conditions on w so that the equation $z = wA(z)$ has a solution in $(0, r_a)$.

The following properties hold:

- ▶ Since $A_i \geq 0$, the function $A(z)$ is convex and increasing in $(0, r_a)$;

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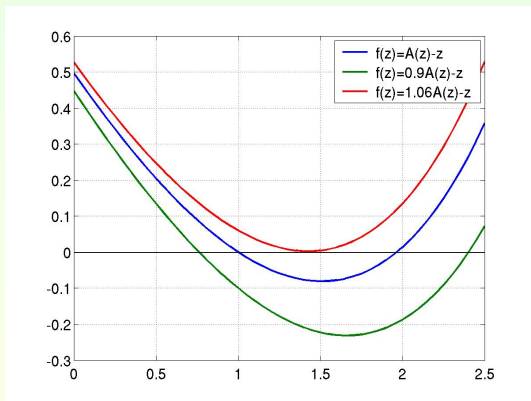
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The convergence radius R is the superior extremum of the $w \in (0, r_a)$ such that $z = wA(z)$ has a solution in $(0, r_a)$

Scalar case



Here $R = 1.06$

Convergence radius in the scalar case

The convergence radius of $G(w)$ is

$$R = \frac{1}{A'(\sigma)}$$

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Can we extend these properties to the block case?

General case

Let μ_w and \mathbf{u}_w be the Perron eigenvalue/eigenvector of $G(w)$.



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From

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one has

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Therefore, μ_w solves the *scalar equation*

$$z = w\rho(A(z)). \quad (1)$$

Intuition: the convergence radius is the superior extremum of the $w \in (0, r_a)$ such that (1) has a solution

Some properties

Let $\theta(z)$ be the spectral radius of $A(z) = \sum_{i=-1}^{\infty} z^{i+1} A_i$.

The following properties hold [Gail,Hantler,Taylor 95]:

- ▶ $\theta(z)$ is a real analytic function in $(0, r_a)$;
- ▶ $\theta(z)$ is strictly increasing in $(0, r_a)$;
- ▶ the function $\log \theta(e^t)$ is convex and increasing in $(-\infty, \log r_a)$.



Main result

Theorem

Let $\theta(z) = \rho(A(z))$. The following properties hold:

1. the equation

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2. the equation $z = w\theta(z)$ has a solution in $(0, r_a)$ for any $0 < w \leq 1/\theta'(\sigma)$;
3. the convergence radius of $G(w)$ is $R = 1/\theta'(\sigma) = \sigma/\theta(\sigma)$.



Fixed point iteration

Goal: Compute the solution $\sigma \in (0, r_a)$ of $z = \frac{\theta(z)}{\theta'(z)}$.

Set, say, $z_0 = 1$ and compute

$$z_{k+1} = \frac{\theta(z_k)}{\theta'(z_k)}, \quad k \geq 0.$$



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- ▶ $\theta'(z_k)$?

Computation of $\theta'(z)$

1. There exist vectors $\mathbf{u}(z)$ and $\mathbf{v}(z)$, analytic for $0 < z < r_a$, such that

$$\mathbf{u}(z)^T A(z) = \theta(z) \mathbf{u}(z)^T, \quad A(z) \mathbf{v}(z) = \theta(z) \mathbf{v}(z)$$



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2. Taking derivative in $A(z) \mathbf{v}(z) = \theta(z) \mathbf{v}(z)$ and multiplying the result on the left by $\mathbf{u}(z)^T$, yields

$$\theta'(z) = \frac{\mathbf{u}(z)^T A'(z) \mathbf{v}(z)}{\mathbf{u}(z)^T \mathbf{v}(z)}.$$

Newton's method

Newton's method applied to the function $g(z) = \theta(z) - z\theta'(z)$, yields the sequence

$$z_{k+1} = z_k + \frac{\theta(z_k) - z_k\theta'(z_k)}{z_k\theta''(z_k)}, \quad k \geq 0,$$

starting, say, from $z_0 = 1$.

How compute $\theta''(z_k)$?



Computation of $\theta''(z)$

Differentiate twice $A(z)\mathbf{v}(z) = \theta(z)\mathbf{v}(z)$, and multiply the result on the left by $\mathbf{u}(z)^T$.

We get

$$\theta''(z) = \frac{\mathbf{u}(z)^T (A''(z)\mathbf{v}(z) + 2A'(z)\mathbf{v}'(z) - 2\theta'(z)\mathbf{v}'(z))}{\mathbf{u}(z)^T \mathbf{v}(z)},$$

where $\mathbf{v}'(z)$ solves the singular system

$$(A(z) - \theta(z)I)\mathbf{v}'(z) = -(A'(z) - \theta'(z))\mathbf{v}(z).$$

Convergence

- ▶ We expect a local linear convergence for the fixed point iteration, and a local quadratic convergence for Newton's method.
- ▶ A detailed convergence analysis seems difficult to be performed.
- ▶ Experimentally, we observed that the convergence of the fixed point iteration obtained starting with $z_0 = 1$ is monotonic.



Numerical experiments

Consider the case

$$A_{-1} = \begin{bmatrix} 0 & p \\ 0 & 0 \end{bmatrix}, \quad A_0 = \begin{bmatrix} 0.8 - p & 0 \\ 1 & 0 \end{bmatrix}, \quad A_1 = \begin{bmatrix} 0 & 0.2 \\ 0 & 0 \end{bmatrix},$$

and $A_i = 0$ for $i > 1$, with $0 < p < 0.8$. This problem is positive recurrent if $p > 0.2$, and null recurrent if $p = 0.2$.

We implemented the algorithms in Matlab, and considered the cases obtained with $p = 0.1 \cdot k$, $k = 1, 2, \dots, 7$.

The iterations have been halted if $|z_{k+1} - z_k| < 10^{-10}$.



Numerical results

p	0.1	0.2	0.3	0.4	0.5	0.6	0.7
Newton	5	1	5	5	6	6	6
F.P.I.	81	1	52	47	43	39	36
R	1.0135	1	1.0068	1.0222	1.0424	1.0657	1.0911

Table: Number of iterations and value of the radius R

Further developments

- ▶ Convergence analysis of the algorithms
- ▶ Design of algorithms for computing an arbitrary number of matrices G_n
- ▶ Probabilistic interpretation of the formula $R = 1/\theta'(\sigma)$.

