

A semilinear hyperbolic system with space-dependent and nonlinear damping (part I)

Debora Amadori

University of L'Aquila (Italy)



VIII Partial differential equations, optimal design and numerics

Benasque, August 20, 2019

Joint work with Fatima Aqel, Edda Dal Santo (L'Aquila)

The problem

$$\begin{cases} \partial_t \rho + \partial_x J = 0 \\ \partial_t J + \partial_x \rho = -2k(x)g(J) \end{cases} \quad x \in I = (0, 1), \quad t > 0$$

- Initial conditions: $(\rho_0, J_0) \in L^\infty(I)$
- Boundary conditions: $J(0, t) = J(1, t) = J_b \in \mathbb{R}$

The problem

$$\begin{cases} \partial_t \rho + \partial_x J = 0 \\ \partial_t J + \partial_x \rho = -2k(x)g(J) \end{cases} \quad x \in I = (0, 1), \quad t > 0$$

- Initial conditions: $(\rho_0, J_0) \in L^\infty(I)$
- Boundary conditions: $J(0, t) = J(1, t) = J_b \in \mathbb{R}$
- Damping term: $k(x) \geq 0, k \in L^1(I); \quad g \in \mathbf{C}^1(\mathbb{R}), \quad g(J)J \geq 0$

The problem

$$\begin{cases} \partial_t \rho + \partial_x J = 0 \\ \partial_t J + \partial_x \rho = -2k(x)g(J) \end{cases} \quad x \in I = (0, 1), \quad t > 0$$

- Initial conditions: $(\rho_0, J_0) \in L^\infty(I)$
- Boundary conditions: $J(0, t) = J(1, t) = J_b \in \mathbb{R}$
- Damping term: $k(x) \geq 0, k \in L^1(I); \quad g \in \mathbf{C}^1(\mathbb{R}), \quad g(J)J \geq 0$

Relation with the semilinear wave equation:

$$\begin{cases} -u_x = \rho \\ u_t = J \end{cases} \quad \Rightarrow \quad \boxed{\partial_{tt}u - \partial_{xx}u + 2k(x)g(\partial_t u) = 0}$$

The problem

$$\begin{cases} \partial_t \rho + \partial_x J = 0 \\ \partial_t J + \partial_x \rho = -2k(x)g(J) \end{cases} \quad x \in I = (0, 1), \quad t > 0$$

- Initial conditions: $(\rho_0, J_0) \in L^\infty(I)$
- Boundary conditions: $J(0, t) = J(1, t) = J_b \in \mathbb{R}$
- Damping term: $k(x) \geq 0, k \in L^1(I); \quad g \in \mathbf{C}^1(\mathbb{R}), \quad g(J)J \geq 0$

Relation with the semilinear wave equation:

$$\begin{cases} -u_x = \rho \\ u_t = J \end{cases} \quad \Rightarrow \quad \boxed{\partial_{tt} u - \partial_{xx} u + 2k(x)g(\partial_t u) = 0}$$

- Fixed string at both ends:

$$u(0, t) = u(1, t) = 0 \quad \Leftrightarrow \quad J_b = 0, \quad \int_I \rho_0(x) dx = 0.$$

Stationary equations

$$\partial_x J = 0, \quad \partial_x \rho = -2k(x)g(J)$$

- Stationary solution:

$$J(x) = J_b, \quad \rho(x) = -2g(J) \int_0^x k(y) dy + C,$$

C uniquely determined by $\int_I \rho_0$.

Stationary equations

$$\partial_x J = 0, \quad \partial_x \rho = -2k(x)g(J)$$

- Stationary solution:

$$J(x) = J_b, \quad \rho(x) = -2g(J) \int_0^x k(y) dy + C,$$

C uniquely determined by $\int_I \rho_0$.

- Without loss of generality:

$$J_b = 0, \quad \int_I \rho_0 dx = 0, \quad J(x) = \rho(x) = 0 \quad \forall x$$

General target

- (A) to study the decay properties of the solutions as $t \rightarrow \infty$;
- (B) to provide approximations with good accuracy for large t .

What we did... and what we do

- For the Cauchy problem...

Rigorous L^1 error estimates for suitable approximations of

$$\partial_t \rho + \partial_x J = 0, \quad \partial_t J + \partial_x \rho = -2k(x)g(\rho, J)$$

based on Bressan-Liu-Yang [functional](#) for hyperbolic systems of conservation laws

[A. - Gosse, Springerbriefs 2015]

What we did... and what we do

- For the Cauchy problem...

Rigorous L^1 error estimates for suitable approximations of

$$\partial_t \rho + \partial_x J = 0, \quad \partial_t J + \partial_x \rho = -2k(x)g(\rho, J)$$

based on Bressan-Liu-Yang **functional** for hyperbolic systems of conservation laws
[A. - Gosse, Springerbriefs 2015]

- ... and for the Initial-Boundary Value problem (IBVP)

- No direct approach using the BLY functional, due to the reflection of waves at the two boundaries
- An alternative approach inspired by the **probabilistic interpretation** of the damped wave equation (using **matrix analysis**)

What we did... and what we do

- For the Cauchy problem...

Rigorous L^1 error estimates for suitable approximations of

$$\partial_t \rho + \partial_x J = 0, \quad \partial_t J + \partial_x \rho = -2k(x)g(\rho, J)$$

based on Bressan-Liu-Yang **functional** for hyperbolic systems of conservation laws
[A. - Gosse, Springerbriefs 2015]

- ... and for the Initial-Boundary Value problem (IBVP)

- No direct approach using the BLY functional, due to the reflection of waves at the two boundaries
- An alternative approach inspired by the **probabilistic interpretation** of the damped wave equation (using **matrix analysis**)

\rightsquigarrow Decay in L^∞ as $t \rightarrow \infty$

[Haraux (2009): Decay in L^p , $2 \leq p \leq \infty$, sufficiently regular data. Chitour, Marx, Prieur (2019)]

A probabilistic interpretation of telegrapher's equation

Goldstein, Kac (1956)

- Particles moving either to the left or to the right, speed = ± 1 .
Time step τ , space step δ
- $\alpha(x, t)$: probability that a particle at (x, t) arrived from the left,
 $\beta(x, t)$: probability that a particle at (x, t) arrived from the right

A probabilistic interpretation of telegrapher's equation

Goldstein, Kac (1956)

- Particles moving either to the left or to the right, **speed** = ± 1 .
Time step τ , space step δ
- $\alpha(x, t)$: probability that a particle at (x, t) arrived from the **left**,
 $\beta(x, t)$: probability that a particle at (x, t) arrived from the **right**
- λ : **fixed** rate of reversal, $1 - \lambda\tau > 0$:

$$\alpha(x, t) = (1 - \lambda\tau)\alpha(x - \delta, t - \tau) + \lambda\tau \beta(x - \delta, t - \tau)$$

$$\beta(x, t) = (1 - \lambda\tau)\beta(x + \delta, t - \tau) + \lambda\tau \alpha(x + \delta, t - \tau)$$

A probabilistic interpretation of telegrapher's equation

Goldstein, Kac (1956)

- Particles moving either to the left or to the right, **speed** = ± 1 .
Time step τ , space step δ
- $\alpha(x, t)$: probability that a particle at (x, t) arrived from the **left**,
 $\beta(x, t)$: probability that a particle at (x, t) arrived from the **right**
- λ : **fixed** rate of reversal, $1 - \lambda\tau > 0$:

$$\alpha(x, t) = (1 - \lambda\tau)\alpha(x - \delta, t - \tau) + \lambda\tau\beta(x - \delta, t - \tau)$$

$$\beta(x, t) = (1 - \lambda\tau)\beta(x + \delta, t - \tau) + \lambda\tau\alpha(x + \delta, t - \tau)$$

- As $\tau = \delta \rightarrow 0$: a **linear** system

$$\partial_t\alpha + \partial_x\alpha = -\lambda\alpha + \lambda\beta$$

$$\partial_t\beta - \partial_x\beta = \lambda\alpha - \lambda\beta$$

A probabilistic interpretation for the telegrapher's equation

$$\partial_t(\alpha + \beta) + \partial_x(\alpha - \beta) = 0$$

$$\partial_t(\alpha - \beta) + \partial_x(\alpha + \beta) = -2\lambda(\alpha - \beta)$$

- For $v = \alpha + \beta$:

$$\partial_t^2 v - \partial_x^2 v + 2\lambda \cdot (\partial_t v) = 0.$$

A probabilistic interpretation for the telegrapher's equation

$$\partial_t(\alpha + \beta) + \partial_x(\alpha - \beta) = 0$$

$$\partial_t(\alpha - \beta) + \partial_x(\alpha + \beta) = -2\lambda(\alpha - \beta)$$

- For $v = \alpha + \beta$:

$$\partial_t^2 v - \partial_x^2 v + 2\lambda \cdot (\partial_t v) = 0.$$

Remark:

The damping term is linear.

In our case, the rate of reversal λ is not constant but depends on x and on the solution through $g'(J)$.

Question: Is there a counterpart of this derivation for nonlinear damping?

A probabilistic interpretation for the telegrapher's equation

$$\partial_t(\alpha + \beta) + \partial_x(\alpha - \beta) = 0$$

$$\partial_t(\alpha - \beta) + \partial_x(\alpha + \beta) = -2\lambda(\alpha - \beta)$$

- For $v = \alpha + \beta$:

$$\partial_t^2 v - \partial_x^2 v + 2\lambda \cdot (\partial_t v) = 0.$$

Remark:

The damping term is linear.

In our case, the rate of reversal λ is not constant but depends on x and on the solution through $g'(J)$.

Question: Is there a counterpart of this derivation for nonlinear damping?

[Katsoulakis-Tzavaras JSP 1999]

The "Well Balanced" (WB) approximation

Let $N \in 2\mathbb{N}$, $\Delta x = \Delta t = 1/N$, $x_j = j\Delta x$, $j = 0, \dots, N$ and set

$$\mu_N = \sum_{j=1}^{2N-1} \left(\int_{x_{j-1}}^{x_j} k(x) dx \right) \delta_{\{x_j\}}$$

Consider

$$\begin{cases} \partial_t \rho + \partial_x J & = 0, \\ \partial_t J + \partial_x \rho + 2g(J)\mu_N & = 0, \end{cases}$$

with

- initial data $(\rho_0^{\Delta x}, J_0^{\Delta x})$ piecewise constant, being constant on each cell
- boundary conditions $J^{\Delta x}(0, t) = J^{\Delta x}(1, t) = 0$.

The "Well Balanced" (WB) approximation

Let $N \in 2\mathbb{N}$, $\Delta x = \Delta t = 1/N$, $x_j = j\Delta x$, $j = 0, \dots, N$ and set

$$\mu_N = \sum_{j=1}^{2N-1} \left(\int_{x_{j-1}}^{x_j} k(x) dx \right) \delta_{\{x_j\}}$$

Consider

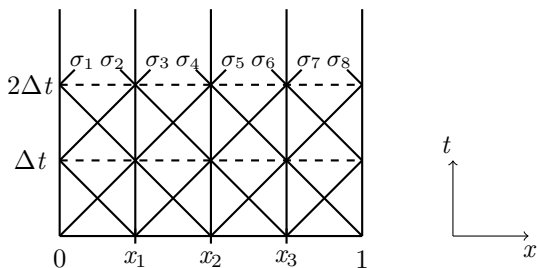
$$\begin{cases} \partial_t \rho + \partial_x J & = 0, \\ \partial_t J + \partial_x \rho + 2g(J)\mu_N & = 0, \end{cases}$$

with

- initial data $(\rho_0^{\Delta x}, J_0^{\Delta x})$ piecewise constant, being constant on each cell
- boundary conditions $J^{\Delta x}(0, t) = J^{\Delta x}(1, t) = 0$.

An approximate solution $(\rho^{\Delta x}, J^{\Delta x})(x, t)$ is an **exact** solution to the initial-boundary value problem above.

Waves in the WB approximation



WB scheme, $N = 4$. The segments with speed $0, \pm 1$ correspond to the location of the discontinuities.

The size of waves:

± 1 -waves:

$$\sigma_{\pm 1} = \Delta J = \pm \Delta \rho,$$

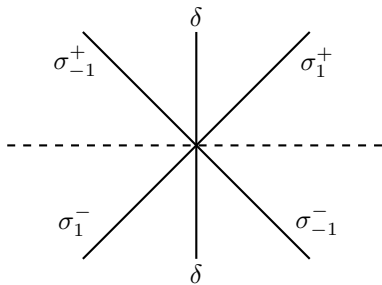
0 -waves:

$$\delta = \int_{x_{j-1}}^{x_j} k(y) dy$$

The vector size:

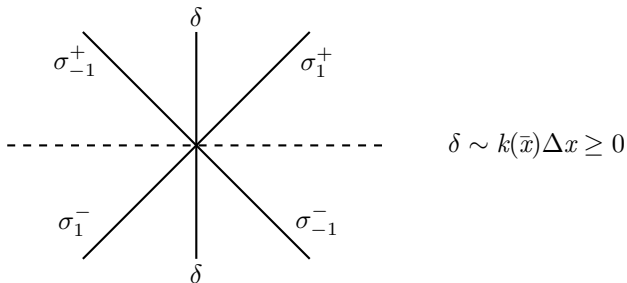
$$\sigma(t) = (\sigma_1, \dots, \sigma_{2N})(t)$$

Interactions



$$\delta \sim k(\bar{x})\Delta x \geq 0$$

Interactions



Wave sizes change according to:

$$\begin{pmatrix} \sigma_{-1}^+ \\ \sigma_1^- \end{pmatrix} = \begin{pmatrix} 1-c & c \\ c & 1-c \end{pmatrix} \begin{pmatrix} \sigma_{-1}^- \\ \sigma_1^+ \end{pmatrix} \quad c \sim g'(J)\delta.$$

\rightsquigarrow A 2×2 doubly stochastic matrix if $g' \geq 0$

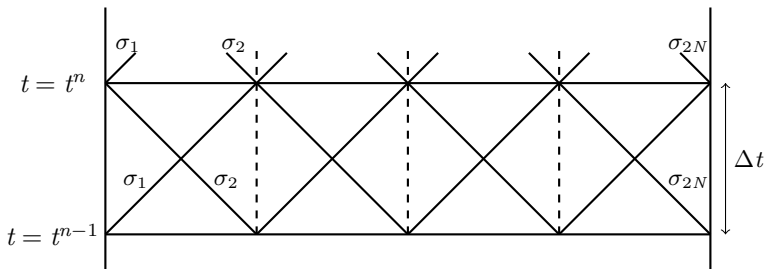
The transition matrix

Let

$$\boldsymbol{\sigma}(t) = (\sigma_1, \dots, \sigma_{2N}) \in \mathbb{R}^{2N}, \quad N = (\Delta x)^{-1}$$

be the vector of sizes of waves at time t , in increasing space order. Then for $t^n = n\Delta t$, $n \geq 1$:

$$\boldsymbol{\sigma}(t^n+) = B(\mathbf{c}^n) \boldsymbol{\sigma}(t^{n-1}+)$$



The transition matrix

Let

$$\boldsymbol{\sigma}(t) = (\sigma_1, \dots, \sigma_{2N}) \in \mathbb{R}^{2N}, \quad N = (\Delta x)^{-1}$$

be the vector of sizes of waves at time t , in increasing space order. Then for $t^n = n\Delta t$, $n \geq 1$:

$$\boldsymbol{\sigma}(t^n+) = B(\mathbf{c}^n) \boldsymbol{\sigma}(t^{n-1}+)$$

$$\mathbf{c}^n = (c_1^n, \dots, c_{N-1}^n) \in \mathbb{R}^{N-1} \quad c_j^n \sim g'(J_j^n) \delta_j$$

- $B(\mathbf{c}) \in M_{2N}(\mathbb{R})$ is **doubly stochastic**. The eigenvalues λ_i satisfy $|\lambda_i| \leq 1$.

The transition matrix

Let

$$\boldsymbol{\sigma}(t) = (\sigma_1, \dots, \sigma_{2N}) \in \mathbb{R}^{2N}, \quad N = (\Delta x)^{-1}$$

be the vector of sizes of waves at time t , in increasing space order. Then for $t^n = n\Delta t$, $n \geq 1$:

$$\boldsymbol{\sigma}(t^n+) = B(\mathbf{c}^n) \boldsymbol{\sigma}(t^{n-1}+)$$

$$\mathbf{c}^n = (c_1^n, \dots, c_{N-1}^n) \in \mathbb{R}^{N-1} \quad c_j^n \sim g'(J_j^n) \delta_j$$

- $B(\mathbf{c}) \in M_{2N}(\mathbb{R})$ is **doubly stochastic**. The eigenvalues λ_i satisfy $|\lambda_i| \leq 1$.
- The coefficients \mathbf{c}^n depend on time when g is non linear, and depend on space if $k(x)$ is not constant.

The transition matrix

Let

$$\boldsymbol{\sigma}(t) = (\sigma_1, \dots, \sigma_{2N}) \in \mathbb{R}^{2N}, \quad N = (\Delta x)^{-1}$$

be the vector of sizes of waves at time t , in increasing space order. Then for $t^n = n\Delta t$, $n \geq 1$:

$$\boldsymbol{\sigma}(t^n+) = B(\mathbf{c}^n) \boldsymbol{\sigma}(t^{n-1}+)$$

$$\mathbf{c}^n = (c_1^n, \dots, c_{N-1}^n) \in \mathbb{R}^{N-1} \quad c_j^n \sim g'(J_j^n) \delta_j$$

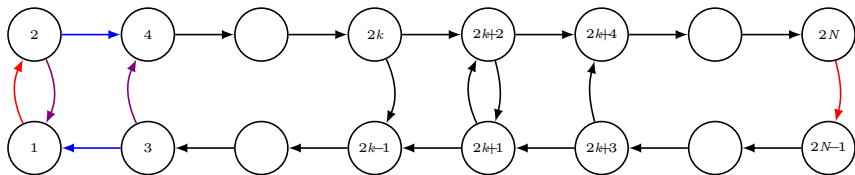
- $B(\mathbf{c}) \in M_{2N}(\mathbb{R})$ is **doubly stochastic**. The eigenvalues λ_i satisfy $|\lambda_i| \leq 1$.
- The coefficients \mathbf{c}^n depend on time when g is non linear, and depend on space if $k(x)$ is not constant.

GOAL:

Determine **spectral properties** of the matrices $B(\mathbf{c})$

The transition matrix

$$B(\mathbf{c}) = \begin{pmatrix} 0 & \mathbf{1} & 0 & 0 & \cdots & 0 & 0 & 0 & 0 \\ c_1 & 0 & 0 & 1 - c_1 & \cdots & 0 & 0 & 0 & 0 \\ 1 - c_1 & 0 & 0 & c_1 & & \vdots & \vdots & & \\ \vdots & \vdots & & \vdots & \vdots & \vdots & \vdots & & \\ 0 & 0 & 0 & 0 & \cdots & c_{N-1} & 0 & 0 & 1 - c_{N-1} \\ 0 & 0 & 0 & 0 & \cdots & 1 - c_{N-1} & 0 & 0 & c_{N-1} \\ 0 & 0 & 0 & 0 & \cdots & 0 & 0 & \mathbf{1} & 0 \end{pmatrix}$$



The transition matrix / Properties

- If $c_j \cdot c_{j+1} > 0$ for some j , then the eigenvalues with max. modulus are exactly **two** ($\lambda = \pm 1$) and they are **simple**.

The transition matrix / Properties

- If $c_j \cdot c_{j+1} > 0$ for some j , then the eigenvalues with max. modulus are exactly **two** ($\lambda = \pm 1$) and they are **simple**.
- In the special case $\mathbf{c} = c(1, \dots, 1)$ ($k(x) = \bar{k} > 0$, g' constant), then

$$B(\mathbf{c}) = (1 - c)B(\mathbf{0}) + cB_1$$

is a **Birkhoff decomposition** for $B(\mathbf{c})$, where $B(\mathbf{0})$, B_1 are **permutation matrices**.

The transition matrix / Properties

- If $c_j \cdot c_{j+1} > 0$ for some j , then the eigenvalues with max. modulus are exactly **two** ($\lambda = \pm 1$) and they are **simple**.

- In the special case $\mathbf{c} = c(1, \dots, 1)$ ($k(x) = \bar{k} > 0$, g' constant), then

$$B(\mathbf{c}) = (1 - c)B(\mathbf{0}) + cB_1$$

is a **Birkhoff decomposition** for $B(\mathbf{c})$, where $B(\mathbf{0})$, B_1 are **permutation matrices**.

- **Birkhoff Theorem:** A $n \times n$ matrix is doubly stochastic if and only if it is a **convex combination** of (at most $n!$) **permutation matrices**.

The transition matrix / Properties

- If $c_j \cdot c_{j+1} > 0$ for some j , then the eigenvalues with max. modulus are exactly **two** ($\lambda = \pm 1$) and they are **simple**.

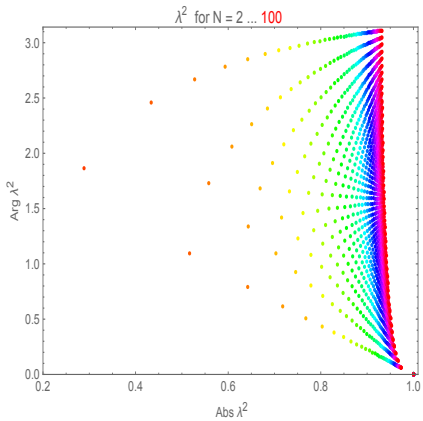
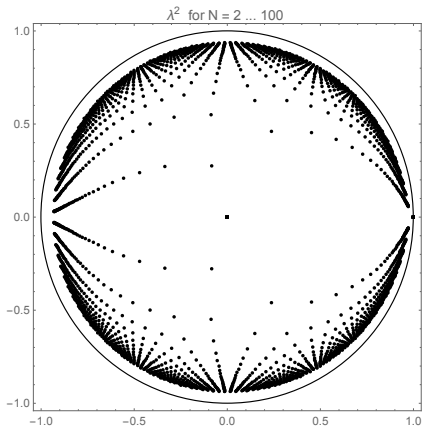
- In the special case $\mathbf{c} = c(1, \dots, 1)$ ($k(x) = \bar{k} > 0$, g' constant), then

$$B(\mathbf{c}) = (1 - c)B(\mathbf{0}) + cB_1$$

is a **Birkhoff decomposition** for $B(\mathbf{c})$, where $B(\mathbf{0})$, B_1 are **permutation matrices**.

- **Birkhoff Theorem:** A $n \times n$ matrix is doubly stochastic if and only if it is a **convex combination** of (at most $n!$) **permutation matrices**.
- References on doubly stochastic matrices (books): **Horn–Johnson**, **Bapat–Raghavan**, **D.Serre**

Some plots



Iterations of the transition matrix

- By iteration,

$$\sigma(t^n+) = \mathcal{B}_n \sigma(0+), \quad \mathcal{B}_n \doteq [B^{(n)} B^{(n-1)} \dots B^{(2)} B^{(1)}] \in M_{2N}.$$

Iterations of the transition matrix

- By iteration,

$$\sigma(t^n+) = \mathcal{B}_n \sigma(0+), \quad \mathcal{B}_n \doteq [B^{(n)} B^{(n-1)} \dots B^{(2)} B^{(1)}] \in M_{2N}.$$

- Spectral radius of each $B^{(n)}=1$. The eigenspace of the eigenv. ± 1 is **the same for every** $B^{(n)}$.

Iterations of the transition matrix

- By iteration,

$$\sigma(t^{n+}) = \mathcal{B}_n \sigma(0+), \quad \mathcal{B}_n \doteq [B^{(n)} B^{(n-1)} \dots B^{(2)} B^{(1)}] \in M_{2N}.$$

- Spectral radius of each $B^{(n)} = 1$. The eigenspace of the eigenv. ± 1 is **the same for every** $B^{(n)}$.

Good news:

Because of the boundary conditions $J(0, t) = J(1, t)$, the projection of $\sigma(t)$ onto the eigenspace for $\lambda = 1$ is zero. Also the one for $\lambda = -1$ is harmless.

Problem:

Estimate the **second maximal modulus** of the eigenvalues of $B^{(n)}$ and of \mathcal{B}_n .

Two parameters, both $\rightarrow \infty$: n and $N = \Delta x^{-1}$

Iterations of the transition matrix

- By iteration,

$$\sigma(t^{n+}) = \mathcal{B}_n \sigma(0+), \quad \mathcal{B}_n \doteq [B^{(n)} B^{(n-1)} \dots B^{(2)} B^{(1)}] \in M_{2N}.$$

- Spectral radius of each $B^{(n)} = 1$. The eigenspace of the eigenv. ± 1 is **the same for every** $B^{(n)}$.

Good news:

Because of the boundary conditions $J(0, t) = J(1, t)$, the projection of $\sigma(t)$ onto the eigenspace for $\lambda = 1$ is zero. Also the one for $\lambda = -1$ is harmless.

Problem:

Estimate the **second maximal modulus** of the eigenvalues of $B^{(n)}$ and of \mathcal{B}_n .

Two parameters, both $\rightarrow \infty$: n and $N = \Delta x^{-1}$

Possible tools:

Nonhomogeneous Markov chains... Joint Spectral Radius... **Matrix theory!**

An estimate for the eigenvalues

Write $B(\mathbf{c})$ as a perturbation:

$$B(\mathbf{c}) = B(\mathbf{0}) + E(\mathbf{c})$$

with

$$\|E(\mathbf{c})\| = \max_{\|v\|=1} \|E(\mathbf{c})v\| = 2 \max_{j=1, \dots, N-1} c_j \rightarrow 0 \quad N \rightarrow \infty$$

An estimate for the eigenvalues

Write $B(\mathbf{c})$ as a perturbation:

$$B(\mathbf{c}) = B(\mathbf{0}) + E(\mathbf{c})$$

with

$$\| \| E(\mathbf{c}) \| \| = \max_{\|v\|=1} \|E(\mathbf{c})v\| = 2 \max_{j=1, \dots, N-1} c_j \rightarrow 0 \quad N \rightarrow \infty$$

A Rayleigh quotient formula

$$\lambda_\ell = \mu_\ell + \frac{\mathbf{y}_\ell^* E(\mathbf{c}) \mathbf{x}_\ell}{\mathbf{y}_\ell^* \mathbf{x}_\ell} + \mathcal{O}(\| \| E(\mathbf{c}) \| \|^2), \quad \mu_\ell = e^{\frac{i\pi\ell}{N}}$$

where $\mathbf{x}_\ell, \mathbf{y}_\ell \in \mathbb{C}^{2N}$ are resp. the right and left eigenvector associated to μ_ℓ

An estimate for the eigenvalues

Write $B(\mathbf{c})$ as a perturbation:

$$B(\mathbf{c}) = B(\mathbf{0}) + E(\mathbf{c})$$

with

$$\|E(\mathbf{c})\| = \max_{\|v\|=1} \|E(\mathbf{c})v\| = 2 \max_{j=1, \dots, N-1} c_j \rightarrow 0 \quad N \rightarrow \infty$$

A Rayleigh quotient formula

$$\lambda_\ell = \mu_\ell + \frac{\mathbf{y}_\ell^* E(\mathbf{c}) \mathbf{x}_\ell}{\mathbf{y}_\ell^* \mathbf{x}_\ell} + \mathcal{O}(\|E(\mathbf{c})\|^2), \quad \mu_\ell = e^{\frac{i\pi\ell}{N}}$$

where $\mathbf{x}_\ell, \mathbf{y}_\ell \in \mathbb{C}^{2N}$ are resp. the right and left eigenvector associated to μ_ℓ

Lemma:

If $k(x) \geq \bar{k} \cdot \chi_{(\alpha, \beta)}$ with $\bar{k} > 0$ and if $\inf g' > 0$, then there exists $C = C(\alpha, \beta) > 0$ such that for every $\mu_\ell \neq \pm 1$ and N large enough one has

$$\left| \mu_\ell + \frac{\mathbf{y}_\ell^* E(\mathbf{c}) \mathbf{x}_\ell}{\mathbf{y}_\ell^* \mathbf{x}_\ell} \right| \leq 1 - \frac{\bar{k}(\inf g')}{N} C.$$

END OF PART I

... To be continued ...