

State estimation for linear age-structured population diffusion models

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Valenciennes, Stability of non-conservative systems
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The problem in a glance

A classical model for **age-space structured populations** is given by

$$\left\{ \begin{array}{l} \frac{\partial p}{\partial t}(a, x, t) + \frac{\partial p}{\partial a}(a, x, t) \\ \quad = -\mu(a)p(a, x, t) + k\Delta p(a, x, t), \quad a \in (0, a^*), \quad x \in \Omega, \quad t > 0, \\ p(a, x, t) = 0, \quad a \in (0, a^*), \quad x \in \partial\Omega, \quad t > 0, \\ p(a, x, 0) = p_0(a, x), \quad a \in (0, a^*), \quad x \in \Omega, \\ p(0, x, t) = \int_0^{a^*} \beta(a)p(a, t, x) da, \quad x \in \Omega, \quad t > 0. \end{array} \right.$$

- $p(a, x, t)$: distribution density of the population of age a at spatial position x at time t ;
- a^* : maximal life expectancy;
- k : diffusion coefficient;
- $\mu(a), \beta(a)$: death and birth rates (independent of x);

The problem in a glance

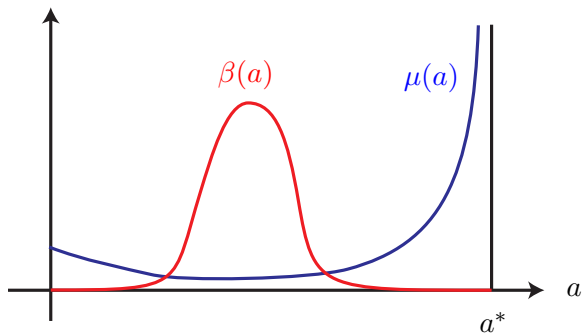
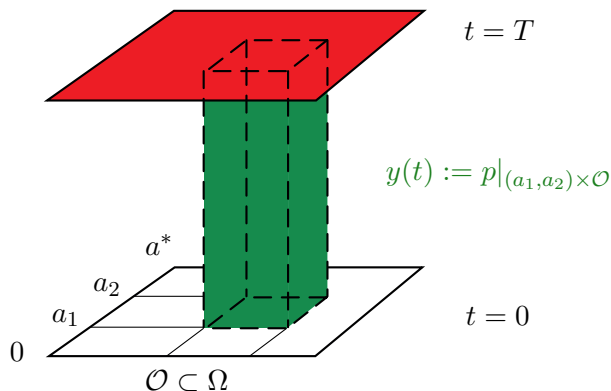


Figure: Typical birth and death rates.

The problem in a glance



Estimation problem

Knowing the output $y(t) := p|_{(a_1, a_2) \times \mathcal{O}}$ (but assuming that p_0 is unknown), estimate $p(a, x, T)$ for all $a \in (0, a^*)$ and $x \in \Omega$, as $T \rightarrow +\infty$.

The problem in a glance

$$\begin{cases} \dot{p}(t) = Ap(t), & t \in (0, T) \\ p(0) = p_0, \\ y(t) = Cp(t), & t \in (0, T), \end{cases}$$

where $C \in \mathcal{L}(X, Y)$, $Y := L^2((a_1, a_2) \times \mathcal{O})$ is defined by

$$C\varphi := \varphi|_{(a_1, a_2) \times \mathcal{O}} \text{ for all } \varphi \in X.$$

We introduce the **Luenberger observer**

$$\begin{cases} \dot{\hat{p}}(t) = A\hat{p}(t) + L(C\hat{p}(t) - y(t)), & t \in (0, T) \\ \hat{p}(0) = 0, \end{cases}$$

where $L \in \mathcal{L}(Y, X)$ is a linear operator to be defined.

Then the **error** $e := \hat{p} - p$ satisfies

$$\begin{cases} \dot{e}(t) = (A + LC)e(t), & t \in (0, T) \\ e(0) = -p_0. \end{cases}$$

The problem in a glance

Goal

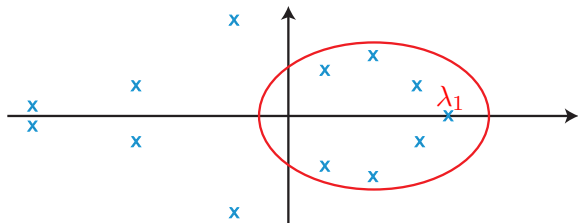
Find L such that $e^{t(A+LC)}$ exponentially stable (detectability).

The problem in a glance

Goal

Find L such that $e^{t(A+LC)}$ exponentially stable (detectability).

How?



Spectrum of A : an infinite number of stable modes
and a finite number of unstable modes.

Design an infinite dimensional Luenberger observer via a finite dimensional stabilizing operator.

- **Population Dynamics**

- **Semigroup properties:** Song et al., Chan, Guo, Li et al., Langlais, Walker
- **Controllability problems:** Ainseba, Anita, Iannelli, Langlais, Echarroudi, Maniar, Traoré, Kavian
- **Inverse problems:** Traoré, Rundell, Di Blasio, Lorenzi, Perasso, Picart
- **Numerical aspects:** Lopez, Triggiani, Milner, Kim, Huyer, Ayati, Dupont, Pelovska, Gerardo-Giorda

- **State Space Splitting**

- **Abstract setting:** Russel, Triggiani, Jacobson & Nett, Jacob & Zwart
- **Stabilization of PDE:** Barbu & Triggiani, Raymond et al., Badra & Takahashi

Outline

- 1 Spectral properties of the operator
- 2 Detectability
- 3 Application : observer design for populations dynamics
- 4 Numerical results

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The model

$$\left\{ \begin{array}{l} \frac{\partial p}{\partial t}(a, x, t) = -\frac{\partial p}{\partial a}(a, x, t) \\ \quad -\mu(a)p(a, x, t) + k\Delta p(a, x, t), \quad a \in (0, a^*), \quad x \in \Omega, \quad t > 0, \\ p(a, x, t) = 0, \quad a \in (0, a^*), \quad x \in \partial\Omega, \quad t > 0, \\ p(a, x, 0) = p_0(a, x), \quad a \in (0, a^*), \quad x \in \Omega, \\ p(0, x, t) = \int_0^{a^*} \beta(a)p(a, t, x) da, \quad x \in \Omega, \quad t > 0. \end{array} \right.$$

Assumptions

Typical assumptions on the birth and death rates β and μ :

- $\beta \in L^\infty(0, a^*)$, $\beta \geq 0$ a.e. in $(0, a^*)$;
- $\mu \in L^1_{\text{loc}}(0, a^*)$, $\mu \geq 0$ a.e. in $(0, a^*)$ and

$$\lim_{a \rightarrow a^*} \int_0^a \mu(s) ds = +\infty.$$

We also introduce the function

$$\Pi(a) := \exp\left(-\int_0^a \mu(s) ds\right)$$

which represents the probability to survive at age $a > 0$.

In particular

$$\lim_{a \rightarrow a^*} \Pi(a) = 0.$$

First order formulation

We introduce the Hilbert space $X := L^2((0, a^*) \times \Omega)$ and let A be defined by:

$$\mathcal{D}(A) = \left\{ \varphi \in X \cap L^2((0, a^*), H_0^1(\Omega)) \mid -\frac{\partial \varphi}{\partial a} - \mu \varphi + k \Delta \varphi \in X; \right. \\ \left. \begin{aligned} &\varphi(a, \cdot)|_{\partial \Omega} = 0 \text{ for almost all } a \in (0, a^*); \\ &\varphi(0, x) = \int_0^{a^*} \beta(a) \varphi(a, x) da \text{ for almost all } x \in \Omega \end{aligned} \right\}$$

$$A\varphi = -\frac{\partial \varphi}{\partial a} - \mu \varphi + k \Delta \varphi, \quad \forall \varphi \in \mathcal{D}(A).$$

The population dynamics problem reads then

$$\begin{cases} \dot{p}(t) = Ap(t), & t > 0 \\ p(0) = p_0. \end{cases}$$

Theorem (Chan and Guo, 1989)

- *A is the infinitesimal generator of a C_0 -semigroup e^{tA} on X .*
- *If $p_0 \in X$, there exists a unique solution $p \in C([0, \infty), X)$.*
- *If $p_0 \in \mathcal{D}(A)$, there exists a unique solution $p \in C([0, \infty), \mathcal{D}(A)) \cap C^1([0, \infty), X)$.*

Diffusion free model

McKendrick–Von Foerster model (1959) describes the diffusion free case ($k = 0$) :

$$\left\{ \begin{array}{ll} \frac{\partial p}{\partial t}(a, t) = -\frac{\partial p}{\partial a}(a, t) - \mu(a)p(a, t), & a \in (0, a^*), t > 0, \\ p(a, 0) = p_0(a), & a \in (0, a^*), \\ p(0, t) = \int_0^{a^*} \beta(a)p(a, t) da, & t > 0. \end{array} \right.$$

Diffusion free model

The population operator A_0 corresponding to the above system is defined as follows

$$\mathcal{D}(A_0) = \left\{ \varphi \in L^2(0, a^*) \mid -\frac{d\varphi}{da} - \mu\varphi \in L^2(0, a^*); \right. \\ \left. \varphi(0) = \int_0^{a^*} \beta(a)\varphi(a) da \right\}.$$

$$A_0\varphi = -\frac{d\varphi}{da} - \mu\varphi, \quad \forall \varphi \in \mathcal{D}(A_0).$$

Then the McKendrick–Von Foerster model reads then

$$\begin{cases} \dot{p}(t) = A_0 p(t), & t > 0 \\ p(0) = p_0. \end{cases}$$

Theorem (Song et al., 1982)

- 1 A_0 has *compact resolvent* and its spectrum is constituted of a *countable (infinite) set of isolated eigenvalues* with finite algebraic multiplicity.
- 2 The eigenvalues $(\lambda_n^0)_{n \geq 1}$ of A_0 (counted without multiplicity) the (complex) solutions of the *characteristic equation*

$$F(\lambda) := \int_0^{a^*} \beta(a) \Pi(a) e^{-\lambda a} da = 1.$$

- 3 The eigenvalues $(\lambda_n^0)_{n \geq 1}$ are of *geometric multiplicity* one:

$$\varphi_n^0(a) = e^{-\lambda_n^0 a} \Pi(a) = e^{-\lambda_n^0 a - \int_0^a \mu(s) ds}.$$

- 4 Every vertical strip of the complex plane contains a *finite number* of eigenvalues of A_0 .

Theorem (Song et al., 1982)

The operator A_0 has a unique real eigenvalue λ_1^0 . Moreover:

- 1 λ_1^0 is of *algebraic multiplicity one*;
- 2 $\lambda_1^0 > 0$ (< 0) $\iff F(0) = \int_0^{a^*} \beta(a)\Pi(a) da > 1$ (< 1);
- 3 λ_1^0 is a *real dominant eigenvalue*:

$$\lambda_1^0 > \operatorname{Re}(\lambda_n^0), \quad \forall n \geq 2.$$

Back to the problem with diffusion

$$\mathcal{D}(A) = \left\{ \varphi \in X \cap L^2((0, a^*), H_0^1(\Omega)) \mid -\frac{\partial \varphi}{\partial a} - \mu \varphi + k \Delta \varphi \in X; \right. \\ \left. \begin{aligned} &\varphi(a, \cdot)|_{\partial \Omega} = 0 \text{ for almost all } a \in (0, a^*); \\ &\varphi(0, x) = \int_0^{a^*} \beta(a) \varphi(a, x) da \text{ for almost all } x \in \Omega \end{aligned} \right\}$$
$$A\varphi = -\partial_a \varphi - \mu \varphi + k \Delta \varphi, \quad \forall \varphi \in \mathcal{D}(A).$$

Let $0 < \lambda_1^D < \lambda_2^D \leq \lambda_3^D \leq \dots$ be the increasing sequence of eigenvalues of $-k\Delta$ with Dirichlet boundary conditions and let $(\varphi_n^D)_{n \geq 1}$ be a corresponding orthonormal basis of $L^2(\Omega)$.

Spectral properties

Theorem (Chan and Guo, 1989)

- ① A has compact resolvent and its (pure point) spectrum is

$$\sigma(A) = \{ \lambda_i^0 - \lambda_j^D \mid i, j \in \mathbb{N}^* \}$$

- ② The eigenspace associated to an eigenvalue λ of A is given by

$$\text{Span} \left\{ \varphi_i^0(a) \varphi_j^D(x) = e^{-\lambda_i^0 a} \Pi(a) \varphi_j^D(x) \mid \lambda_i^0 - \lambda_j^D = \lambda \right\}.$$

- ③ The real eigenvalue λ_1 of A is dominant:

$$\lambda_1 = \lambda_1^0 - \lambda_1^D > \text{Re}(\lambda), \quad \forall \lambda \in \sigma(A), \lambda \neq \lambda_1.$$

- ④ λ_1 is a simple eigenvalue, the corresponding eigenspace being generated by

$$\varphi_1(a, x) := \varphi_1^0(a) \varphi_1^D(x) = e^{-\lambda_1^0 a} \Pi(a) \varphi_1^D(x).$$

Spectral properties

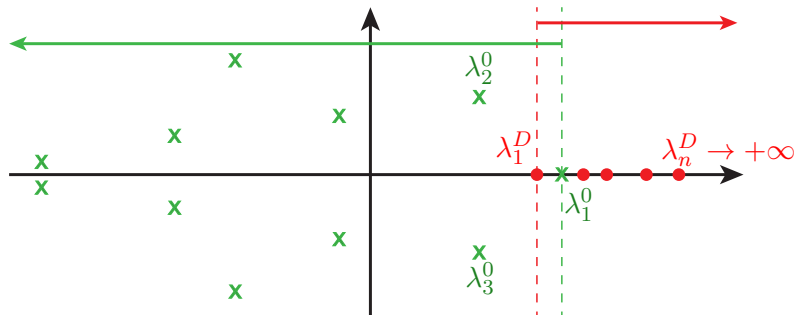


Figure: Spectra of A_0 and $-k\Delta$.

In this example, there is only 1 unstable eigenvalue λ_1 :

$$\operatorname{Re}(\lambda_n^0) < \lambda_1^D < \lambda_1^0 < \lambda_2^D, \quad \forall n \geq 2$$

$$\implies \lambda_1 = \lambda_1^0 - \lambda_1^D > 0, \quad \operatorname{Re}(\lambda_n) < 0, \quad \forall n \geq 2$$

Compactness & Stability

Proposition (Chan and Guo, 1989)

The semigroup e^{tA} generated on X by A is compact for $t \geq a^$.*

This implies in particular that (see Zabczyk, 1975)

$$\omega_a(A) = \omega_0(A)$$

where $\omega_a(A) := \lim_{t \rightarrow +\infty} t^{-1} \ln \|e^{tA}\|$ denotes the **growth bound** of e^{tA} and $\omega_0(A) := \sup \{\operatorname{Re} \lambda \mid \lambda \in \sigma(A)\}$ the **spectral bound** of A .

Consequence

The above condition ensures that the **exponential stability** of e^{tA} is equivalent to the condition

$$\omega_0(A) = \sup \{\operatorname{Re} \lambda \mid \lambda \in \sigma(A)\} < 0.$$

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Abstract framework

Consider

- $A : \mathcal{D}(A) \rightarrow X$ with compact resolvent on a Hilbert space X generating a C_0 -semigroup in X ,
- $C \in \mathcal{L}(X, Y)$, where Y is another Hilbert space.

We assume

(A1) A admits M eigenvalues (counted without multiplicities) with real part greater or equal than 0:

$$\cdots \leq \operatorname{Re} \lambda_{M+2} \leq \operatorname{Re} \lambda_{M+1} < 0 \leq \operatorname{Re} \lambda_M \leq \cdots \leq \operatorname{Re} \lambda_2 \leq \operatorname{Re} \lambda_1.$$

(A2) We have the equality

$$\omega_a(A) = \omega_0(A).$$

Detectability

Definition

The pair (A, C) is **detectable** if there exists $L \in \mathcal{L}(Y, X)$ such that $(A + LC)$ generates an exponentially stable semigroup.

We are going to show that:

Spectral observability of unstable eigenfunctions of A

$$(A\varphi = \lambda\varphi \text{ for } \lambda \in \Sigma_+ \text{ and } C\varphi = 0) \implies \varphi = 0$$



Detectability of the finite dimensional system (A^+, C^+)

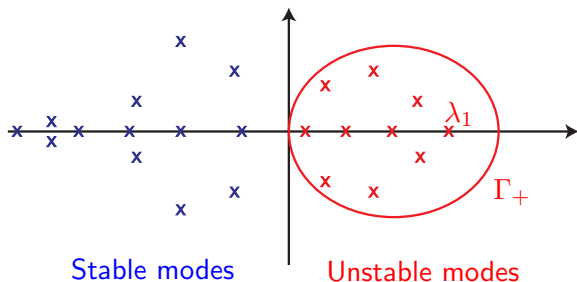


Detectability of the infinite dimensional system (A, C)

Projection operator

We set $\Sigma_+ := \{\lambda_1, \dots, \lambda_M\}$ and let Γ_+ be a positively oriented curve enclosing Σ_+ but no other point of the spectrum of A . Let $P_+ : X \rightarrow X$ be the projection operator defined by

$$P_+ := -\frac{1}{2\pi i} \int_{\Gamma_+} (\xi - A)^{-1} d\xi.$$



Splitting

We set $X_+ := P_+X$ and $X_- := (I - P_+)X$, and then P_+ provides the following decomposition of X

$$X = X_+ \oplus X_-.$$

Following Russell and Triggiani, we can decompose our system into two subsystems :

- a finite dimensional system to be stabilized,
- a stable infinite dimensional system.

More precisely, X_+ and X_- are invariant subspaces under A (since A and P_+ commute) and the spectra of the restricted operators $A|_{X_+}$ and $A|_{X_-}$ are respectively Σ_+ and $\Sigma_- := \sigma(A) \setminus \Sigma_+$. We also define

$$A_+ := A|_{\mathcal{D}(A) \cap X_+} : \mathcal{D}(A) \cap X_+ \longrightarrow X_+,$$

$$A_- := A|_{\mathcal{D}(A) \cap X_-} : \mathcal{D}(A) \cap X_- \longrightarrow X_-.$$

Splitting

If A is **diagonalizable**, the space $X_+ = P_+X$ is the **finite dimensional** space spanned by the **eigenfunctions** of A associated to the **unstable eigenvalues**:

$$X_+ = \bigoplus_{k=1}^M \text{Ker}(A - \lambda_k).$$

and

$$\dim X_+ = \sum_{k=1}^M m_k^G.$$

where $m_k^G := \dim \text{Ker}(A - \lambda_k)$ is the **geometric multiplicity** of λ_k .

Splitting

In the general case, the space X_+ is the **finite dimensional** space spanned by the **generalized eigenfunctions** of A associated to the **unstable eigenvalues**:

$$X_+ = \bigoplus_{k=1}^M \text{Ker} (A - \lambda_k)^{m_k^P}$$

where m_k^P is the **multiplicity of the pole** λ_k in the resolvent $(A - \lambda)^{-1}$.

The space $\text{Ker} (A - \lambda_k)^{m_k^P}$ is called the generalized eigenspace associated to λ_k . Its dimension m_k^A is the **algebraic multiplicity** of λ_k .

$$\dim X_+ = \sum_{k=1}^M m_k^A.$$

Detectability result : from L_+ to L

Theorem

Let

- $Q_+ : Y \rightarrow Y_+ := CX_+$ be the orthogonal projection operator from Y to Y_+ ,
- $i_{X_+} : X_+ \rightarrow X$ be the embedding operator from X_+ into X .

Set

$$C_+ = Ci_{X_+} \in \mathcal{L}(X_+, Y_+)$$

and assume that the finite dimensional projected system (A_+, C_+) is detectable through $L_+ \in \mathcal{L}(Y_+, X_+)$.

Then, the infinite dimensional system (A, C) is detectable through

$$L = i_{X_+}L_+Q_+ \in \mathcal{L}(Y, X).$$

Proof

For $L \in \mathcal{L}(Y, X)$, consider the system

$$\dot{z}(t) = (A + LC)z(t).$$

If we write $z = z_+ + z_-$ where $z_+ := P_+z$ and $z_- := (I - P_+)z$, by applying P_+ and $(I - P_+)$ to the above equation, we obtain a corresponding splitting of the system into two subsystems:

$$\begin{cases} \dot{z}_+(t) &= A_+z_+(t) + P_+LCz(t), \\ \dot{z}_-(t) &= A_-z_-(t) + (I - P_+)LCz(t). \end{cases}$$

Taking $L = i_{X_+}L_+Q_+$ and using the identities $P_+i_{X_+} = \text{Id}_{X_+}$ and $(I - P_+)i_{X_+} = 0$, we obtain

$$\begin{cases} \dot{z}_+(t) &= A_+z_+(t) + L_+Q_+Cz(t), \\ \dot{z}_-(t) &= A_-z_-(t). \end{cases}$$

It follows from assumption **(A2)** that z_- is exponentially stable:

$$\|z_-(t)\| \leq K e^{-\omega_- t} \|z_-(0)\|$$

where $0 < \omega_- \leq -\operatorname{Re} \lambda_{M+1}$. On the other hand, by using $C_+ = Q_+ C i_{X_+}$ and since $i_{X_+} z_+ = z_+$, we have

$$\begin{aligned} \dot{z}_+(t) &= A_+ z_+(t) + L_+ Q_+ C (z_+(t) + z_-(t)) \\ &= A_+ z_+(t) + L_+ Q_+ C i_{X_+} z_+(t) + L_+ Q_+ C z_-(t) \\ &= (A_+ + L_+ C_+) z_+(t) + L_+ Q_+ C z_-(t). \end{aligned}$$

Proof

Using Duhamel's formula, we get

$$z_+(t) = \mathbb{T}_t^+ z_+(0) + \int_0^t \mathbb{T}_{t-s}^+ L_+ Q_+ C z_-(s) ds,$$

where \mathbb{T}_t^+ is the semigroup generated by $A_+ + L_+ C_+$, which is exponentially stable by the detectability assumption, i.e. there exists $\omega_+ > 0$ such that

$$\|\mathbb{T}_t^+ x\| \leq K e^{-\omega_+ t} \|x\| \quad \forall x \in X_+, \forall t > 0.$$

Combined with exponential stability of z_- , this yields

$$\|z_+(t)\| \leq K \left\{ e^{-\omega_+ t} \|z_+(0)\| + \|L_+\| \|C\| \int_0^t e^{-\omega_+(t-s)} e^{-\omega_- s} \|z_-(0)\| ds \right\},$$

and consequently

$$\|z_+(t)\| \leq K \left(e^{-\omega_+ t} + \|L_+\| \|C\| \frac{e^{-\omega_+ t} - e^{-\omega_- t}}{\omega_- - \omega_+} \right) \|z_0\|.$$

Proof

It is then sufficient to choose ω_+ small enough such that $0 < \omega_+ < \omega_-$ to have the exponential decay of $t \mapsto z_+(t)$:

$$\|z_+(t)\| \leq K e^{-\omega_+ t} \|z_0\|, \quad t > 0.$$

We have thus proved the exponential decay of $z = z_+ + z_-$. ■

Hautus test

The following result provide a sufficient condition of Hautus type for the detectability of the finite dimensional projected system (A_+, C_+) .

Proposition

If the spectral observability condition (Hautus test)

$$(A\varphi = \lambda\varphi \text{ for } \lambda \in \Sigma_+ \text{ and } C\varphi = 0) \implies \varphi = 0$$

is satisfied, then (A_+, C_+) is detectable.

Proof : Since $C_+z_+ = Cz_+$ for any $z_+ \in X_+$, if the Hautus test is satisfied, then it is clear that the following Hautus test is also satisfied:

$$(\varphi \in \mathcal{D}(A) \cap X_+ \mid A_+\varphi = \lambda\varphi \text{ and } C_+\varphi = 0) \implies \varphi = 0.$$

As the above system is finite dimensional, (A_+, C_+) is detectable. ■

Corollaire

If the Hautus test is satisfied, then (A, C) is detectable via the stabilizing output injection operator L defined previously.

- The matrices A_+ and C_+ are in practice of **small size** : their dimensions are respectively $\dim X_+ \times \dim X_+$ and $\dim Y_+ \times \dim X_+$.
- The stabilizing operator L_+ of the finite dimensional system (A_+, C_+) can be determined by solving a finite dimensional algebraic Riccati equation.

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Assumptions **(A1)** ($M < \infty$ unstable eigenvalues) and **(A2)** ($\omega_a(A) = \omega_0(A)$) are satisfied for our population model and the problem of determining the stabilizing operator L for (A, C) fits into the framework described above.

It only remains to verify that the Hautus test is satisfied for our system (A, C) :

Lemma

If $\varphi \in \mathcal{D}(A)$ satisfies $A\varphi = \lambda\varphi$ for $\lambda \in \Sigma_+$ and $C\varphi = 0$, then φ vanishes identically.

Proof of the spectral observability

Let λ be an unstable eigenvalue of A and let $\varphi \in \mathcal{D}(A)$ satisfying $A\varphi = \lambda\varphi$. Decomposing $\varphi(0, x)$ in the basis of $L^2(\Omega)$ constituted of the eigenfunctions of $-k\Delta$, the unique solution of the evolution system

$$\begin{cases} \frac{\partial \varphi}{\partial a}(a, x) = k\Delta\varphi(a, x) - (\lambda + \mu)\varphi(a, x), & a \in (0, a^*), x \in \Omega, \\ \varphi(a, x) = 0, & a \in (0, a^*), x \in \partial\Omega, \\ \varphi(0, x) = \sum_{j \in \mathbb{N}} \alpha_j \varphi_j^D(x), & x \in \Omega, \end{cases}$$

is given by

$$\varphi(a, x) = \sum_{j \in \mathbb{N}} \alpha_j e^{-(\lambda + \lambda_j^D)a} \Pi(a) \varphi_j^D(x).$$

Plugging the above expression in the renewal equation, we obtain

$$\sum_{j \in \mathbb{N}} \alpha_j \varphi_j^D(x) = \sum_{j \in \mathbb{N}} \alpha_j \left(\int_0^{a^*} \beta(a) e^{-(\lambda + \lambda_j^D)a} \Pi(a) \right) \varphi_j^D(x).$$

We see that is equivalent to, for any $j \in \mathbb{N}$, either $\alpha_j = 0$, either $\lambda + \lambda_j^D$ solves the characteristic equation of the diffusion free problem.

Proof of the spectral observability (end)

Consequently, we have

$$\varphi(a, x) = \sum_{j|\lambda+\lambda_j^D \in \sigma(A_0)} \alpha_j e^{-(\lambda+\lambda_j^D)a} \Pi(a) \varphi_j^D(x).$$

The condition $C\varphi = 0$ reads then

$$\sum_{j|\lambda+\lambda_j^D \in \sigma(A_0)} \alpha_j e^{-(\lambda+\lambda_j^D)a} \varphi_j^D|_{\mathcal{O}}(x) = 0, \quad a \in (a_1, a_2).$$

Since the eigenfunctions of $-k\Delta$ with Dirichlet boundary conditions are analytic, we immediately obtain that $\varphi = 0$. ■

Main result

Theorem

Let $p_0 \in X$ and assume that $y(t) = p|_{(a_1, a_2) \times \mathcal{O}}$ ($t > 0$) is known.
Let \hat{p} the observer defined by

$$\begin{cases} \dot{\hat{p}}(t) = A\hat{p}(t) + L(C\hat{p}(t) - y(t)), & t \in (0, T) \\ \hat{p}(0) = 0, \end{cases}$$

where $L \in \mathcal{L}(Y, X)$ is the stabilizing operator defined by

$$L = i_{X_+} L_+ Q_+ \in \mathcal{L}(Y, X).$$

Then, there exist $M, \omega > 0$ such that

$$\|\hat{p}(t) - p(t)\| \leq M e^{-\omega t} \|p_0\|, \quad t > 0.$$

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Full observation in age

Taking $\Omega = (0, \pi)$ and assuming that p_0 is an **unknown initial data**, we want to **estimate p at time $t = T$** where p solves:

$$\left\{ \begin{array}{l} \partial_t p(a, x, t) + \partial_a p(a, x, t) \\ \quad = -\mu(a)p(a, x, t) + \partial_{xx} p(a, x, t), \quad a \in (0, a^*), x \in (0, \pi), t > 0, \\ p(a, 0, t) = p(a, \pi, t) = 0, \quad a \in (0, a^*), t > 0, \\ p(a, x, 0) = p_0(a, x), \quad a \in (0, a^*), x \in (0, \pi), \\ p(0, x, t) = \int_0^{a^*} \beta(a)p(a, x, t) da, \quad x \in (0, \pi), t > 0, \end{array} \right.$$

provided we **know the observation**

$$y(t) = p(t)|_{(0, a^*) \times (\pi/3, 2\pi/3)}, \quad t \in (0, T).$$

The fertility and mortality functions

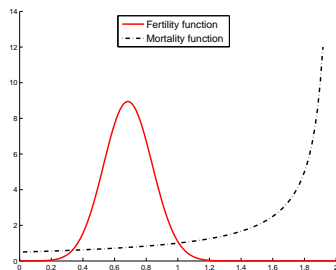


Figure: The fertility and mortality functions.

Taking $a^* = 2$, we choose the fertility and mortality function to be

$$\beta(a) = 10 a(a^* - a) \exp \left\{ -20(a - a^*/3)^2 \right\}, \quad \mu(a) = (a^* - a)^{-1}.$$

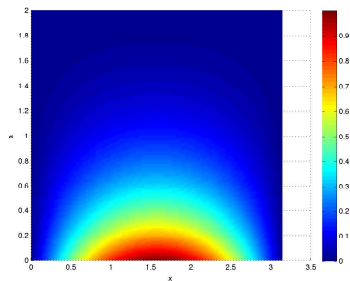
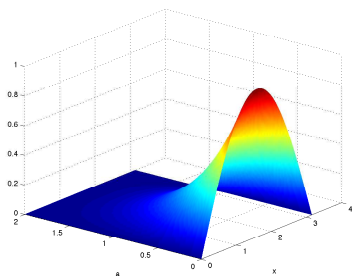
Note that the function $\Pi(a)$ can be computed explicitly:

$$\Pi(a) = \exp \left(- \int_0^a \mu(s) ds \right) = \frac{a^* - a}{a^*}.$$

First test : initial state = unstable eigenfunction

Under these assumptions, there is a unique unstable eigenvalue $\lambda_1 = \lambda_1^0 - \pi^2$ (where $\lambda_1^0 \in \mathbb{R}$ satisfies $F(\lambda_1^0) = 1$). Computing numerically this value, we obtain that $\lambda_1 = 0.239$. We first choose as initial state an eigenfunction corresponding to λ_1

$$p_0(a, x) = \varphi_1(a, x) = \varphi_1^0(a)\varphi_1^D(x) = \frac{a^* - a}{a^*} e^{-\lambda_1^0 a} \sin(x).$$



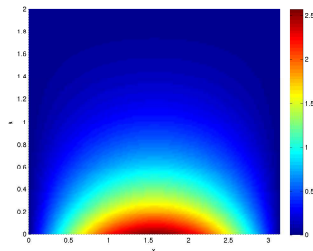
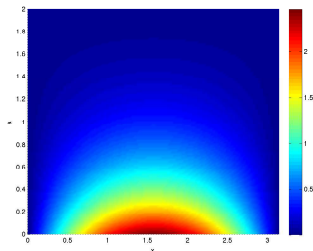
Estimated and exact solution at time $t = T$

The exact solution is:

$$p(a, x, t) = e^{\lambda_1 t} p_0(a, x).$$

We take: $T = 2a^*$, with $a^* = 2$.

Using $N_x = 100$, $N_a = 120$ and $N_t = 2N_a$, we obtain an L^2 relative error of 4.07%.

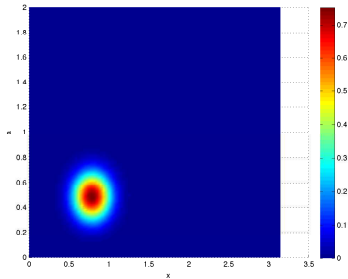
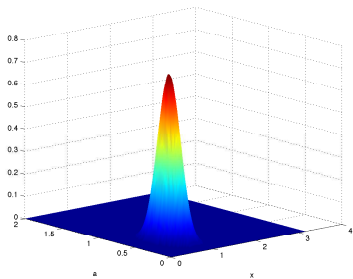


Estimated (left) and exact (right) solution at time $t = T$.

Second test : initial state = Gaussian function

We choose a space-aged localized initial distribution of population of gaussian type:

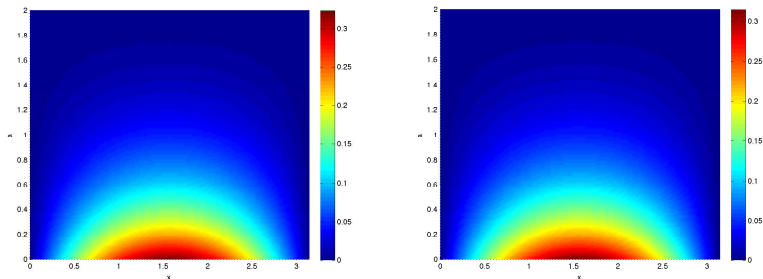
$$p_0(a, x) = \exp \left\{ - \left(30(a - a^*/4)^2 + 20(x - \ell/4)^2 \right) \right\}.$$



Gaussian initial state (3D and 2D representations).

Estimated and exact solution at time $t = T$

We obtain an L^2 relative error of 2.99%, 9.6% and 16.2% respectively for 5%, 10% and 15% of noise¹.

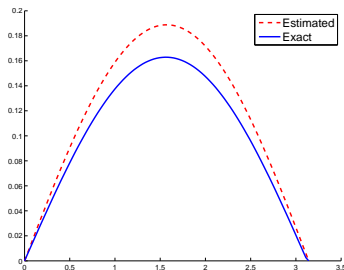
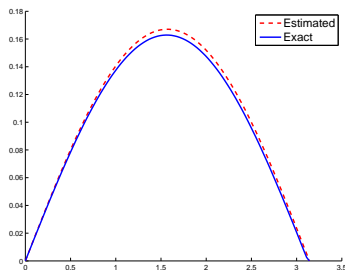


Estimated (left) and exact (right) solution at time $t = T$ (5% of noise).

¹ “Exact” solution refers here to a numerical solution computed numerically.

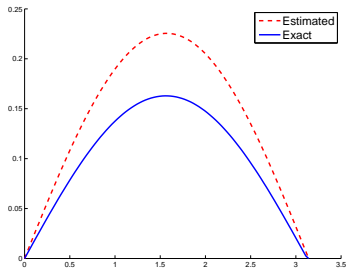
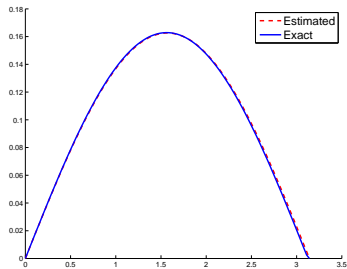
Estimated and exact final total population

$$P_T(x) = \int_0^{a^*} p(a, T, x) da \quad \text{and} \quad \hat{P}_T(x) = \int_0^{a^*} \hat{p}(a, T, x) da.$$



Estimated (dashed line) and exact total population at time $t = T$ with 5% of noise (left) and 15% of noise (right).

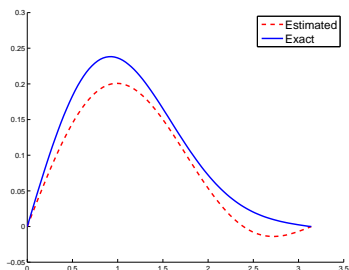
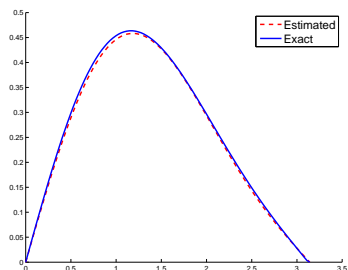
Distributed observation in space and age



Estimated (dashed line) and exact total population at time $t = T$ with age observation in $\left(0, \frac{a^*}{20}\right)$ (left) and $\left(\frac{a^*}{2}, a^*\right)$ (right).

Influence of the observation time

We consider a configuration with **two unstable eigenvalues** and we investigate the **influence of T** . For $T = 0.5a^*$, we obtain a relative error of 27% for the population density, but we still obtain a reasonable approximation for the total population.



Estimated (dashed line) and exact total population at time $t = T$, for $T = a^*$ (left) and for $T = 0.5a^*$ (right).

Conclusion

1 Other models

- Other **outputs** : $y(x, t) = \int_{a_1}^{a_2} p(a, x, t) da, \quad x \in \mathcal{O}.$
- **Space** dependent coefficients : $\beta(a, \mathbf{x}), \quad \mu(a, \mathbf{x}).$
- **Nonlinearities**: $\beta(a, x, P)$ and $\mu(a, x, P)$ where

$$P(x, t) := \int_0^{a^*} p(a, x, t) da.$$

- adaptative observer which gives an estimation of p and k .

2 Approximation

- **Convergence** analysis and error estimates
- **Uniform** exponential stability (with respect to Δa et h)

Particular case: $A_+ := A|_{\mathcal{D}(A) \cap X_+}$ diagonalizable

The results collected here can be found in Barbu and Triggiani 2004. We assume that $A_+ := A|_{\mathcal{D}(A) \cap X_+}$ is diagonalizable. For simplicity, we denote by N the number of unstable eigenvalues of A counted with multiplicities (still denoted λ_k , $k = 1, \dots, N$). This implies in particular that the unstable space is

$$X_+ = \bigoplus_{k=1}^N \text{Ker}(A - \lambda_k).$$

We denote then by $(\varphi_k)_{1 \leq k \leq N}$ a basis of X_+ . Denote by ψ_k an eigenfunction of A^* corresponding to the unstable eigenvalue $\overline{\lambda_k}$ ($1 \leq k \leq N$). It can be shown that the family $(\psi_k)_{1 \leq k \leq N}$ can be chosen such that $(\varphi_k)_{1 \leq k \leq N}$ and $(\psi_k)_{1 \leq k \leq N}$ form bi-orthogonal sequences, in the sense that $(\varphi_k, \psi_m)_X = \delta_{km}$. It follows then that the projection operator $P_+ \in \mathcal{L}(X, X_+)$ can be expressed as

$$P_+ z = \sum_{k=1}^N (z, \psi_k)_X \varphi_k \quad (z \in X).$$

Since

$$X_+ = P_+ X = \text{Span} \{ \varphi_k, 1 \leq k \leq N \},$$

it follows that

$$Y_+ = C X_+ = \text{Span} \{ C \varphi_k, 1 \leq k \leq N \}.$$

Assume now that the family

$$(C \varphi_k)_{1 \leq k \leq N} \text{ is linearly independent in } X. \quad (1)$$

This property holds true in the case of internal observation.

Therefore

$$\dim Y_+ = \dim X_+ = N.$$

We denote by \mathbb{G} the Hermitian matrix of size $N \times N$ defined by

$$\mathbb{G} = ((C \varphi_i, C \varphi_j)_Y)_{1 \leq i \leq N, 1 \leq j \leq N}.$$

It is not difficult to prove that (1) is equivalent to the fact that \mathbb{G} is invertible.

The orthogonal projection operator Q_+

Lemma

Assume that property (1) holds true. Then, for any $y \in Y$, then Q_+y is defined by

$$Q_+y = \sum_{i=1}^N (y, \eta_i)_Y C\varphi_i,$$

where

$$\eta_i = \sum_{j=1}^N \alpha_{ij} C\varphi_j$$

and

$$(\alpha_{ij})_{1 \leq i \leq N, 1 \leq j \leq N} = \mathbb{G}^{-1}.$$

From L_+ to L

The (finite dimensional) operator $C_+ \in \mathcal{L}(X_+, Y_+)$ satisfies $C_+ \varphi_k = C \varphi_k$ for any $k \in \{1, \dots, N\}$. Note that C_+ is nothing but the identity matrix when we choose as basis for X_+ and Y_+ respectively $(\varphi_k)_{1 \leq k \leq N}$ and $(C \varphi_k)_{1 \leq k \leq N}$. Therefore, using these bases, $A_+ + L_+ C_+$ is a Hurwitz matrix provided $\text{diag}(\lambda_1, \dots, \lambda_N) + L_+$ is Hurwitz. It is thus sufficient to take $L_+ = -\sigma I$ with

$$\sigma > \text{Re } \lambda_1$$

to ensure the stability of $A_+ + L_+ C_+$.

The corresponding operator $L \in \mathcal{L}(Y, X)$ for every $y \in Y$

$$Ly = L_+ Q_+ y = L_+ \left(\sum_{i=1}^N (y, \eta_i)_Y C \varphi_i \right) = -\sigma \sum_{i=1}^N (y, \eta_i)_Y \varphi_i,$$

and, following Theorem 6, $A + LC$ generates an exponentially stable semigroup.

Goal

Under these assumptions, there is a unique unstable eigenvalue $\lambda_1 = \lambda_1^0 - 1$ (where $\lambda_1^0 \in \mathbb{R}$ satisfies $F(\lambda_1^0) = 1$). Computing numerically this value, we obtain that $\lambda_1 = 0.239$.

The observation operator $C \in \mathcal{L}(X, Y)$ is given by

$$C\varphi = \varphi|_{(0, a^*) \times (\pi/3, 2\pi/3)}, \quad \forall \varphi \in X$$

where $X = L^2((0, a^*) \times (0, \pi))$ and $Y = L^2((0, a^*) \times (\pi/3, 2\pi/3))$.

In order to estimate $p(T)$, we use the **observer** designed previously.

As the unstable space is the one-dimensional space

$$X_+ = \text{Ker}(A - \lambda_1) = \text{Span}\{\varphi_1\} = \text{Span}\{\varphi_1^0(a)\varphi_1^D(x)\},$$

the observer involves the stabilizing output injection operator L defined by

$$Ly = -\sigma (y, \eta_1)_Y \varphi_1 \quad (y \in Y),$$

where $\sigma > \lambda_1$ (gain coefficient) and

$$\eta_1 = \alpha_{11} C\varphi_1 = \frac{C\varphi_1}{\|C\varphi_1\|_Y^2}.$$

Goal

The **observer** solves then the following system

$$\left\{ \begin{array}{l} \partial_t \widehat{p}(a, x, t) + \partial_a \widehat{p}(a, x, t) + \mu(a) \widehat{p}(a, x, t) \\ \quad - \partial_{xx} \widehat{p}(a, x, t) + \sigma (C \widehat{p}, \eta_1)_Y \varphi_1(a, x) \\ \quad = \sigma (\mathbf{y}, \eta_1)_Y \varphi_1(a, x), \\ \widehat{p}(a, 0, t) = \widehat{p}(a, \pi, t) = 0, \\ \widehat{p}(a, x, 0) = 0, \\ \widehat{p}(0, x, t) = \int_0^{a^*} \beta(a) \widehat{p}(a, x, t) da, \end{array} \right. \quad \begin{array}{l} a \in (0, a^*), x \in (0, \pi), t > 0, \\ a \in (0, a^*), t > 0, \\ a \in (0, a^*), x \in (0, \pi), \\ x \in (0, \pi), t > 0. \end{array}$$

Goal: compare $p(T)$ and $\widehat{p}(T)$

Main difficulties concerning the discretization

- Singular behavior of the coefficient μ

↪ **rescaling** the problem: introduce the auxiliary variable

$$u(a, x, t) = \frac{p(a, x, t)}{\Pi(a)} = \exp\left(\int_0^{a^*} \mu(s) ds\right) p(a, x, t)$$

$$\left\{ \begin{array}{l} \partial_t u(a, x, t) + \partial_a u(a, x, t) - \partial_{xx} u(a, x, t) = 0, \\ u(a, 0, t) = u(a, \pi, t) = 0, \\ u(a, x, 0) = u_0(a, x) = p_0(a, x) / \Pi(a), \\ u(0, x, t) = \int_0^{a^*} m(a) u(a, x, t) da, \quad \text{where } m(a) = \beta(a) \Pi(a) \end{array} \right.$$

- discretization of the renewal eq.: $p(0, x, t) = \int_0^{a^*} \beta(a) p(a, x, t) da$

$$\hookrightarrow u(0, x, n\Delta t) = \int_0^{a^*} m(a) u(a, x, (n-1)\Delta t) da$$

- presence of the extra term in the observer equation $(C\hat{p}, \eta_1)_Y \varphi_1$,
↪ **introduce** $\theta(t) = (C\Pi\hat{u}, \eta_1)_Y$ which satisfies

$$\left\{ \begin{array}{l} \dot{\theta}(t) = -(C\Pi\partial_a \hat{u}, \eta_1)_Y + (C\Pi\partial_{xx} \hat{u}, \eta_1)_Y - \sigma\theta(t) + \sigma(y, \eta_1)_Y \\ \theta(0) = 0 \end{array} \right.$$

Rescaling the open loop problem

First of all, in order to overcome the difficulties due to singular behavior of the coefficient μ , we introduce the auxiliary variable

$$u(a, x, t) = \frac{p(a, x, t)}{\Pi(a)} = \exp\left(\int_0^{a^*} \mu(s) ds\right) p(a, x, t).$$

One can easily check that u satisfies

$$\left\{ \begin{array}{ll} \partial_t u(a, x, t) + \partial_a u(a, x, t) - \partial_{xx} u(a, x, t) = 0, & a \in (0, a^*), x \in (0, \pi), t > 0, \\ u(a, 0, t) = u(a, \pi, t) = 0, & a \in (0, a^*), t > 0, \\ u(a, x, 0) = u_0(a, x), & a \in (0, a^*), x \in (0, \pi), \\ u(0, x, t) = \int_0^{a^*} m(a) u(a, x, t) da, & x \in (0, \pi), t > 0, \end{array} \right.$$

where we have set $u_0(a, x) = p_0(a, x)/\Pi(a)$ and where $m(a) = \beta(a)\Pi(a)$ stands for the maternity function.

Finite difference discretization in time

Let $u^n(a, x)$ be an approximation of $u(a, x, t^n)$, where $t^n = n\Delta t$, $0 \leq n \leq N_t$, $\Delta t = T/N_t$ is a discretization of $(0, T)$. Starting from $u^0(a, x) = u_0(a, x)$, we construct u^n for $n \geq 1$ using an Euler's backwards scheme

$$\left\{ \begin{array}{ll} \frac{u^n(a, x) - u^{n-1}(a, x)}{\Delta t} + \partial_a u^n(a, x) - \partial_{xx} u^n(a, x) = 0, & a \in (0, a^*), x \in (0, \pi), \\ u^n(a, 0) = u^n(a, \pi) = 0, & a \in (0, a^*), \\ u^0(a, x) = u_0(a, x), & a \in (0, a^*), x \in (0, \pi), \\ u^n(0, x) = \int_0^{a^*} m(a) u^{n-1}(a, x) da, & x \in (0, \pi). \end{array} \right.$$

Finite difference discretization in space

Denoting by $u_i^n(a)$ an approximation of $u^n(x_i, a)$ (where $x_i = ih = i\ell/(N_x + 1)$, with $0 \leq i \leq N_x + 1$) and using a classical centered approximation for the second order derivative in space, the above system yields

$$\begin{cases} \frac{d\mathbf{U}^n}{da}(a) + \frac{1}{h^2}\mathbb{K}\mathbf{U}^n(a) + \frac{1}{\Delta t}\mathbf{U}^n(a) = \frac{1}{\Delta t}\mathbf{U}^{n-1}(a), \\ \mathbf{U}^n(0) = \int_0^{a^*} m(a)\mathbf{U}^{n-1}(a) da, \\ \mathbf{U}^0(a) = \mathbf{U}_0(a), \end{cases}$$

where

$$\mathbf{U}^n(a) = \begin{pmatrix} u_1^n(a) \\ \vdots \\ \vdots \\ u_{N_x}^n(a) \end{pmatrix}, \mathbf{U}_0(a) = \begin{pmatrix} u_0(a, x_1) \\ \vdots \\ \vdots \\ u_0(a, x_{N_x}) \end{pmatrix}, \mathbb{K} = \begin{pmatrix} 2 & -1 & & \\ -1 & 2 & -1 & 0 \\ & \ddots & \ddots & \ddots \\ 0 & -1 & 2 & -1 \\ & & -1 & 2 \end{pmatrix}.$$

Finite difference discretization in age

We use a Crank-Nicholson scheme. Denoting by $u_i^{n,k}$ an approximation of $u_i^n(a^k)$, where $a^k = k\Delta a$, $0 \leq k \leq N_a$, $\Delta t = a^*/N_a$, and by

$$\mathbf{U}^{n,k} := \begin{pmatrix} u_1^{n,k} \\ \vdots \\ u_{N_x}^{n,k} \end{pmatrix}$$

an approximation of $\mathbf{U}^n(a^k)$, we move from age a^{k-1} to age a^k following

$$\begin{aligned} \frac{1}{\Delta a} (\mathbf{U}^{n,k} - \mathbf{U}^{n,k-1}) + \frac{1}{h^2} \mathbb{K} \left(\frac{\mathbf{U}^{n,k} + \mathbf{U}^{n,k-1}}{2} \right) + \frac{1}{\Delta t} \left(\frac{\mathbf{U}^{n,k} + \mathbf{U}^{n,k-1}}{2} \right) \\ = \frac{1}{\Delta t} \left(\frac{\mathbf{U}^{n-1,k} + \mathbf{U}^{n-1,k-1}}{2} \right), \end{aligned}$$

with the initial conditions

$$\begin{cases} \mathbf{U}^{0,k} = \mathbf{U}_0(a^k), & \forall k = 0, \dots, N_a, \\ \mathbf{U}^{n,0} = \sum_{k=0}^{N_a} \omega_k m(a^k) \mathbf{U}^{n-1,k} \simeq \int_0^{a^*} m(a) \mathbf{U}^{n-1}(a) da. \end{cases}$$

The algorithm

- 1 For $n = 0$: Initialization of $\mathbf{U}^{0,k}$
- 2 For $n = 1, \dots, N_t$:
 - $k = 0$: Initialization of $\mathbf{U}^{n,0}$ using the values of $(\mathbf{U}^{n-1,j})_{j=0}^{N_a}$:

$$\mathbf{U}^{n,0} = \sum_{k=0}^{N_a} \omega_k m(a^k) \mathbf{U}^{n-1,k}$$

- For $k = 1, \dots, N_a$, $\mathbf{U}^{n,k} = \begin{pmatrix} u_1^{n,k} \\ \vdots \\ u_{N_x}^{n,k} \end{pmatrix}$ solves the linear system
$$\mathbb{A} \mathbf{U}^{n,k} = \mathbf{b}^{n,k}$$

where

$$\mathbb{A} = \left(\Delta t + \frac{1}{2} \Delta a \right) \mathbb{I} + \frac{1}{2} \frac{\Delta t \Delta a}{h^2} \mathbb{K}$$

$$\mathbf{b}^{n,k} = \frac{\Delta a}{2} \left(\mathbf{U}^{n-1,k} + \mathbf{U}^{n-1,k-1} \right) + \left[\left(\Delta t - \frac{\Delta a}{2} \right) \mathbb{I} - \frac{\Delta t \Delta a}{2h^2} \mathbb{K} \right] \mathbf{U}^{n,k-1}$$

- $\mathbf{Y}^{n,k} = \begin{pmatrix} y_1^{n,k} \\ \vdots \\ y_N^{n,k} \end{pmatrix}$ where $y_i^{n,k} = \Pi(a^k) u_i^{n,k}$ if $\ell_1 \leq ih \leq \ell_2$ and

Discretization of the closed loop system : observer design

$$\left\{ \begin{array}{l} \partial_t \hat{p}(a, x, t) + \partial_a \hat{p}(a, x, t) + \mu(a) \hat{p}(a, x, t) \\ \quad - \partial_{xx} \hat{p}(a, x, t) + \sigma (C \hat{p}, \eta_1)_Y \varphi_1(a, x) = \sigma (\mathbf{y}, \eta_1)_Y \varphi_1(a, x), \\ \hat{p}(a, 0, t) = \hat{p}(a, \pi, t) = 0, \\ \hat{p}(a, x, 0) = 0, \\ \hat{p}(0, x, t) = \int_0^{a^*} \beta(a) \hat{p}(a, x, t) da. \end{array} \right.$$

Rescaling the problem

First of all, we introduce the auxiliary variable

$$\hat{u}(a, x, t) = \frac{\hat{p}(a, x, t)}{\Pi(a)} = \exp\left(\int_0^{a^*} \mu(s) ds\right) \hat{p}(a, x, t).$$

One can easily check that \hat{u} satisfies

$$\left\{ \begin{array}{l} \partial_t \hat{u}(a, x, t) + \partial_a \hat{u}(a, x, t) - \partial_{xx} \hat{u}(a, x, t) \\ \quad + \sigma(C\Pi\hat{u}, \eta_1)_Y v_1(a, x) = \sigma(y, \eta_1)_Y v_1(a, x), \\ \hat{u}(a, 0, t) = \hat{u}(a, \pi, t) = 0, \\ \hat{u}(a, x, 0) = 0, \\ \hat{u}(0, x, t) = \int_0^{a^*} m(a) \hat{u}(a, x, t) da, \end{array} \right.$$

where we have set $v_1(a, x) = \varphi_1(a, x)/\Pi(a)$.

Discretization of the term $(C\Pi\widehat{u}, \eta_1)_Y$

Let us introduce

$$\theta(t) = (C\Pi\widehat{u}, \eta_1)_Y.$$

Using the fact that $(C\Pi v_1, \eta_1)_Y = 1$, we remark that θ satisfies

$$\dot{\theta}(t) = - (C\Pi\partial_a\widehat{u}, \eta_1)_Y + (C\Pi\partial_{xx}\widehat{u}, \eta_1)_Y - \sigma\theta(t) + \sigma(\mathbf{y}, \eta_1)_Y.$$

Consequently,

$$\left\{ \begin{array}{l} \dot{\theta}(t) = - (C\Pi\partial_a\widehat{u}(t), \eta_1)_Y + (C\Pi\partial_{xx}\widehat{u}(t), \eta_1)_Y - \sigma\theta(t) + \sigma(\mathbf{y}(t), \eta_1)_Y, \\ \partial_t\widehat{u}(a, x, t) + \partial_a\widehat{u}(a, x, t) - \partial_{xx}\widehat{u}(a, x, t) + \sigma\theta(t)v_1(a, x) \\ \quad = \sigma(\mathbf{y}, \eta_1)_Y v_1(a, x), \\ \theta(0) = 0, \\ \widehat{u}(a, 0, t) = \widehat{u}(a, \ell, t) = 0, \\ \widehat{u}(a, x, 0) = 0, \\ \widehat{u}(0, x, t) = \int_0^{a^*} m(a)\widehat{u}(a, x, t) da. \end{array} \right.$$

Finite difference discretization in time

Let $\hat{u}^n(a, x)$ (resp. θ^n , $y^n(a, x)$) be an approximation of $\hat{u}(a, x, t^n)$ (resp. $\theta(t^n)$, $y(a, x, t^n)$), where $t^n = n\Delta t$, $0 \leq n \leq N_t$, $\Delta t = T/N_t$ is a discretization of $(0, T)$. Starting from $\theta^0 = 0$ and $\hat{u}^0(a, x) = 0$, we construct θ^n and \hat{u}^n for $n \geq 1$ using an Euler's backwards scheme

$$\left\{ \begin{array}{l} \frac{1}{\Delta t} (\theta^n - \theta^{n-1}) = - (C\Pi\partial_a\hat{u}^{n-1}, \eta_1)_Y + (C\Pi\partial_{xx}\hat{u}^{n-1}, \eta_1)_Y \\ \quad - \sigma\theta^n + \sigma(y^n, \eta_1)_Y, \\ \\ \frac{1}{\Delta t} (\hat{u}^n(a, x) - \hat{u}^{n-1}(a, x)) + \partial_a\hat{u}^n(a, x) - \partial_{xx}\hat{u}^n(a, x) + \sigma\theta^n v_1 \\ \quad = \sigma(y^n, \eta_1)_Y v_1, \\ \\ \theta^0 = 0, \\ \\ \hat{u}^n(a, 0) = \hat{u}^n(a, \ell) = 0, \\ \\ \hat{u}^0(a, x) = 0, \\ \\ \hat{u}^n(0, x) = \int_0^{a^*} m(a)\hat{u}^{n-1}(a, x) da. \end{array} \right.$$

Finite difference discretization in space

Denoting by $\widehat{u}_i^n(a)$ an approximation of $\widehat{u}^n(x_i, a)$, the above system yields

$$\left\{ \begin{array}{l} \frac{1}{\Delta t} \theta^n = \frac{1}{\Delta t} \theta^{n-1} - \sigma \theta^n - h \left(\int_0^{a^*} \Pi(a) \boldsymbol{\eta}_1(a)^T \partial_a \widehat{\mathbf{U}}^{n-1}(a) da \right) \\ \quad - \frac{1}{h} \left(\int_0^{a^*} \Pi(a) \boldsymbol{\eta}_1(a)^T \mathbb{K} \widehat{\mathbf{U}}^{n-1}(a) da \right) + \sigma h \left(\int_0^{a^*} \boldsymbol{\eta}_1(a)^T \mathbf{Y}^n(a) da \right), \\ \frac{d\widehat{\mathbf{U}}^n}{da}(a) + \frac{1}{h^2} \mathbb{K} \widehat{\mathbf{U}}^n(a) + \frac{1}{\Delta t} \widehat{\mathbf{U}}^n(a) + \sigma \theta^n \mathbf{V}_1(a) \\ \quad = \frac{1}{\Delta t} \widehat{\mathbf{U}}^{n-1}(a) + \sigma h \left(\int_0^{a^*} \boldsymbol{\eta}_1(a)^T \mathbf{Y}^n(a) da \right) \mathbf{V}_1(a), \\ \theta^0 = 0, \\ \widehat{\mathbf{U}}^n(0) = \int_0^{a^*} m(a) \widehat{\mathbf{U}}^{n-1}(a) da, \\ \widehat{\mathbf{U}}^0(a) = 0, \end{array} \right.$$

where

$$\widehat{\mathbf{U}}^n(a) = \begin{pmatrix} \widehat{u}_1^n(a) \\ \vdots \\ \widehat{u}_{N_x}^n(a) \end{pmatrix}, \mathbf{V}_1(a) = \begin{pmatrix} v_1(a, x_1) \\ \vdots \\ v_1(a, x_{N_x}) \end{pmatrix}, \boldsymbol{\eta}_1(a) = \begin{pmatrix} \eta_1(a, x_1) \\ \vdots \\ \eta_1(a, x_{N_x}) \end{pmatrix}.$$

Finite difference discretization in age

We use a Crank-Nicholson scheme. Denoting by $\widehat{u}_i^{n,k}$ an approximation of

$$\widehat{u}_i^n(a^k), \text{ where } a^k = k\Delta a, 0 \leq k \leq N_a, \Delta t = a^*/N_a, \text{ and by } \widehat{\mathbf{U}}^{n,k} := \begin{pmatrix} \widehat{u}_1^{n,k} \\ \vdots \\ \widehat{u}_{N_x}^{n,k} \end{pmatrix}$$

an approximation of $\widehat{\mathbf{U}}^n(a^k)$, we move from age a^{k-1} to age a^k following

$$\begin{aligned} \frac{1}{\Delta t}\theta^n &= \frac{1}{\Delta t}\theta^{n-1} - \sigma\theta^n - h\Delta a \left(\sum_{k=1}^{N_a} \Pi(k\Delta a) (\boldsymbol{\eta}_1^k)^T \left(\frac{\widehat{\mathbf{U}}^{n-1,k} - \widehat{\mathbf{U}}^{n-1,k-1}}{\Delta a} \right) \right) \\ &- \frac{\Delta a}{h} \left(\sum_{k=1}^{N_a} \Pi(k\Delta a) (\boldsymbol{\eta}_1^k)^T \mathbb{K} \widehat{\mathbf{U}}^{n-1,k} \right) + \sigma h \Delta a \left(\sum_{k=1}^{N_a} (\boldsymbol{\eta}_1^k)^T \mathbf{Y}^{n,k} \right), \end{aligned}$$

and

$$\begin{aligned} \frac{1}{\Delta a} \left(\widehat{\mathbf{U}}^{n,k} - \widehat{\mathbf{U}}^{n,k-1} \right) &+ \frac{1}{h^2} \mathbb{K} \left(\frac{\widehat{\mathbf{U}}^{n,k} + \widehat{\mathbf{U}}^{n,k-1}}{2} \right) + \frac{1}{\Delta t} \left(\frac{\widehat{\mathbf{U}}^{n,k} + \widehat{\mathbf{U}}^{n,k-1}}{2} \right) \\ + \sigma \theta^n \mathbf{V}_1^k &= \frac{1}{\Delta t} \left(\frac{\widehat{\mathbf{U}}^{n-1,k} + \widehat{\mathbf{U}}^{n-1,k-1}}{2} \right) + \sigma h \Delta a \left(\sum_{j=1}^{N_a} (\boldsymbol{\eta}_1^j)^T \mathbf{Y}^{n,j} \right) \mathbf{V}_1^k, \end{aligned}$$

with the initial conditions

$$\left\{ \begin{array}{l} \theta^0 = 0, \\ \hat{\mathbf{U}}^{0,k} = 0, \quad \forall k = 0, \dots, N_a, \\ \hat{\mathbf{U}}^{n,0} = \sum_{k=0}^{N_a} \omega_k m(a^k) \hat{\mathbf{U}}^{n-1,k}. \end{array} \right.$$

Here

$$\mathbf{V}_1^k = \begin{pmatrix} v_1(k\Delta a, x_1) \\ \vdots \\ v_1(k\Delta a, x_{N_x}) \end{pmatrix}, \quad \boldsymbol{\eta}_1^k = \begin{pmatrix} \eta_1(k\Delta a, x_1) \\ \vdots \\ \eta_1(k\Delta a, x_{N_x}) \end{pmatrix}, \quad \mathbf{Y}^{n,k} = \begin{pmatrix} y^n(k\Delta a, x_1) \\ \vdots \\ y^n(k\Delta a, x_{N_x}) \end{pmatrix}.$$

- 1 For $n = 0$: Initialization of θ^0 and $\widehat{\mathbf{U}}^{0,k}$ ($k = 0, \dots, N_a$) at 0.
- 2 For $n = 1, \dots, N_t$:
 - Calculate θ^n using the values of θ^{n-1} and $(\widehat{\mathbf{U}}^{n-1,j})_{j=0}^{N_a}$:

$$\theta^n = \frac{1}{1 + \sigma \Delta t} \left(\theta^{n-1} - h \Delta t \sum_{k=1}^{N_a} \Pi(k \Delta a) (\boldsymbol{\eta}_1^k)^{\mathbf{T}} \left(\widehat{\mathbf{U}}^{n-1,k} - \widehat{\mathbf{U}}^{n-1,k-1} \right) - \frac{\Delta a \Delta t}{h} \sum_{k=1}^{N_a} \Pi(k \Delta a) (\boldsymbol{\eta}_1^k)^{\mathbf{T}} \mathbb{K} \widehat{\mathbf{U}}^{n-1,k} + \sigma h \Delta a \Delta t \sum_{k=1}^{N_a} (\boldsymbol{\eta}_1^k)^{\mathbf{T}} \mathbf{Y}^{n,k} \right)$$

- $k = 0$: Initialization of $\widehat{\mathbf{U}}^{n,0}$ using the values of $(\widehat{\mathbf{U}}^{n-1,j})_{j=0}^{N_a}$:

$$\widehat{\mathbf{U}}^{n,0} = \sum_{k=0}^{N_a} \omega_k m(k \Delta a) \widehat{\mathbf{U}}^{n-1,k}$$

- For $k = 1, \dots, N_a$, solve the linear system

$$\mathbb{A} \widehat{\mathbf{U}}^{n,k} = \widehat{\mathbf{b}}^{n,k}$$

where

$$\widehat{\mathbf{b}}^{n,k} = \frac{\Delta a}{2} \left(\widehat{\mathbf{U}}^{n-1,k} + \widehat{\mathbf{U}}^{n-1,k-1} \right) + \left[\left(\Delta t - \frac{\Delta a}{2} \right) \mathbb{I} - \frac{\Delta t \Delta a}{2h^2} \mathbb{K} \right] \widehat{\mathbf{U}}^{n,k-1} - \sigma \Delta a \Delta t \theta^n \mathbf{V}_1^k + \sigma h (\Delta a)^2 \Delta t \left(\sum_{j=1}^{N_a} (\boldsymbol{\eta}_1^j)^{\mathbf{T}} \mathbf{Y}^{n,j} \right) \mathbf{V}_1^k$$

End of the algorithm

- $\widehat{\mathbf{P}}^{n,k} = \begin{pmatrix} \widehat{p}_1^{n,k} \\ \vdots \\ \widehat{p}_{N_x}^{n,k} \end{pmatrix}$ where $\widehat{p}_i^{n,k} = \Pi(a^k)\widehat{u}_i^{n,k}$.