

Wiener-Hopf Factorization for Time-Inhomogeneous Markov Chains

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Motivation

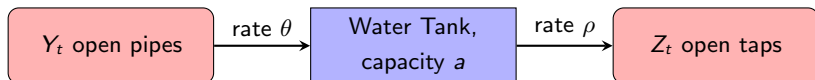
In many applications one needs to compute conditional expectations of the form

$$\mathbb{E}\left(e^{-c\tau} f(\widehat{X}_\tau) \mid X_s\right), \quad s \geq 0, \quad c \in (0, \infty),$$

where X is a Markov process, defined on a stochastic basis $(\Omega, \mathcal{F}, \mathbb{F}, \mathbb{P})$, \widehat{X} is a process related to X (usually $\widehat{X} = X$), and τ is an \mathbb{F} -stopping time.

- **Monte Carlo method**: slow and inaccurate, curse of dimensionality.
- **Feynman-Kac representation** (for \mathbb{R}^d -valued X): not easy in practice, especially if d is large or the associated integro-PDE is highly nonlinear.
- **Wiener-Hopf Factorization**: pure analytic method.

Example: A Simple Fluid Model (Rogers' 94)



- The water volume in the tank at time t , ξ_t , satisfies

$$\frac{d\xi_t}{dt} = \theta Y_t - \rho Z_t, \quad \text{if } 0 < \xi_t < a.$$

- We model $X_t = (Y_t, Z_t)$ by a finite-state time-homogeneous Markov chain.
- Let $v(y, z) := \theta y - \rho z \neq 0$. Then $\xi(t) = \xi_0 + \int_0^t v(X_s) ds$, $t \geq 0$. Let

$$\tau_\ell^+ := \inf \{u \geq 0 : \xi(u) > \ell\}, \quad \tau_\ell^- := \inf \{u \geq 0 : \xi(u) < \ell\}, \quad \ell \geq 0.$$

- τ_ℓ^+ ($\ell \in (\xi_0, a]$) is the first time the tank has ℓ amount of water.
- If $a = \infty$, τ_ℓ^- ($\ell \in [0, \xi_0)$) is the first time the tank has ℓ water amount left.
- The problem of interest is to find the **joint distribution of $(\tau_\ell^\pm, X_{\tau_\ell^\pm})$** , or

$$\mathbb{E} \left(e^{-c\tau_\ell^\pm} \mathbf{1}_{\{X_{\tau_\ell^\pm} = x'\}} \mid X_0 = x \right), \quad c > 0.$$

Example: Two Barrier Ruin Problem (Avram, Pistorius & Usabel' 03)

- The reserve process U of an insurance company is modeled by

$$U_t = u - \sum_{k=1}^{N_t} Z_k + pt, \quad t \geq 0,$$

where $u > 0$ is the initial capital, $p > 0$ is the premium rate.

- The claims $(Z_k)_{k \in \mathbb{N}}$ are i.i.d. random variables with common **phase-type** (n, β, \mathbf{B}) distribution.
- N is an independent renewal process with inter-arrival distribution of **phase-type** (m, α, \mathbf{A}) .
- Let $\tau_K := \inf\{t > 0 : U_t \notin [0, K]\}$, $K > 0$. The ruin problem is to find expressions for

$$\mathbb{E}\left(e^{-\delta\tau_K} \mathbf{1}_{\{U_{\tau_K}=K\}} \mid U_0 = u\right) \quad \text{and} \quad \mathbb{E}\left(e^{-\delta\tau_K} \mathbf{1}_{\{U_{\tau_K} \leq 0\}} \mid U_0 = x\right),$$

for any $\delta > 0$.

Example: CDS on Defaultable Stock in Markov-Modulated Models

- Let Z be a continuous-time Markov chain with finite state space $E \cup \partial$, where ∂ is an absorbing cemetery state. Let $\zeta := \inf\{t \geq 0 : Z_t = \partial\}$.
- Consider a defaultable stock with pre-default dynamics $S_t = e^{X_t}$, where

$$X_t = X_0 + \int_0^t b(Z_s) ds + \int_0^t \sigma(Z_s) dW_s, \quad t \geq 0.$$

- W is a standard Brownian motion, and is **independent** of Z .
- The default occurs when X_t reaches the barrier $\ell < X_0$.
- Consider a credit default swap written on S . The expected payment when default occurs is then given by

$$\mathbb{E}\left(h(Z_{\tau_\ell^-}) \mathbf{1}_{\{\tau_\ell^- < \zeta\}} \mid Z_0 = i, X_0 = x\right).$$

Literature on Applications

- Fluid Flow Models: Rogers '94, Rogers & Shi '94, Asmussen '95
- Ruin Problem: Avram, Pistorius, & Usabel '03, Avram & Usabel' 04
- Option Pricing: Guo & Zhang' 04, Jobert & Rogers '06, Jiang & Pistorius '08, Levendorskiĭ '08, Mijatović & Pistorius '11
- Optimal Control: Jiang & Pistorius '12
- Battery Charges, Network Loading, etc.

Theoretical Significance

- A significant aspect of the Wiener-Hopf factorization theory for Markov processes is the **probabilistic interpretation** of purely algebraic factorizations of linear operators that are related to these processes.
- The results of the factorization are interpreted in terms of functionals of appropriately **time-changed** processes that are related to the underlying Markov process.
- The Wiener-Hopf factorization is a vital component in the theory of Markov processes path decomposition or splitting time theorems.

Time-Inhomogeneous Wiener-Hopf Factorization

- The various existing forms of the Wiener-Hopf factorization for Markov chains, strong Markov processes, Lévy processes, and Markov additive process, have been obtained and applied only in the **time-homogeneous** case.
- However, there are abundant real life dynamical systems that are modeled in terms of **time-inhomogeneous** processes, and yet the corresponding Wiener-Hopf factorization theory is not available for those models.
- Our research is aimed at developing and applying the Wiener-Hopf factorization framework for a large class of time-inhomogeneous processes.

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Spaces of Matrices

- For any $n \in \mathbb{N}$, let $\mathcal{Q}(n)$ be the set of $n \times n$ generator matrices, i.e.,

$$Q(i, j) \geq 0, \quad i \neq j; \quad \sum_{j \in E} Q(i, j) \leq 0, \quad \text{for } Q \in \mathcal{Q}(n).$$

- Let $\mathcal{Q}_0(n) \subset \mathcal{Q}(n)$ be the set of **irreducible** $n \times n$ generator matrices.
- $Q \in \mathcal{Q}_0(n)$ is called **recurrent** if all its rows sum up to zero; otherwise, Q is called **transient**.
- For any $n, n' \in \mathbb{N}$, let $\mathcal{P}(n, n')$ be the set of $n \times n'$ matrices whose rows are sub-probability vectors.

The Partition of State Space

- Let \mathbf{E} be a finite state space with cardinality m , and let ∂ be a coffin state.
- Let $v : \mathbf{E} \cup \{\partial\} \rightarrow \mathbb{R}$ and $\sigma : \mathbf{E} \cup \{\partial\} \rightarrow \mathbb{R}$ with $v(\partial) = \sigma(\partial) = 0$. Denote by

$$V := \text{diag}\{v(i), i \in \mathbf{E}\}, \quad \Sigma := \text{diag}\{\sigma(i), i \in \mathbf{E}\}.$$

Assume that v and σ are NOT equal to zero simultaneously on \mathbf{E} .

- The state space \mathbf{E} is partitioned into $\mathbf{E} = \mathbf{E}_0 \cup \mathbf{E}_+ \cup \mathbf{E}_-$, where

$$\mathbf{E}_0 := \{i \in \mathbf{E} : \sigma(i) \neq 0\}, \quad \mathbf{E}_{\pm} := \{i \in \mathbf{E} : \sigma(i) = 0, \pm v(i) > 0\}.$$

The cardinalities of \mathbf{E}_0 and \mathbf{E}_{\pm} are denoted respectively by m_0 and m_{\pm} .

The Probabilistic Model

- Let $(\Omega, \mathcal{F}, \mathbb{F}, \mathbb{P})$ be a stochastic basis, where \mathbb{F} is the augmented filtration generated by X and W .
- Let $X = (X_t)_{t \geq 0}$ be a continuous-time Markov chain taking values in $\mathbf{E} \cup \{\partial\}$, with generator $Q \in \mathcal{Q}(m)$.
- Let $W = (W_t)_{t \geq 0}$ be a standard Brownian motion, independent of X .

- Define the additive functional with noise

$$\varphi_t := \int_0^t v(X_s) ds + \int_0^t \sigma(X_s) dW_s, \quad t \geq 0.$$

- For any $\ell \geq 0$, define the passage times

$$\tau_\ell^\pm := \inf \{t \geq 0 : \pm \varphi_t > \ell\}.$$

- The **time-changed** processes $(X_{\tau_\ell^+})_{\ell \geq 0}$ and $(X_{\tau_\ell^-})_{\ell \geq 0}$ are continuous-time Markov chains with respective state spaces $\mathbf{E}_+ \cup \mathbf{E}_0 \cup \{\partial\}$ and $\mathbf{E}_- \cup \mathbf{E}_0 \cup \{\partial\}$.

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An Algebraic Factorization for Generator Matrices

Consider first the case when $\sigma \equiv 0$. Then $v \neq 0$ on \mathbf{E} , and

$$\mathbf{E}_0 = \emptyset, \quad \mathbf{E}_{\pm} = \{i \in \mathbf{E} : \pm v(i) > 0\}.$$

Theorem 2.1 (Barlow, Rogers, & Williams '80)

Let $Q \in \mathcal{Q}(m)$ and $c > 0$. Then, there exists a **unique** quadruple of matrices $(\Pi_c^+, \tilde{Q}_c^+, \Pi_c^-, \tilde{Q}_c^-)$, where $\Pi_c^{\pm} \in \mathcal{P}(m_{\mp}, m_{\pm})$ and $\tilde{Q}_c^{\pm} \in \mathcal{Q}(m_{\pm})$, such that

$$V^{-1}(Q - cI) = \begin{pmatrix} I^+ & \Pi_c^- \\ \Pi_c^+ & I^- \end{pmatrix} \begin{pmatrix} \tilde{Q}_c^+ & 0 \\ 0 & -\tilde{Q}_c^- \end{pmatrix} \begin{pmatrix} I^+ & \Pi_c^- \\ \Pi_c^+ & I^- \end{pmatrix}^{-1}, \quad (2.1)$$

where I (respectively, I^{\pm}) is the $m \times m$ (respectively, $m_{\pm} \times m_{\pm}$) identity matrix.

Probabilistic Interpretation of Matrix Wiener-Hopf

Theorem 2.2 (Barlow, Rogers, & Williams '80)

$$\Pi_c^\pm(i, j) = \mathbb{E}\left(e^{-c\tau_0^\pm} \mathbf{1}_{\{X_{\tau_0^\pm}=j\}} \mid X_0 = i\right), \quad i \in \mathbf{E}_\mp, \quad j \in \mathbf{E}_\pm, \quad (2.2)$$

$$e^\ell \tilde{Q}_c^\pm(i, j) = \mathbb{E}\left(e^{-c\tau_\ell^\pm} \mathbf{1}_{\{X_{\tau_\ell^\pm}=j\}} \mid X_0 = i\right), \quad i, j \in \mathbf{E}_\pm, \quad \ell > 0. \quad (2.3)$$

- \tilde{Q}_c^\pm is the generator matrix of the Markov chain obtained by first killing X at rate c , and then applying the time-change $(\tau_\ell^\pm)_{\ell \geq 0}$.
- If $c = 0$, the factorization (2.1) still holds true, but **not necessarily unique** when row sums of Q are all zero. In this case, the quadruple $(\Pi_0^+, \tilde{Q}_0^+, \Pi_0^-, \tilde{Q}_0^-)$ given in Theorem 2.2 is not the only solution to (2.1).
- By Theorems 2.1 and 2.2, in order to compute expectation of the form (2.2) or (2.3), one only needs to solve the algebraic equation (2.1).

Computing Π_c^\pm and \tilde{Q}_c^\pm

- The “plus” and “minus” parts in (2.1) can be separated as

$$V^{-1}(Q - cI) \begin{pmatrix} I^+ \\ \Pi_c^+ \end{pmatrix} = \begin{pmatrix} I^+ \\ \Pi_c^+ \end{pmatrix} \tilde{Q}_c^+, \quad (2.4)$$

$$V^{-1}(Q - cI) \begin{pmatrix} I^- \\ \Pi_c^- \end{pmatrix} = - \begin{pmatrix} I^- \\ \Pi_c^- \end{pmatrix} \tilde{Q}_c^-. \quad (2.5)$$

Hence, (Π_c^+, \tilde{Q}_c^+) and (Π_c^-, \tilde{Q}_c^-) can be computed independently.

- Write Q and V in their respective block forms:

$$Q = \begin{pmatrix} A & B \\ C & D \end{pmatrix} \quad \text{and} \quad V = \begin{pmatrix} V^+ & \mathbf{0} \\ \mathbf{0} & V^- \end{pmatrix}.$$

Then, $\tilde{Q}_c^+ = (V^+)^{-1}((A - cI^+) + B\Pi_c^+)$, and Π_c^+ satisfies following **algebraic Riccati equation**

$$-(V^-)^{-1}(D - cI^-)\Pi_c^+ + \Pi_c^+(V^+)^{-1}(A - cI^+) + \Pi_c^+(V^+)^{-1}B\Pi_c^+ - (V^-)^{-1}C = \mathbf{0}.$$

Similar computation can be done for (Π_c^-, \tilde{Q}_c^-) .

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Existence and Uniqueness

Consider the general case when σ is any function on \mathbf{E} . Let $Q \in \mathcal{Q}_0(m)$, namely, Q is **irreducible**.

Theorem 2.3 (Jiang & Pistorius '08)

Let $Q \in \mathcal{Q}_0(m)$ be **transient**. Then there exists a **unique** quadruple of matrices $(\Pi^+, \tilde{Q}^+, \Pi^-, \tilde{Q}^-)$, where $\Pi^\pm \in \mathcal{P}(m_\mp, m_\pm + m_0)$ and $\tilde{Q}^\pm \in \mathcal{Q}(m_\pm + m_0)$, such that

$$\frac{1}{2}\Sigma^2 \begin{pmatrix} I^+ & \mathbf{0} \\ \mathbf{0} & I_0 \\ & \Pi^+ \end{pmatrix} (\tilde{Q}^+)^2 - V \begin{pmatrix} I^+ & \mathbf{0} \\ \mathbf{0} & I_0 \\ & \Pi^+ \end{pmatrix} \tilde{Q}^+ + Q \begin{pmatrix} I^+ & \mathbf{0} \\ \mathbf{0} & I_0 \\ & \Pi^+ \end{pmatrix} = \mathbf{0}, \quad (2.6)$$

$$\frac{1}{2}\Sigma^2 \begin{pmatrix} & \Pi^- \\ I_0 & \mathbf{0} \\ \mathbf{0} & I^- \end{pmatrix} (\tilde{Q}^-)^2 + V \begin{pmatrix} & \Pi^- \\ I_0 & \mathbf{0} \\ \mathbf{0} & I^- \end{pmatrix} \tilde{Q}^- + Q \begin{pmatrix} & \Pi^- \\ I_0 & \mathbf{0} \\ \mathbf{0} & I^- \end{pmatrix} = \mathbf{0}. \quad (2.7)$$

Probabilistic Interpretation

Theorem 2.4 (Jiang & Pistorius '08)

$$\begin{aligned}\Pi^\pm(i, j) &= \mathbb{P}\left(X_{\tau_0^\pm} = j, \tau_0^\pm < \infty \mid X_0 = i\right), \quad i \in \mathbf{E}_\mp, j \in \mathbf{E}_\pm \cup \mathbf{E}_0, \\ e^{\ell \tilde{Q}^\pm}(i, j) &= \mathbb{P}\left(X_{\tau_\ell^\pm} = j, \tau_\ell^\pm < \infty \mid X_0 = i\right), \quad i, j \in \mathbf{E}_\pm, \ell > 0.\end{aligned}$$

- When $\sigma \equiv 0$ (so that $\mathbf{E}_0 = \emptyset$ and $\mathbf{E} = \mathbf{E}_+ \cup \mathbf{E}_-$), (2.6)–(2.7) reduce to the “plus” and “minus” part of (2.1), given as in (2.4)–(2.5), respectively.
- When σ is constant (so that $\mathbf{E}_+ = \mathbf{E}_- = \emptyset$ and $\mathbf{E} = \mathbf{E}_0$), (2.6)–(2.7) reduce to Kennedy & Williams '90.

Application: Two-Sided Exit Problem

- The two-sided exit problem of φ from the interval $[k, \ell]$, $-\infty < k < \ell < \infty$, is to find the **joint distribution** of the position (X_τ, φ_τ) at the first-exit time

$$\tau = \tau_{k,\ell} := \inf \{t \geq 0 : \varphi_t \notin [k, \ell]\}.$$

- Such problem can be solved explicitly in terms of the output of the Wiener-Hopf factorization $(\Pi^+, \tilde{Q}^+, \Pi^-, \tilde{Q}^-)$.
- To this end, define

$$W^+ := \begin{pmatrix} I^+ & \mathbf{0} \\ \mathbf{0} & I_0 \\ & \Pi^+ \end{pmatrix}, \quad W^- := \begin{pmatrix} \Pi^- & \\ I_0 & \mathbf{0} \\ \mathbf{0} & I^- \end{pmatrix}, \quad J^+ := \begin{pmatrix} I^+ & \mathbf{0} \\ \mathbf{0} & I_0 \\ \mathbf{0} & \end{pmatrix}, \quad J^- := \begin{pmatrix} \mathbf{0} & \\ I_0 & \mathbf{0} \\ \mathbf{0} & I^- \end{pmatrix}$$

$$Z^+ := \begin{pmatrix} \mathbf{0} & I_0 \\ \Pi^+ & \end{pmatrix} e^{\tilde{Q}^+(\ell-k)}, \quad Z^- := \begin{pmatrix} \Pi^- & \\ \mathbf{0} & I_0 \end{pmatrix} e^{\tilde{Q}^-(\ell-k)}.$$

Application: Two-Sided Exit Problem (Cont'd)

$$\Psi^+(a) := \left(\mathbf{W}^+ e^{\tilde{\mathbf{Q}}^+(\ell-a)} - \mathbf{W}^- e^{\tilde{\mathbf{Q}}^-(a-k)} \mathbf{Z}^+ \right) (\mathbf{I}_0^+ - \mathbf{Z}^- \mathbf{Z}^+)^{-1},$$

$$\Psi^-(a) := \left(\mathbf{W}^- e^{\tilde{\mathbf{Q}}^-(\ell-a)} - \mathbf{W}^+ e^{\tilde{\mathbf{Q}}^+(a-k)} \mathbf{Z}^- \right) (\mathbf{I}_0^- - \mathbf{Z}^+ \mathbf{Z}^-)^{-1},$$

$$\Psi_0(s, a) := \left(e^{sa} \mathbf{I} - e^{s\ell} \Psi^+(a) \mathbf{J}^+ - e^{sk} \Psi^-(a) \mathbf{J}^- \right) \left(-\frac{1}{2} \Sigma^2 s^2 - \mathbf{V}s - \mathbf{Q} \right)^{-1}.$$

Proposition 2.5 (Jiang & Pistorius '08)

Let h^\pm and h be real-valued functions on $\mathbf{E}_0 \cup \mathbf{E}_\pm$ and \mathbf{E} , respectively. For any $a \in [k, \ell]$ and $i \in \mathbf{E}$,

$$\mathbb{E} \left(h^+(X_\tau) \mathbf{1}_{\{\varphi_\tau = \ell\}} \mathbf{1}_{\{\tau < \zeta\}} \mid X_0 = i, \varphi_0 = a \right) = \mathbf{e}_i^\top \Psi^+(a) h^+,$$

$$\mathbb{E} \left(h^-(X_\tau) \mathbf{1}_{\{\varphi_\tau = k\}} \mathbf{1}_{\{\tau < \zeta\}} \mid X_0 = i, \varphi_0 = a \right) = \mathbf{e}_i^\top \Psi^-(a) h^-,$$

$$\mathbb{E} \left(e^{s\varphi_\zeta} h(X_{\zeta-}) \mathbf{1}_{\{\zeta < \tau\}} \mid X_0 = i, \varphi_0 = a \right) = \mathbf{e}_i^\top \Psi_0(s, a) \mathbf{H}(-\mathbf{Q}) \mathbf{1}.$$

where $\mathbf{H} := \text{diag}\{h(i), i \in \mathbf{E}\}$.

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Highly Non-trivial Extension to Time-Inhomogeneous Framework

- Let X be a **time-inhomogeneous** Markov chain taking values in $\mathbf{E} \cup \partial$, with generator function $Q_t \in \mathcal{Q}(m)$, $t \geq 0$.
- The factorization of $V^{-1}(Q_t - cI)$ can be done for each $t \geq 0$ **separately**, exactly as described in Theorem 2.1.
- However, the resulting matrices $\Pi_c^\pm(t)$ and $\tilde{Q}_c^\pm(t)$, $t \geq 0$, do not have similar **probabilistic interpretations** as shown in Theorem 2.2.
- In particular, they are not useful for computing expectations of the form

$$\mathbb{E} \left(e^{-c\tau_\ell^\pm(s)} \mathbf{1}_{\{X_{\tau_\ell^\pm(s)}=j\}} \mid X_s = i \right), \quad (3.1)$$

where

$$\tau_\ell^\pm(s) := \inf \left\{ t \geq s : \int_s^t v(X_u) du > \ell \right\}.$$

First Attempt: Piecewise Constant Generator

- Bielecki, Cialenco, Gong, & Huang '18: Wiener-Hopf Factorization for time-inhomogeneous Markov chain X with **piecewise-constant** generator Λ :

$$\Lambda_t = \Lambda_k, \quad s_{k-1} \leq t < s_k, \quad k = 1, \dots, n; \quad \Lambda_t = \Lambda_{n+1}, \quad t \geq s_n,$$

where $\Lambda_1, \dots, \Lambda_{n+1} \in \mathcal{Q}(m)$, $n \in \mathbb{N}$, $0 = s_0 < s_1 < \dots < s_n$.

- Time-inhomogeneous Markov chains with piecewise-constant generators are adequate models for, among others, seasonal phenomena, Erlang loss systems with moving boundaries, or structural breaks in credit migrations.
- The main goal is to apply the Wiener-Hopf factorization technique to compute expectations of the form (3.1).
- This special setup allows to use some appropriately tailored and original randomization techniques.

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The Time-Inhomogeneous Generator

In Bielecki, Cheng, Cialenco, & Gong '19, we study Wiener-Hopf Factorization for time-inhomogeneous Markov chain with **continuous bounded** generators.

- Let \mathbf{E} be a finite set with $|\mathbf{E}| = m > 1$.
- Let $(\Lambda_s)_{s \in \mathbb{R}_+}$, $\mathbb{R}_+ := [0, \infty)$, where $\Lambda_s \in \mathcal{Q}(m)$.
- **Main Assumptions:**
 - ◇ $\exists K \in (0, \infty)$, such that $|\Lambda_s(i, j)| \leq K$, for all $i, j \in \mathbf{E}$ and $s \in \mathbb{R}_+$;
 - ◇ $(\Lambda_s)_{s \in \mathbb{R}_+}$, considered as a mapping from \mathbb{R}_+ to the set of $m \times m$ generator matrices, is **continuous** with respect to s on \mathbb{R}_+ .

The Partition of State Space

- Recall $v : \mathbf{E} \rightarrow \mathbb{R} \setminus \{0\}$, $V := \text{diag}\{v(i) : i \in \mathbf{E}\}$.
- Assume that both \mathbf{E}_+ and \mathbf{E}_- are non-empty, where

$$\mathbf{E}_+ := \{i \in \mathbf{E} : v(i) > 0\} \quad \text{and} \quad \mathbf{E}_- := \{i \in \mathbf{E} : v(i) < 0\}.$$

- Accordingly, we write Λ_s and V in the block form :

$$\Lambda_s = \begin{matrix} & \mathbf{E}_+ & \mathbf{E}_- \\ \mathbf{E}_+ & \begin{pmatrix} A_s & B_s \\ C_s & D_s \end{pmatrix} \\ \mathbf{E}_- & \end{matrix}, \quad V = \begin{matrix} & \mathbf{E}_+ & \mathbf{E}_- \\ \mathbf{E}_+ & \begin{pmatrix} V_+ & 0 \\ 0 & V_- \end{pmatrix} \\ \mathbf{E}_- & \end{matrix}.$$

- Denote by $\mathcal{X} := \mathbb{R}_+ \times \mathbf{E}$ and $\mathcal{X}_\pm := \mathbb{R}_+ \times \mathbf{E}_\pm$.
- Define $\tilde{\Lambda} : L^\infty(\mathcal{X}) \rightarrow L^\infty(\mathcal{X})$, associated with $(\Lambda_s)_{s \in \mathbb{R}_+}$, by

$$(\tilde{\Lambda}g)(s, i) := (\Lambda_s g(s, \cdot))(i), \quad (s, i) \in \mathcal{X},$$

and similarly for \tilde{A} , \tilde{B} , \tilde{C} , and \tilde{D} .

The Time-Inhomogeneous Markov Family

- Let Ω be the collection of \mathbf{E} -valued càdlàg functions on \mathbb{R}_+ .
- There exists a **standard** Markov family

$$\mathcal{M} := \{(\Omega, \mathcal{F}, \mathbb{F}_s, (X_t)_{t \in [s, \infty)}, \mathbb{P}_{s,i}), (s, i) \in \mathcal{X}\},$$

such that the associated evolution system $\mathbf{U} := (U_{s,t})_{0 \leq s \leq t < \infty}$, defined by

$$(U_{s,t}f)(i) := \mathbb{E}_{s,i}(f(X_t)), \quad 0 \leq s \leq t < \infty, \quad i \in \mathbf{E},$$

for any $f : \mathbf{E} \rightarrow \mathbb{R}$, admits the generator $\mathbf{\Lambda}$, namely

$$\lim_{h \downarrow 0} \frac{1}{h} ((U_{s,s+h}f)(i) - f(i)) = (\Lambda_s f)(i), \quad \text{for any } (s, i) \in \mathcal{X}.$$

- Define the passage times

$$\tau_\ell^\pm(s) := \inf \left\{ t \in [s, \infty) : \pm \int_s^t v(X_u) du > \ell \right\}.$$

Main Goal

- **Goal:** To derive a **Wiener-Hopf type method** for computing expectations of the form

$$\mathbb{E}_{s,i} \left(g(\tau_\ell^\pm, X_{\tau_\ell^\pm}) \right),$$

for $g \in L^\infty(\mathcal{X})$, $\ell \in \mathbb{R}_+$, and $(s, i) \in \mathcal{X}$.

- Note that $X_{\tau_\ell^\pm(s)} \in \mathbf{E}_\pm \cup \{\partial\}$. Hence, it is sufficient to compute expectations of the following form

$$\mathbb{E}_{s,i} \left(g^\pm(\tau_\ell^\pm, X_{\tau_\ell^\pm}) \right), \tag{3.2}$$

for $g^\pm \in L^\infty(\mathcal{X}_\pm)$, $\ell \in \mathbb{R}_+$, and $(s, i) \in \mathcal{X}$.

Some Important Operators

- $J^\pm : L^\infty(\mathcal{X}_\pm) \rightarrow L^\infty(\mathcal{X}_\mp)$ is defined as

$$(J^\pm g^\pm)(s, i) := \mathbb{E}_{s, i} \left(g^\pm \left(\tau_0^\pm, X_{\tau_0^\pm} \right) \right), \quad (s, i) \in \mathcal{X}_\mp.$$

- For any $\ell \in \mathbb{R}_+$, $\mathcal{P}_\ell^\pm : L^\infty(\mathcal{X}_\pm) \rightarrow L^\infty(\mathcal{X}_\pm)$ is defined as

$$(\mathcal{P}_\ell^\pm g^\pm)(s, i) := \mathbb{E}_{s, i} \left(g^\pm \left(\tau_\ell^\pm, X_{\tau_\ell^\pm} \right) \right), \quad (s, i) \in \mathcal{X}_\pm.$$

- For any $(s, i) \in \mathcal{X}_\pm$, we define

$$(G^\pm g^\pm)(s, i) := \lim_{\ell \rightarrow 0^+} \frac{1}{\ell} (\mathcal{P}_\ell^\pm g^\pm(s, i) - g^\pm(s, i)),$$

for any $g^\pm \in C_0(\mathcal{X}_\pm)$ such that the above limit exists and is finite.

- It can be shown that any expectation of the form (3.2) can be represented in terms of the operators J^\pm and \mathcal{P}_ℓ^\pm .

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Existence and Uniqueness

Theorem 3.1 (Bielecki, Cialenco, Cheng, & Gong '19)

There exists a **unique** quadruple of operators (S^+, H^+, S^-, H^-) which solves the following operator equation: for any $g^\pm \in C_0^1(\mathcal{X}_\pm)$,

$$V^{-1} \left(\frac{\partial}{\partial s} + \tilde{\Lambda} \right) \begin{pmatrix} I^+ & S^- \\ S^+ & I^- \end{pmatrix} \begin{pmatrix} g^+ \\ g^- \end{pmatrix} = \begin{pmatrix} I^+ & S^- \\ S^+ & I^- \end{pmatrix} \begin{pmatrix} H^+ & 0 \\ 0 & -H^- \end{pmatrix} \begin{pmatrix} g^+ \\ g^- \end{pmatrix},$$

subject to the conditions below:

- (a) $S^\pm : C_0(\mathcal{X}_\pm) \rightarrow C_0(\mathcal{X}_\mp)$ is a bounded operator such that
- (i) for any $g^\pm \in C_c(\mathcal{X}_\pm)$ with $\text{supp } g^\pm \subset [0, \eta_{g^\pm}] \times \mathbf{E}_\pm$ for some constant $\eta_{g^\pm} \in (0, \infty)$, we have $\text{supp } S^\pm g^\pm \subset [0, \eta_{g^\pm}] \times \mathbf{E}_\mp$;
 - (ii) for any $g^\pm \in C_0^1(\mathcal{X}_\pm)$, we have $S^\pm g^\pm \in C_0^1(\mathcal{X}_\mp)$.
- (b) H^\pm is the **strong generator** of a strongly continuous positive contraction semigroup $(Q_\ell^\pm)_{\ell \in \mathbb{R}_+}$ on $C_0(\mathcal{X}_\pm)$ with domain $\mathcal{D}(H^\pm) = C_0^1(\mathcal{X}_\pm)$.

Probabilistic Interpretation

Theorem 3.2 (Bielecki, Cialenco, Cheng, & Gong '19)

For any $g^\pm \in C_0(\mathcal{X}_\pm)$, we have

$$S^\pm g^\pm = J^\pm g^\pm \quad \text{and} \quad Q_\ell^\pm g^\pm = P_\ell^\pm g^\pm, \quad \text{for any } \ell \in \mathbb{R}_+,$$

Moreover, G^\pm is the **strong generator** of $(P_\ell^\pm)_{\ell \in \mathbb{R}_+}$ on $C_0(\mathcal{X}_\pm)$ with domain $\mathcal{D}(G^\pm) = C_0^1(\mathcal{X}_\pm)$.

From Theorems 3.1 and 3.2, we see that (J^+, G^+, J^-, G^-) is the unique quadruple of operators, subject to conditions (a) and (b), that solves the operator equation: for any $g^\pm \in C_0^1(\mathcal{X}_\pm)$,

$$V^{-1} \left(\frac{\partial}{\partial s} + \tilde{\Lambda} \right) \begin{pmatrix} I^+ & J^- \\ J^+ & I^- \end{pmatrix} \begin{pmatrix} g^+ \\ g^- \end{pmatrix} = \begin{pmatrix} I^+ & J^- \\ J^+ & I^- \end{pmatrix} \begin{pmatrix} G^+ & 0 \\ 0 & -G^- \end{pmatrix} \begin{pmatrix} g^+ \\ g^- \end{pmatrix}. \quad (3.3)$$

Remarks

- $\partial/\partial s + \tilde{Q}$ is the generator of the following **time-homogenized** process \tilde{X} on the enlarged probability space $\tilde{\Omega} := \mathbb{R}_+ \times \Omega$:

$$\tilde{X}_t(\tilde{\omega}) = \tilde{X}_t((s, \omega)) = (s + t, X_{t+s}(\omega)).$$

The equations (3.3) can be regarded as the Wiener-Hopf factorization corresponding to the **time-homogeneous** Markov process \tilde{X} . Note that this is not a direct application of **Barlow, Rogers, & Williams '80** since the state space of \tilde{X} is no longer a finite space.

- The operators J^\pm and G^\pm are counterparts of the matrices Π_c^\pm and \tilde{Q}_c^\pm in Theorem 2.1. When X is a **time-homogeneous** Markov chain with generator matrix Q , by taking $g^\pm(s, i) = e^{-cs} \mathbf{1}_{\{i=k\}}$, for $c \geq 0$ and $k \in \mathbf{E}_\pm$, we reduce (3.3) to the time-homogeneous factorization (2.1).

Computing the Operators J^\pm and G^\pm

- By Theorems 3.1 and 3.2, for any $g^\pm \in C_0^1(\mathcal{X}_\pm)$, $J^\pm g^\pm$ and $G^\pm g^\pm$ (and thus $\mathcal{P}_\ell^\pm g^\pm$, for any $\ell \in \mathbb{R}_+$) can be computed from the Wiener-Hopf equation (3.3) subject to conditions (a) and (b).
- Note that (3.3) can be decomposed into:

$$V^{-1} \left(\frac{\partial}{\partial s} + \tilde{\Lambda} \right) \begin{pmatrix} I^+ \\ J^+ \end{pmatrix} g^+ = \begin{pmatrix} I^+ \\ J^+ \end{pmatrix} G^+ g^+, \quad g^+ \in C_0^1(\mathcal{X}_+), \quad (3.4)$$

$$V^{-1} \left(\frac{\partial}{\partial s} + \tilde{\Lambda} \right) \begin{pmatrix} I^- \\ J^- \end{pmatrix} g^- = - \begin{pmatrix} I^- \\ J^- \end{pmatrix} G^- g^-, \quad g^- \in C_0^1(\mathcal{X}_-).$$

Hence, one can compute $J^+ g^+$ and $G^+ g^+$ (and thus $\mathcal{P}_\ell^+ g^+$) separately from $J^- g^-$ and $G^- g^-$ (and thus $\mathcal{P}_\ell^- g^-$).

Computing the Operators J^\pm and G^\pm (Cont'd)

- Using the block form of $\tilde{\Lambda}$ and V , (3.4) can be written as

$$(V^+)^{-1} \left(\frac{\partial}{\partial s} + \tilde{A} + \tilde{B}J^+ \right) g^+ = G^+ g^+, \quad (3.5)$$

$$(V^-)^{-1} \left(\frac{\partial}{\partial s} J^+ + \tilde{C} + \tilde{D}J^+ \right) g^+ = J^+ G^+ g^+. \quad (3.6)$$

- From (3.5), we see that G^+ is determined by J^+ .
- Furthermore, from (3.5) and (3.6) we deduce the following operator Riccati equation for J^+ :

$$\left(J^+ (V^+)^{-1} \tilde{B} J^+ + J^+ (V^+)^{-1} \left(\frac{\partial}{\partial s} + \tilde{A} \right) - (V^-)^{-1} \left(\frac{\partial}{\partial s} + \tilde{D} \right) J^+ - (V^-)^{-1} \tilde{C} \right) g^+ = 0.$$

- In conclusion, solving (3.4) for G^+ and J^+ boils down to solving the above Riccati equation for J^+ .

Computing the Operators J^\pm and G^\pm (Cont'd)

- In practice, take $g_j^\pm \in C_0^1(\mathcal{X}_\pm)$ with

$$g_j^\pm(s, i) := e^{-cs} \mathbf{1}_{\{j\}}(i), \quad (s, i) \in \mathcal{X}_\pm,$$

for any $c > 0$ and $j \in \mathbf{E}_\pm$.

- By solving $J^\pm g^\pm$ and $G^\pm g^\pm$ for such g^\pm , we obtain the **Laplace transform** for the pair $(\tau_\ell^\pm, X_{\tau_\ell^\pm})$.
- We then perform the inverse Laplace transform with respect to c to obtain the joint distribution of $(\tau_\ell^\pm, X_{\tau_\ell^\pm})$ under $\mathbb{P}_{s,i}$, which enables us to compute the expectations (3.2) for any $g^\pm \in L^\infty(\mathcal{X}_\pm)$.

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Future Research

- Noisy Wiener-Hopf factorization for time-inhomogeneous Markov chains
- Wiener-Hopf factorization for time-homogeneous/inhomogeneous strong Markov processes with general state space

The Wiener-Hopf factorization for time-homogeneous strong Markov processes, with general state spaces, was studied by [Williams '08](#), but only for the associated **resolvent operators**.

- **Connection with the classical Wiener-Hopf factorization for Lévy processes and Markov additive processes**
- Wiener-Hopf factorization for time-homogeneous Lévy processes
- Wiener-Hopf factorization for time-homogeneous Markov additive processes
- Ultimate Goal: a **unifying framework** in terms of functional of time-changed processes and in terms of algebraic factorizations of linear operators, which encompasses all the processes above.

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THANK YOU FOR YOUR ATTENTION!