Asymptotic periodic analysis of cyclic stochastic fluid flows with time-varying transition rates

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Outline



2 Asymptotic periodic distribution





Motivation

- We are interested in Stochastic Fluid Models (SFMs) with *time-varying parameters* so that we can model a wider range of problems.
- We consider a Cyclic Stochastic Fluid Model (CSFM) analysed in Margolius and O'Reilly (2016).
- We extend the analysis to the study of its asymptotic periodic distribution.
- We also study some other CSFMs of interest, with boundaries.

Cyclic Stochastic Fluid Model (CSFM)

- CSFM $\{(\widehat{X}(t), J(t)) : t \ge 0\}$ is a process with level variable $X(t) \ge 0$ and phase variable $J(t) \in S$
- $\{J(t) : t \ge 0\}$ is a continuous-time Markov chain with
- time-varying generator $T(t) = [T(t)_{ij}]_{i,j \in S}$
- cycle of length 1 such that T(t) = T(t+1) for all $t \ge 0$
- some finite irreducible state space ${\cal S}$
- fluid rates in $c_i \in \mathbb{R}$, for all $i \in S$, such that

$$\begin{aligned} & \frac{d}{dt}X(t) = c_{J(t)} & \text{when } X(t) > 0, \\ & \frac{d}{dt}X(t) = \max(0, c_{J(t)}) & \text{when } X(t) = 0. \end{aligned}$$

Example (Margolius and O'Reilly, 2016)

Let

$$T(t) = \begin{bmatrix} * & 0 & a_2(t) & 0 & a_3(t) & 0 & 0 & b_1(t) \\ 0 & * & a_1(t) & 0 & 0 & a_3(t) & 0 & b_2(t) \\ b_2(t) & b_1(t) & * & 0 & 0 & 0 & a_3(t) & 0 \\ 0 & 0 & 0 & * & a_1(t) & a_2(t) & 0 & b_3(t) \\ b_3(t) & 0 & 0 & b_1(t) & * & 0 & a_2(t) & 0 \\ 0 & b_3(t) & 0 & b_2(t) & 0 & * & a_1(t) & 0 \\ 0 & 0 & b_3(t) & 0 & b_2(t) & b_1(t) & * & 0 \\ a_1(t) & a_2(t) & 0 & 0 & a_3(t) & 0 & 0 & * \end{bmatrix}$$

where

 $\begin{array}{rcl} a_1(t) &=& 30+25\sin(2\pi t)\\ b_1(t) &=& 30-20\sin(2\pi t)\\ a_2(t) &=& 10+7\sin(2\pi t)\\ b_2(t) &=& 40-35\sin(2\pi t)\\ a_3(t) &=& 50-35\sin(2\pi t)\\ b_3(t) &=& 80+50\sin(2\pi t). \end{array}$

Evolution of level X(t) in time (simulation)



Partitioning

Let

$$\begin{array}{rcl} \mathcal{S}_{+} &=& \{i \in \mathcal{S} : c_{i} > 0\} \\ \mathcal{S}_{-} &=& \{i \in \mathcal{S} : c_{i} < 0\} \\ \mathcal{S}_{0} &=& \{i \in \mathcal{S} : c_{i} = 0\} \end{array}$$

and partition according to $\mathcal{S}=\mathcal{S}_+\cup\mathcal{S}_-\cup\mathcal{S}_0,$ as

$$T(t) = \begin{bmatrix} T_{++}(t) & T_{+-}(t) & T_{+0}(t) \\ T_{-+}(t) & T_{--}(t) & T_{-0}(t) \\ T_{0+}(t) & T_{00}(t) & T_{0-}(t) \end{bmatrix}$$

.

Let

$$\mathbf{C}_{+} = diag(c_{i})_{i \in \mathcal{S}_{+}}, \quad \mathbf{C}_{-} = diag(c_{i})_{i \in \mathcal{S}_{-}}.$$

Density $\pi(t; x)$ at x > 0

For , $t \in [0, 1)$, x > 0,

$$\pi(t; \mathbf{x}) = \begin{bmatrix} \pi(t; \mathbf{x})_+ & \pi(t; \mathbf{x})_- & \pi(t; \mathbf{x})_0 \end{bmatrix},$$

such that for all $t \in [0, 1)$ and $i \in S$,

$$\pi(t;x)_i = \frac{\partial}{\partial x} \sum_{n=0}^{\infty} P(X(t+n) \le x, J(t+n) = i).$$



Mass p(t) at x = 0

For *t* ∈ [0, 1),

$$p(t) = \begin{bmatrix} \mathbf{0} & p_-(t) & p_0(t) \end{bmatrix},$$

where, for all $i \in S_{-} \cup S_{0}$,

$$p(t)_i = \sum_{n=0}^{\infty} P(X(t+n) = 0, J(t+n) = i).$$



Matrix $\hat{U}_0(s, t)$

For all s, t with $0 \le s \le t$, define $U_0(s, t)$ which solves

$$\begin{array}{lll} \displaystyle \frac{\partial}{\partial t} U_0(s,t) &=& U_0(s,t) T_{00}(t) \\ \displaystyle \frac{\partial}{\partial s} U_0(s,t) &=& -T_{00}(s) U_0(s,t) \\ \displaystyle U_0(s,s) &=& \mathbf{I}. \end{array}$$

where $U_0(s, t) = \mathbf{0}$ for t < s, and

$$\hat{U}_0(s,t)=\sum_{n=0}^{\infty}U_0(s,t+n).$$



Physical interpretation of $\hat{U}_0(s, t)$

For all $i, j \in S_0$,

 $[U_0(s,t)]_{ij} = P[J(t) = j \mid J(s) = i, J(u) \in \mathcal{S}_0, s \le u \le t]$



 $[\hat{U}_0(s,t)]_{ij} = \sum_{n=0}^{\infty} P[J(t+n) = j \mid J(s) = i, J(u) \in \mathcal{S}_0, s \le u \le t+n]$



Key fluid generator $\hat{Q}(s, t)$

For all s, t, with $0 \le s, t \le 1$, define

$$\hat{Q}(\boldsymbol{s},t) = \left[egin{array}{cc} \hat{Q}_{++}(\boldsymbol{s},t) & \hat{Q}_{+-}(\boldsymbol{s},t) \ \hat{Q}_{-+}(\boldsymbol{s},t) & \hat{Q}_{--}(\boldsymbol{s},t) \end{array}
ight],$$

with

$$\begin{aligned} \hat{Q}_{++}(s,t) &= \mathbf{C}_{+}^{-1}[T_{++}(s)I(s=t) + T_{+0}(s)\hat{U}_{0}(s,t)T_{0+}(t)] \\ \hat{Q}_{--}(s,t) &= \mathbf{C}_{-}^{-1}[T_{--}(s)I(s=t) + T_{-0}(s)\hat{U}_{0}(s,t)T_{0-}(t)] \\ \hat{Q}_{+-}(s,t) &= \mathbf{C}_{+}^{-1}[T_{+-}(s)I(s=t) + T_{+0}(s)\hat{U}_{0}(s,t)T_{0-}(t)] \\ \hat{Q}_{-+}(s,t) &= \mathbf{C}_{-}^{-1}[T_{-+}(s)I(s=t) + T_{-0}(s)\hat{U}_{0}(s,t)T_{0+}(t)]. \end{aligned}$$

Key matrix $\hat{\Psi}(s, t)$

For all s, t with $0 \le s \le t$, define matrix $\Psi(s, t) = [\Psi(s, t)_{ij}]_{i \in S_+, j \in S_-}$ such that, with $\theta_s(x) = \inf\{t > s : X(t) = x\}$, for all $i \in S_+, j \in S_-$,

$$\Psi(s,t)_{ij}dt = P\left[t < \theta_s(z) \le t + dt, J(\theta_s(z)) = j \mid X(s) = z, J(s) = i\right],$$

and matrix $\hat{\Psi}(s, t) = [\hat{\Psi}(s, t)_{ij}]$ such that

$$\hat{\Psi}(\boldsymbol{s},t) = \sum_{n=0}^{\infty} \Psi(\boldsymbol{s},t+n).$$

Physical interpretation of $\hat{\Psi}(s, t)$

 $\hat{\Psi}(s,t)_{ij}$ is the probability that

- given start from level 0 in phase $i \in S_+$ at time s,
- the process $\{(\widehat{X}(t), J(t)) : t \ge 0\}$ first returns to level 0
- at some time $u \in \bigcup_{n=0}^{\infty} (t + n, t + n + dt]$,

• and does so in phase $j \in S_-$.



Expression for $\hat{\Psi}(s, t)$ (Margolius and O'Reilly, 2016)

By conditioning on level y (highest peak / lowest trough)

$$\begin{split} \hat{\Psi}(\boldsymbol{s},t) &= \int_{\boldsymbol{y}=0}^{\infty} \int_{\nu=0}^{1} \int_{\boldsymbol{u}=0}^{1} \boldsymbol{e}^{\hat{A}(\boldsymbol{s},\boldsymbol{u})\boldsymbol{y}} \\ &\times \left(\hat{Q}_{+-}(\boldsymbol{u},\nu) - \int_{\eta=0}^{1} \int_{\theta=0}^{1} \hat{\Psi}(\boldsymbol{u},\eta) \hat{Q}_{-+}(\eta,\theta) \hat{\Psi}(\theta,\nu) d\theta d\eta \right) \\ &\times \boldsymbol{e}^{\hat{B}(\nu,t)\boldsymbol{y}} d\boldsymbol{u} d\nu d\boldsymbol{y}, \end{split}$$

where

$$\hat{A}(s,u) = \hat{Q}_{++}(s,u) + \int_{z=0}^{1} \hat{\Psi}(s,z)\hat{Q}_{-+}(z,t)du,$$

$$\hat{B}(\nu,t) = \hat{Q}_{--}(\nu,t) + \int_{z=0}^{1} \hat{Q}_{-+}(\nu,z)\hat{\Psi}(z,t)du.$$

$\hat{\Psi}(s, t)_{ij}$ – Conditioning on highest peak y



$\hat{\Psi}(s, t)_{ij}$ – Conditioning on lowest trough y

Here, $m, r, z, \ell, n = 0, 1, 2...$



Matrix $\hat{U}_1(s, t)$

For all s, t with $0 \le s \le t$, define $U_1(s, t)$ which solves

$$\frac{\partial}{\partial t} U_1(s,t) = U_1(s,t) \begin{bmatrix} T_{--}(t) & T_{-0}(t) \\ T_{0-}(t) & T_{00}(t) \end{bmatrix}$$

$$\frac{\partial}{\partial s} U_1(s,t) = -\begin{bmatrix} T_{--}(s) & T_{-0}(s) \\ T_{0-}(s) & T_{00}(s) \end{bmatrix} U_1(s,t)$$

$$U_1(s,s) = \mathbf{I}$$

and let

$$\hat{U}_1(s,t) = \sum_{n=0}^{\infty} U_1(s,t+n) = (I - U_1(s,s+1))^{-1} U_1(s,t).$$

Physical interpretation of $\hat{U}_1(s, t)$

For $i, j \in S_{-} \cup S_{0}$,

 $[U_1(s,t)]_{ij} = P[J(t) = j \mid J(s) = i, J(u) \in \mathcal{S}_- \cup \mathcal{S}_0, s \le u \le t]$





A related discrete-time Markov chain (DTMC)

Consider a DTMC on $[0, 1) \times S_-$ observed at the epochs when the CSFM hits level 0 from above, with $P = [P(s, t)_{ij}]_{s,t \in [0,1), i, j \in S}$ given by

$$P(s,t)dt = \begin{bmatrix} \mathbf{I} & \mathbf{0} \end{bmatrix} \int_0^1 \hat{U}_1(s,u) \begin{bmatrix} T_{-+}(u) \\ T_{0+}(u) \end{bmatrix} \hat{\Psi}(u,t)du$$

Here, *m*, *n* = 0, 1, 2....



Asymptotic periodic distribution $\xi(t)$ of such DTMC

Let $\xi(t)$, $t \in [0, 1)$, be the unique solution of the set of equations

$$\int_{s=0}^{1} \xi(s) P(s,t)_{ij} ds = \xi(t),$$
$$\int_{t=0}^{1} \xi(t) dt \mathbf{1} = \mathbf{1}.$$

Conditioning used in the derivation of $\pi(t; x)_{-}$

Here, $m, \ell = 0, 1, 2, ...$



Theorem

For all $t \in [0, 1)$ we have

$$\begin{bmatrix} \boldsymbol{p}_{-}(t) & \boldsymbol{p}_{0}(t) \end{bmatrix} = \alpha \int_{\boldsymbol{s}=0}^{1} \begin{bmatrix} \boldsymbol{\xi}(\boldsymbol{s}) & \boldsymbol{0} \end{bmatrix} \hat{U}_{1}(\boldsymbol{s},t) d\boldsymbol{s}, \quad (1)$$

and

$$\begin{bmatrix} \pi(t;x)_{+} & \pi(t;x)_{-} \end{bmatrix} = \int_{s=0}^{1} \begin{bmatrix} p_{-}(s) & p_{0}(s) \end{bmatrix} \begin{bmatrix} T_{-+}(s) \\ T_{0+}(s) \end{bmatrix}$$
$$\times \begin{bmatrix} e^{\hat{A}(s,t)x}(\mathbf{C}_{+})^{-1} & \int_{u=0}^{1} e^{\hat{A}(s,u)x}\hat{\Psi}(u,t)du(\mathbf{C}_{-})^{-1} \end{bmatrix} ds \quad (2)$$

and

$$\pi(t;x)_{0} = \int_{s=0}^{1} \left[\pi(s;x)_{+} \pi(s;x)_{-} \right] \left[\begin{array}{c} T_{+0}(s) \\ T_{-0}(s) \end{array} \right] \hat{U}_{0}(s,t) ds, \quad (3)$$

where $\alpha > 0$ is a normalizing constant that for all $t \in [0, 1)$ solves

$$p(t)\mathbf{1} + \int_{x=0}^{\infty} \pi(t; x) dx \mathbf{1} = 1.$$
 (4)

Normalising constant α

Let
$$(-A^{-1}(s,t)) = \int_{y=0}^{\infty} e^{A(s,t)y} dy$$
. Then

$$\begin{aligned} \alpha &= \left(\int_{s=0}^{1} \left[\begin{array}{cc} \xi(s) & \mathbf{0} \end{array} \right] \hat{U}_{1}(s,t) ds \mathbf{1} \\ &+ \int_{s=0}^{1} \int_{u=0}^{1} \left[\begin{array}{cc} \xi(u) & \mathbf{0} \end{array} \right] \hat{U}_{1}(u,s) \left[\begin{array}{cc} T_{-+}(s) \\ T_{0+}(s) \end{array} \right] du \\ &\times \left[\begin{array}{cc} A^{-1}(s,t)(\mathbf{C}_{+})^{-1} & \int_{v=0}^{1} A^{-1}(s,v) \hat{\Psi}(v,t) dv(\mathbf{C}_{-})^{-1} \end{array} \right] ds \mathbf{1} \\ &+ \int_{s=0}^{1} \int_{u=0}^{1} \int_{v=0}^{1} \left[\begin{array}{cc} \xi(v) & \mathbf{0} \end{array} \right] \hat{U}_{1}(v,u) \left[\begin{array}{cc} T_{-+}(u) \\ T_{0+}(u) \end{array} \right] dv \\ &\times \left[\begin{array}{cc} -A^{-1}(u,s)(\mathbf{C}_{+})^{-1} & -\int_{w=0}^{1} A^{-1}(u,w) \hat{\Psi}(w,s) dw(\mathbf{C}_{-})^{-1} \end{array} \right] du \\ &\times \left[\begin{array}{cc} T_{+0}(s) \\ T_{-0}(s) \end{array} \right] \hat{U}_{0}(s,t) ds \mathbf{1} \right)^{-1}. \end{aligned}$$

Current work



- Proofs, numerical schemes and examples.
- Other cyclic SFMs of interest:
 - Cycle duration constant or with some distribution
 - SFMs with a moving lower boundary, which drops to 0 at the end of a cycle
 - SFMs which restarts at level B > 0 at the start of a cycle, if the level falls below some threshold level 0 < b < B</p>

References

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Thank you





We found Ψ !



Peter found Ψ !

