

Emergence of shocks in large pool credit contagion models and resulting optimal bailout strategies

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with Christa Cuchiero and Stefan Rigger

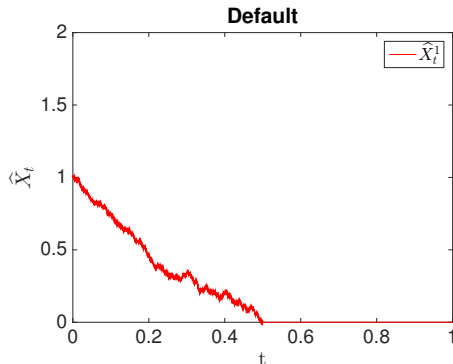
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- ▶ Background structural credit models with mutual obligations.
- ▶ Large pool limit and the supercooled Stefan problem.
- ▶ A central agent controlling the number of defaults:
 - ▶ well-posedness and ‘propagation of chaos’;
 - ▶ numerical solution of mean-field control problem;
 - ▶ illustration of strategies, losses, and cost to central agent.

'Structural' credit risk model

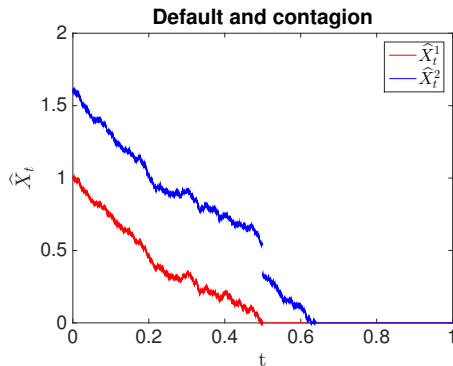
à la Merton, Cox,...

- ▶ The assets of a bank minus its liabilities are modelled by a process X_t .
- ▶ The bank is considered defaulted if X_t hits zero.



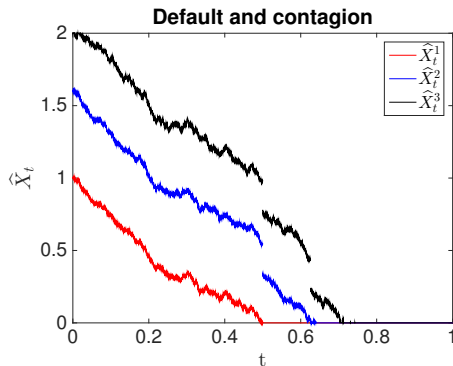
$\hat{X}_t^1$ the stopped process of bank 1.

- ▶ On default, bank 1 fails to meet some of its liabilities to bank 2.
- ▶ Bank 2's equity process jumps downwards.



$\hat{X}_t^1, \hat{X}_t^2$ the stopped processes of banks 1 and 2.

- ▶ On default of bank 2, it fails to meet some of its liabilities to bank 3.
- ▶ Bank 3's equity process jumps downwards.



$\hat{X}_t^1, \hat{X}_t^2, \hat{X}_t^3$ the stopped processes of banks 1, 2, and 3.

We assume the $X^{i,N}$ satisfy, for $i = 1, \dots, N$,

$$X_t^{i,N} = X_{0-}^{i,N} + B_t^{i,N} - \alpha \frac{1}{N} \sum_{i=1}^N \mathbb{1}_{\{\tau_{i,N} \leq t\}},$$

where

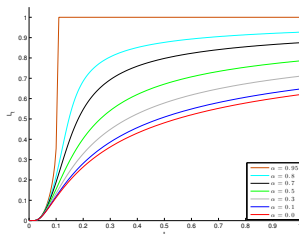
$$\tau_{i,N} = \inf\{t : X_t^{i,N} \leq 0\},$$

- ▶ $X_{0-}^{i,N}$ are non-negative i.i.d. random variables,
- ▶ $(B^{i,N})_{1 \leq i \leq N}$ is an N -dimensional standard Brownian motion, independent of $X_{0-} = (X_{0-}^{i,N})_{1 \leq i \leq N}$,
- ▶ and $\alpha \geq 0$ is a contagion parameter measuring the amount of inter-firm lending.

For $N \rightarrow \infty$, a representative bank follows

$$X_t = X_{0-} + B_t - \alpha \mathbb{P}\left(\inf_{0 \leq s \leq t} X_s \leq 0\right), \quad t \geq 0,$$

where B is a standard Brownian motion independent of X_{0-} .



- ▶ Cartoon model for credit risk with default contagion.
- ▶ Realistic α depends on mutual liabilities, recovery rates, asset vol – anything 0.5 to 5, or even higher (see Lipton, Kaushansky, & R, 2019)
- ▶ Similar models in neuroscience (see Delarue, Inglis, Rubenthaler, & Tanré, 2015)

$$L_t = \mathbb{P}\left(\inf_{0 \leq s \leq t} X_s \leq 0\right), \quad X_{0-} = 0.5$$

Let X_{0-} be a random variable with probability density f .

Consider the problem of finding a non-decreasing Λ such that the stochastic process

$$X_t = X_{0-} + B_t - \Lambda_t, \quad t \geq 0$$

satisfies the constraint

$$\Lambda_t = \alpha \mathbb{P}\left(\inf_{0 \leq s \leq t} X_s \leq 0\right), \quad t \geq 0,$$

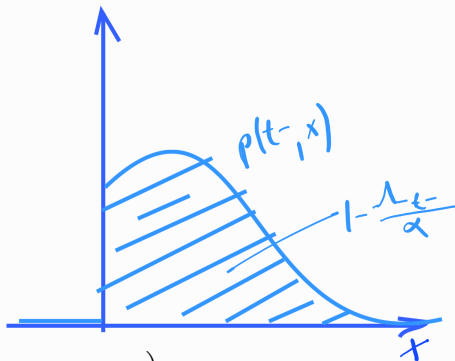
where B is a standard Brownian motion independent of X_{0-} .

Let $\tau := \inf\{t \geq 0 : X_t \leq 0\}$.

'Blow up' scenario

Hambly, Ledger, & Sojmark (2019); see also Delarue, Inglis, Rubenthaler, & Tanré (2015)

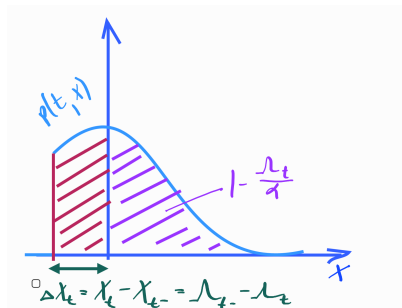
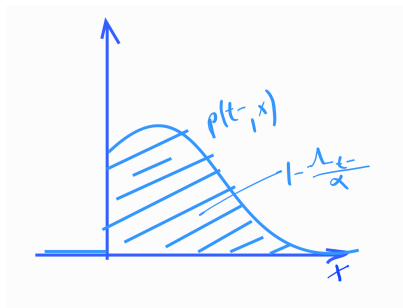
HLS19: If $\alpha > 2\mathbb{E}[X_0]$, $t \rightarrow \Lambda_t$ cannot be continuous for all t .*



- ▶ $\Lambda_t/\alpha = \mathbb{P}\left(\inf_{0 \leq s \leq t} X_s \leq 0\right)$
- ▶ $p(\cdot, t)$ the density of $\hat{X}_t = X_t \mathbb{1}_{\{\tau > t\}}$.

'Blow up' scenario

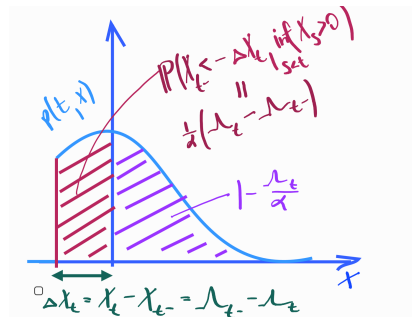
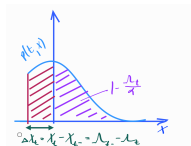
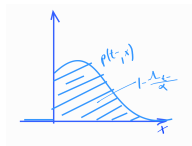
Hambly, Ledger, & Sojmark (2019); see also Delarue, Inglis, Rubenthaler, & Tanré (2015)



► Recall $X_t = X_{0-} + B_t - \Lambda_t$.

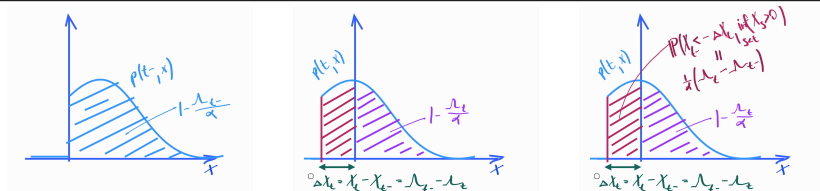
'Blow up' scenario

Hambly, Ledger, & Sojmark (2019); see also Delarue, Inglis, Rubenthaler, & Tanré (2015)



Physical and minimal solutions

Hambly, Ledger, & Sojmark (2019); see also Delarue, Inglis, Rubenthaler, & Tanré (2015)



Definition (Physical and minimal solutions)

Call a solution (X, Λ) *physical* if for all t

$$\Lambda_t - \lim_{s \rightarrow t^-} \Lambda_s = \left\{ \inf_x \left(\mathbb{P}(X_{t-} < x, \inf_{s < t} X_s > 0) < \frac{1}{\alpha} x \right) \right\}.$$

Call a solution $\underline{\Lambda}_t$ *minimal* if for any other solution Λ we have

$$\underline{\Lambda}_t \leq \Lambda_t, \quad t \geq 0.$$

- a. [Existence, Cuchiero, Rigger, & Svaluto-Ferro (2020); see also Ledger and Sojmark (2019)]: If $\mathbb{E}[X_{0-}] < \infty$, there is a unique minimal solution, which is also a physical solution.
- b. [Uniqueness]: Let X_{0-} possess a density f on $[0, \infty)$ that is bounded and changes monotonicity finitely often on compacts. Then physical solutions are unique.

- For any $t > 0$, $\Lambda \in C^1(t, t + \epsilon)$.
- The densities $p(s, \cdot)$, $s \in (t, t + \epsilon)$ are classical solutions of

$$\partial_t p = \frac{1}{2} \partial_{xx} p + \dot{\Lambda}_t \partial_x p, \quad x \geq 0, \quad t \in [0, T],$$

$$\dot{\Lambda}_t = \frac{\alpha}{2} \partial_x p(t, 0), \quad t \in [0, T] \quad \text{and} \quad \Lambda_0 = 0,$$

$$p(0, x) = f(x), \quad x \geq 0 \quad \text{and} \quad p(t, 0) = 0, \quad t \in [0, T].$$

Supercooled Stefan problem

Sherman (1970), Fasano, Primicerio, Howison, & Ockendon (80's), and Stefan (1889) for standard case

The transformation $u(t, x) := p(t, x - \Lambda_t)$, $x \geq \Lambda_t$ leads to

$$\partial_t u = \frac{1}{2} \partial_{xx} u, \quad x \geq \Lambda_t, \quad t \geq 0,$$

$$\dot{\Lambda}_t = \frac{\alpha}{2} \partial_x u(t, \Lambda_t), \quad t \geq 0 \quad \text{and} \quad \Lambda_0 = 0,$$

$$u(0, x) = f(x), \quad x \geq 0 \quad \text{and} \quad u(t, \Lambda_t) = 0, \quad t \geq 0.$$

This is the classical one-dimensional supercooled Stefan problem:

- ▶ $-f$ is the initial temperature in a liquid relative to its freezing point;
- ▶ Λ_t is the location of the liquid-solid boundary at time t ;
- ▶ $-u(t, \cdot)$ is the temperature in the liquid relative to its freezing point at time t ;
- ▶ and $\alpha > 0$ is a physical parameter.

We consider the *supercooled* regime, i.e., when $f \geq 0$.

▶ [Link](#)

A central agent injects cash at a rate $\beta_t^{i,N}$ into bank i ,

$$X_t^{i,N} = X_{0-}^{i,N} + \int_0^t \beta_s^{i,N} ds + B_t^{i,N} - \alpha \frac{1}{N} \sum_{i=1}^N \mathbb{1}_{\{\tau_{i,N} \leq t\}},$$

where $\tau_{i,N} := \inf\{t \geq 0 : X_t^{i,N} \leq 0\}$.

In analogy to before, we will consider the mean-field limit (TBC)

$$\begin{aligned} X_t &= X_{0-} + \int_0^t \beta_s ds + B_t - \Lambda_t, \\ \Lambda_t &= \alpha \mathbb{P}\left(\inf_{0 \leq s \leq t} X_s \leq 0\right). \end{aligned}$$

For given β , the central agent has a total cost

$$C_T(\beta) = \mathbb{E} \left[\int_0^T \beta_t dt \right],$$

and observes losses prior to T ,

$$L_{T-}(\beta) = \mathbb{P} \left(\inf_{0 \leq s < T} X_s(\beta) \leq 0 \right) = \Lambda_{T-}(\beta) / \alpha.$$

We will study the constrained optimisation problem

$$C_T(\beta) \longrightarrow \min_{\beta} \quad \text{subject to} \quad L_{T-}(\beta) \leq \delta.$$

We minimise the following **objective function**

$$\begin{aligned} J(\beta) &= \mathbb{E} \left[\int_0^T \beta_t dt \right] + \gamma \mathbb{P} \left(\inf_{0 \leq s < T} \underline{X}_s(\beta) \leq 0 \right) \\ &= \mathbb{E} \left[\int_0^T \beta_t dt + \gamma \mathbb{1}_{\{\widehat{\underline{X}}_{T-} = 0\}} \right], \end{aligned}$$

where γ traces out optimal pairs $(C_T^*(\gamma), L_{T-}^*(\gamma))$.

As function of δ , γ is a shadow price of preventing defaults,

$$\partial_\delta C_T^*(\delta) = -\gamma(\delta).$$

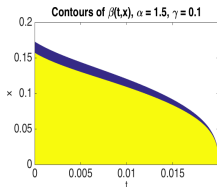
In all examples, we choose a gamma initial density,

$$f(x) = 1/\Gamma(k) \theta^{-k} x^{k-1} e^{-x/\theta}, \quad x \geq 0.$$

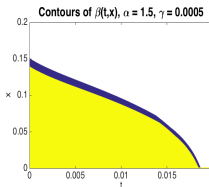
- ▶ Default parameters $k = 2$, $\theta = 1/3$.
- ▶ Smooth solutions for small t , but **blow-up is guaranteed for $\alpha > k\theta = 2/3$** .
- ▶ We will consider various values of **α around 1**.
- ▶ The terminal time is typically **$T = 0.02$** .
- ▶ We fix **$b_{\max} = 30$** at first, but investigate changes later on.

Optimal strategies

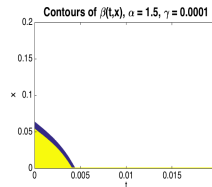
Feedback controls $\beta_t = \beta(t, X_t)$



$\gamma = 0.1$

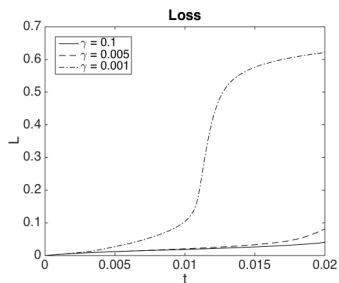


$\gamma = 0.0005$

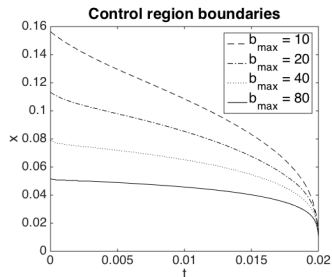


$\gamma = 0.0001$

Contour plots of $(t, x) \rightarrow \beta^*(t, x)$ for $\alpha = 1.5$ and different γ . The white region is $\{\beta^* \leq 0.05 b_{\max}\}$, the (yellow) shaded region $\{\beta^* \geq 0.95 b_{\max}\}$, the dark (blue) zone the transition.



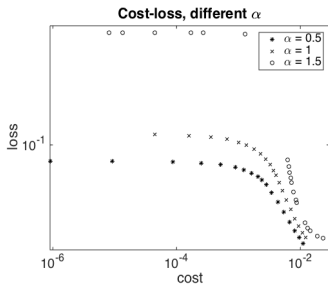
Loss for $\gamma \in \{0.1, 0.005, 0.001\}$



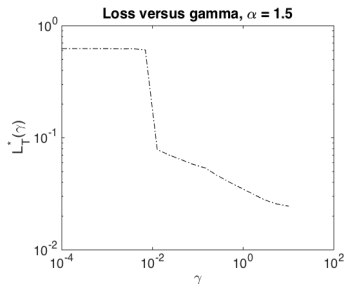
Control regions for different $b_{\max}$

Optimal cost-loss pairs

Small cash withdrawal can cause systemic events.

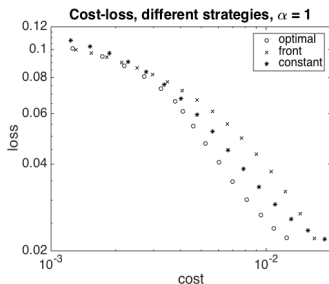


Pairs (C_T^*, L_T^*) for different α

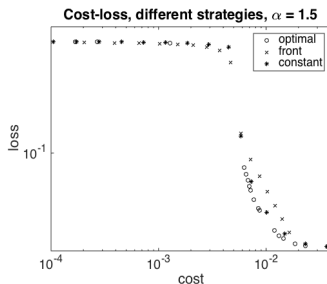


L_T^* as function of γ , $\alpha = 1.5$

Cost C_T^* and loss L_T^* in the optimal regime for logarithmically spaced $\gamma \in [0.0001, 0.1]$.



$\alpha = 1$, no jumps



$\alpha = 1.5$, jump possible

Cost-loss pairs (C_T^*, L_T^*) under optimal strategy compared to those for a

- ▶ **constant strategy**, $(C_T^u(c), L_T^u(c))$, where cash is injected for $0 < X_t \leq c$,
- ▶ **front-up strategy**, $(C_T^f(d), L_T^f(d))$, where X_0 is lifted to $d > 0$.

$$S_T := \{f \in L^2[0, T] \mid 0 \leq f \leq b_{\max} \text{ a.e.}\}$$

$$\mathcal{B}_T := \{\beta \text{ progressively measurable} \mid \mathbb{P}(\beta \in S_T) = 1\}.$$

Theorem (Existence of solutions for general drift)

For any $\beta \in \mathcal{B}_T$, there is a unique minimal solution to

$$X_t = X_{0-} + \int_0^t \beta_s ds + B_t - \alpha \mathbb{P}\left(\inf_{0 \leq s \leq t} X_s \leq 0\right).$$

Theorem (Existence of optimiser)

There is $\beta^* \in \mathcal{B}_T$ such that

$$V_\infty = \inf_{\beta \in \mathcal{B}_T} J(\beta) = J(\beta^*).$$

Propagation of chaos

Consider the N -bank setting

$$X_t^{i,N} := X_{0-}^{i,N} + \int_0^t \beta_s^{i,N} \, ds + B_t^{i,N} - \alpha L_t^N,$$

$$L_t^N := \frac{1}{N} \sum_{i=1}^N \mathbb{1}_{\{\tau^{i,N} \leq t\}}, \quad \text{where} \quad \tau_{i,N} := \inf\{t \geq 0 : X_t^{i,N} \leq 0\},$$

$$V_N := \inf_{\beta^N \in \mathcal{B}} \mathbb{E} \left[\int_0^T \frac{1}{N} \sum_{i=1}^N \beta_s^{i,N} \, ds + \gamma \tilde{L}_T^N(\beta^N) \right],$$

where $\tilde{L}^N(\beta^N)$ is the minimal solution with drift β^N and initial condition $X_{0-}^{i,N} = X_{0-}^N + N^{-\kappa}$ for $\kappa \in (0, 1/2)$ and all $i = 1, \dots, N$.

Theorem

Then it holds that $\lim_{N \rightarrow \infty} V_N = V_\infty$.

Recall $\underline{X}(\beta)$ corresponding to the minimal solution $\underline{\Lambda}(\beta)$, and

$$J(\beta) = \mathbb{E} \left[\int_0^T (\beta_t + \gamma \dot{L}_t) dt \right], \quad L_t = \mathbb{P} \left(\inf_{0 \leq s \leq T} \underline{X}_s(\beta) \leq 0 \right).$$

For regular solutions, $\hat{\underline{X}} = \underline{X}_t \mathbb{1}_{\{\tau > t\}}$ has a sub-probability density p supported on $(0, \infty)$ and an atomic mass at 0:

$$\partial_t p + \partial_x(\beta p) = \frac{1}{2} \partial_{xx} p + \dot{\Lambda}_t \partial_x p, \quad x \geq 0, \quad \Lambda_t = \alpha \left(1 - \int_0^\infty p(t, x) dx \right).$$

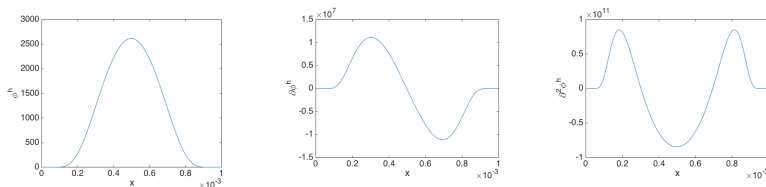
Assuming again regularity, $\dot{L}_t = \frac{1}{2} \partial_x p(t, 0)$ and

$$dX_t = (\beta_t - \alpha \dot{L}_t) dt + dB_t.$$

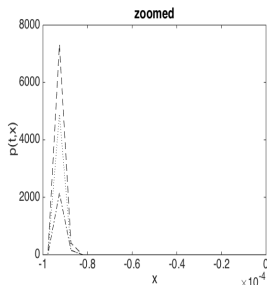
For (small) $h > 0$, approximate $p_x(t, 0)$ by way of the measure ν_t ,

$$\dot{L}_t^h = \frac{1}{2} \int_{-\infty}^{\infty} (-\partial \phi^h)(x) \nu_t(dx) = \frac{1}{2} \langle -\partial \phi^h, \nu_t \rangle.$$

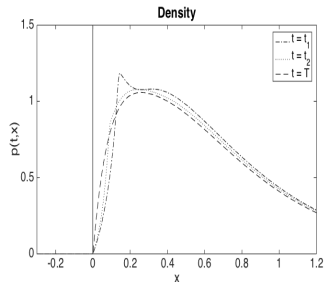
- Dynamics and objective can be rewritten in these terms.
- We smoothly transition SDE coefficients to 0 over $[-h, 0]$.



Smoothed Dirac delta and its derivatives, for $h = 10^{-3}$.



Density $p(t, \cdot)$ for small negative x .



Density $p(t, \cdot)$ in macroscopic range.

Parameters $\alpha = 1.5$, $\gamma = 0.1$; $\Gamma(2, 1/3)$ initial density

We consider the objective

$$\inf_{\beta \in \mathcal{H}^2(\mathbb{R})} \mathbb{E} \left[\int_0^T \left(\beta_t + \frac{\gamma}{2} \langle -\partial \phi^h, \nu \rangle + g(\beta_t) \right) dt \right],$$

where $g(b) = 0$ for $b \in [0, b_{\max}]$ and ∞ otherwise.

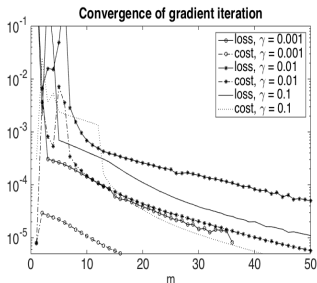
- ▶ We apply the method from R., Stockinger, Zhang, *A fast iterative PDE-based algorithm for feedback controls of nonsmooth mean-field control problems*, arXiv, 2021.
- ▶ We compute the measure by a forward PDE, and the gradient by a backward PDE for a ‘decoupling field’.

Finite difference convergence

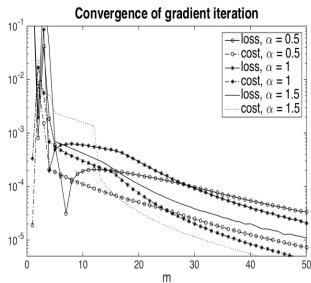
Finite termination

			$10^3 h$	1	2^{-1}	2^{-2}	2^{-3}	2^{-4}	2^{-5}	CPU
$\frac{N}{10^2}$	$\frac{N_x}{10^3}$	$10^3 \theta_N$	ρ_N							(s)
1	3.75	1.956	-2.42	0.5643	0.6430	0.7137	0.832	4.775	0	0.44
2	7.5	-0.806	1.31	0.5663	0.6260	0.6693	0.726	0.838	5.121	1.2
4	15	-0.614	1.82	0.5655	0.6164	0.6481	0.677	0.822	0.033	4.2
8	30	-0.337	1.94	0.5649	0.6118	0.6376	0.658	0.688	0.609	17
16	60	-0.173	—	0.5645	0.6096	0.6327	0.648	0.664	0.689	83
32	120	—	—	0.5643	0.6085	0.6304	0.644	0.654	0.668	427
			$10^2 \vartheta_h$	4.414	2.192	1.354	1.084	1.32	—	
			ϱ_h	2.01	1.61	1.24	0.81	—	—	

Mesh convergence, losses, $\alpha = 1.5$ and $\gamma = 0.1$.



$\alpha = 0.5$, varying γ , $N = 800$



varying α , $\gamma = 1$, $N = 800$

Convergence of C and L in the PGM for varying γ and α . Shown are $|L^{(m+1)} - L^{(m)}|$ and $|C^{(m+1)} - C^{(m)}|$.

$$dX_t^i = b(t, X_t^i, \nu_t^N) dt + \sigma(t, X_t^i) \sqrt{1 - \rho(t, \nu_t^N)^2} dW_t^i + \\ \sigma(t, X_t^i) \rho(t, \nu_t^N) dW_t^0 - \alpha(t) d\mathcal{L}_t^{N, \epsilon},$$

$$\nu_t^N = \frac{1}{N} \sum_{i=1}^N \delta_{X_t^i}, \quad \tau^i = \inf\{t \geq 0 : X_t^i \leq 0\},$$

$$L_t^N = 1 - \nu_t^N(\mathbb{R}^+), \quad \mathcal{L}_t^{N, \epsilon} = \int_0^t \epsilon^{-1} \kappa(\epsilon^{-1}(t-s)) L_s^N ds,$$

where all W^i , W^0 B.M., all independent.

- ▶ W^0 is a common noise;
- ▶ smoothed default contagion through Λ_t .

$$dX_t^i = b(t, X_t^i, \nu_t^N, \gamma_t^i) dt + \sigma(t, X_t^i, \nu_t^N) dW_t^i + \sigma_0(t, X_t^i, \nu_t^N) dW_t^0 \\ - \alpha(t, X_{t-}^i, \nu_{t-}^N) dL_t^N,$$

$$d\Lambda_t^i = \lambda(t, X_t^i, \nu_t^N) dt,$$

$$\nu_t^N = \frac{1}{N} \sum_{i=1}^N \mathbf{1}_{\theta_i > \Lambda_t^i} \delta_{X_t^i}, \quad L_t^N = 1 - \nu_t^N(\mathbb{R}),$$

where all W^i , W^0 B.M. and θ_i i.i.d. exponential, all independent.

- ▶ W^0 is a common noise;
- ▶ γ_t^i are controls;
- ▶ default contagion through Λ_t .

- ▶ C Cuchiero, C Reisinger, S Rigger. Optimal bailout strategies resulting from the drift controlled supercooled Stefan problem. Annals of Operations Research, 2023.
- ▶ B Hambly, A Petronilia, C Reisinger, S Rigger, A Søjmark. Contagious McKean–Vlasov problems with common noise: from smooth to singular feedback through hitting times, arxiv:2307.10800
- ▶ B Hambly, P Jettkant. Control of McKean–Vlasov SDEs with contagion through killing at a state-dependent Intensity, arxiv:2310.15854