

Continuous Auction Models

Gioia Carinci

University of Modena and Reggio Emilia

Joint work with

Pablo A. Ferrari (Univ. Buenos Aires)

Chiara Franceschini, Nicola Manelli (Univ. Modena)

PSPDE XIV, Toulouse, June 2026

1. Introduction
2. Continuous Model
3. Discrete Model
4. Convergence: Discrete \rightarrow Continuous
5. Poisson Random Initial Condition
6. Multi-Velocity Model
7. Conclusions

Setting: Repeated Auctions

Classical auction: few individuals bid for a unique good. **This paper:**

- ▶ one auctioneer, **many** seller agents
- ▶ **successive** auction rounds
- ▶ motivated by *renewable-energy markets*
- ▶ continuous-time, continuous-space scaling limit

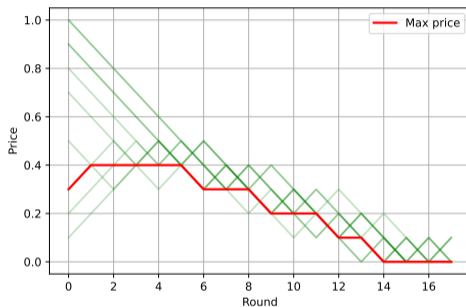


Figure: PSK model: $p=0.3$, $n=10$, $\gamma=0.1$. Green: bid trajectories. Red: max-price q_t^γ .

n sellers, bids configuration $B = [b_1, \dots, b_n : b_i \in \{0, \gamma, 2\gamma, \dots, 1\}]$, where $\gamma = 1/m$, $m \in \mathbb{N}$.

Auction rule — one round

- ▶ Fraction $p = k/n$ awarded to the **lowest np bids** $\Rightarrow np$ winners, $n(1-p)$ losers
- ▶ **Winners** increase bid by γ ; **Losers** decrease bid by γ
- ▶ *Myopic strategy*: each bidder reacts only to win/lose

One-step operator: let $b'_1(B) \leq \dots \leq b'_n(B)$ be the ordered bids, we define

$$T^\gamma B := [b'_i(B) + \gamma(\mathbb{1}\{0 < i \leq np\} - \mathbb{1}\{1 > i > np\}) : i = 1, \dots, n], \quad T_t^\gamma B := (T^\gamma)^{\lfloor t/\gamma \rfloor} B$$

Max price: $q_t^\gamma = b'_{np}(T_t^\gamma B)$ — largest winning bid at time t .

Periodicity around the Max Price

Lemma 1 (Bulk periodicity)

Let all bids \in three consecutive sites, then:

(a) **Bulk** (no boundary interaction):

$$(T^\gamma)^K B = B + \gamma \frac{w - \ell}{w + \ell} \cdot K$$

where the period K is given by

$$K = \frac{w + \ell}{\gcd(w, \ell)}$$

(b) **Boundary:** $[0^{(\ell)}, \gamma^{(w)}]$ stationary for $w < \ell$; $[(1-\gamma)^{(\ell)}, 1^{(w)}]$ stationary for $w > \ell$.

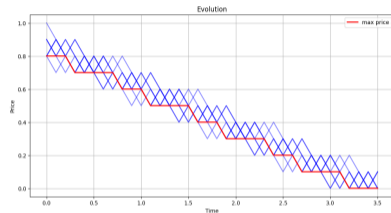


Figure: $w=3, \ell=5$: period $K=8$. Darker: superposed bids.

Continuous Model — Setup and Heuristics

Data:

- ▶ μ : probability measure on $[0, 1]$; $F = \text{CDF}$
- ▶ $p \in (0, 1)$: winner fraction
- ▶ $q_0 = F^{-1}(p)$: initial max price

Heuristics:

- ▶ bids $< q_t$ (*winners*): velocity $+1$
- ▶ bids $> q_t$ (*losers*): velocity -1
- ▶ w_t, l_t : winning/losing mass *at* q_t
- ▶ max-price velocity: $u_t = \frac{w_t - l_t}{w_t + l_t}$

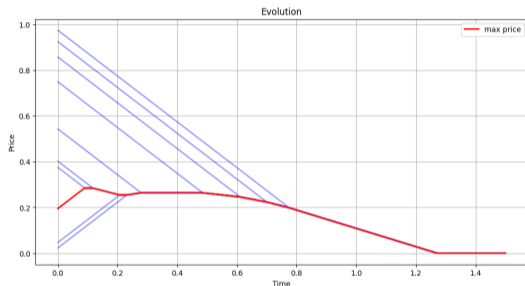


Figure: Continuous model, 10 delta measures, $p=0.3$.
Upon collision with q_t : bids coalesce.

The ODE System

For $p \in (0, 1)$ and initial measure μ with CDF F :

$$\frac{dq_t}{dt} = u_t \cdot \mathbb{1}\{q_t \in (0, 1)\}$$

$$w_t = p - F((q_t - t)^-), \quad \ell_t = F(q_t + t) - p$$

$$u_t = \frac{w_t - \ell_t}{w_t + \ell_t}, \quad q_0 = F^{-1}(p)$$

Bid distribution at time t

$$d(\mu\mathcal{T}_t)(x) = \delta_{q_t}(x) \underbrace{\mu([q_t - t, q_t + t])}_{\text{mass accumulated at } q_t} + d\mu(x-t) \mathbb{1}\{x < q_t\} + d\mu(x+t) \mathbb{1}\{x > q_t\}$$

Define:

$$G(x) := \int_0^x F(z) dz = \int_0^x (x - y) d\mu(y)$$

Existence and Uniqueness

Let

$$\mathcal{G}_t(x) := G(x+t) - G(x-t) = \int_{x-t}^{x+t} F(z) dz = \int_0^1 \Psi_{x,t}(z) d\mu(z)$$

where $\Psi_{x,t}(z) := (x+t-z)_+ - (x-t-z)_+$

Lemma 2

For all $t \geq 0$, the equation $\mathcal{G}_t(x) = 2pt$ with $x \in [q_0 - t, q_0 + t]$ has a *unique solution* $x_t = \mathcal{G}_t^{-1}(2pt)$.

Theorem 3 (Existence & Uniqueness)

The ODE system has a unique solution $(q_t)_{t \geq 0}$:

$$q_t = \begin{cases} \mathcal{G}_t^{-1}(2pt) & \mathcal{G}_t^{-1}(2pt) \in (0, 1) \\ 1 & \mathcal{G}_t^{-1}(2pt) > 1 \\ 0 & \mathcal{G}_t^{-1}(2pt) < 0 \end{cases}$$

Proof Idea: Existence and Uniqueness

By the lemma, for every $t \geq 0$ there is a unique $q_t \in [q_0 - t, q_0 + t]$ such that $\mathcal{G}_t(q_t) = 2pt$. Differentiate this identity:

$$\frac{d}{dt}\mathcal{G}_t(q_t) = 2p.$$

Since

$$\partial_t \mathcal{G}_t(q_t) = F(q_t + t) + F((q_t - t)^-), \quad \partial_x \mathcal{G}_t(q_t) = F(q_t + t) - F((q_t - t)^-),$$

and

$$F(q_t + t) = p + \ell_t, \quad F((q_t - t)^-) = p - w_t,$$

we get

$$2p = (2p + \ell_t - w_t) + (w_t + \ell_t)\dot{q}_t.$$

Therefore

$$\dot{q}_t = \frac{w_t - \ell_t}{w_t + \ell_t}.$$

Uniqueness

Then any solution of the ODE satisfies $\mathcal{G}_t(q_t) = 2pt$. By the lemma, this determines q_t uniquely.

The Baricenter Formula

Define the baricenter of μ on $(\alpha, \beta]$:

$$b[\alpha, \beta] := \frac{1}{F(\beta) - F(\alpha^-)} \int_{(\alpha, \beta]} z d\mu(z)$$

Proposition 4 (Baricenter decomposition)

The unique solution of the ODE system satisfies, for all $t \geq 0$:

$$q_t = b[q_t - t, q_t + t] + t \frac{w_t - \ell_t}{w_t + \ell_t}$$

Physical interpretation

- ▶ $b[q_t - t, q_t + t]$: baricenter at time 0 of bids that have hit q_t by time t
- ▶ Second term: net momentum displacement of that mass
- ▶ *Total momentum is conserved at each collision*

Proof of Equivalence: Theorem 2 \Leftrightarrow Proposition 3

Integration by parts on $\mathcal{G}_t(q_t) = \int_{q_t-t}^{q_t+t} F(z) dz$:

$$\begin{aligned} \int_{q_t-t}^{q_t+t} F(z) dz &= [z F(z)]_{q_t-t}^{q_t+t} - \int_{(q_t-t, q_t+t]} z d\mu(z) \\ &= (q_t+t) \underbrace{F(q_t+t)}_{=p+l_t} - (q_t-t) \underbrace{F((q_t-t)^-)}_{=p-w_t} - (w_t+l_t) b[q_t-t, q_t+t] \end{aligned}$$

Setting $= 2pt$ and simplifying:

$$q_t(w_t+l_t) + t(\ell_t-w_t) - (w_t+l_t) b[q_t-t, q_t+t] = 0$$

$$\Rightarrow \boxed{q_t = b[q_t-t, q_t+t] + t \frac{w_t-\ell_t}{w_t+l_t}}$$

Theorem 2 ($\mathcal{G}_t(q_t) = 2pt$) \Leftrightarrow Proposition 3 (baricenter formula), via integration by parts using $F(q_t+t) = p+l_t$ and $F((q_t-t)^-) = p-w_t$.

Explicit Solution: Uniform Measure

$F(x) = x$, $p \leq \frac{1}{2}$: **If** $p \leq \frac{1}{4}$:

$$q_t = \begin{cases} p & 0 \leq t \leq p \\ -t + 2\sqrt{pt} & p \leq t \leq 4p \\ 0 & t \geq 4p \end{cases}$$

If $\frac{1}{4} < p < \frac{1}{2}$:

$$q_t = \begin{cases} p & 0 \leq t \leq p \\ -t + 2\sqrt{pt} & p \leq t \leq \frac{1}{4p} \\ (2p-1)t + \frac{1}{2} & \frac{1}{4p} \leq t \leq \frac{1}{2(1-2p)} \\ 0 & t \geq \frac{1}{2(1-2p)} \end{cases}$$

If $p = \frac{1}{2}$: $q_t \equiv \frac{1}{2}$.

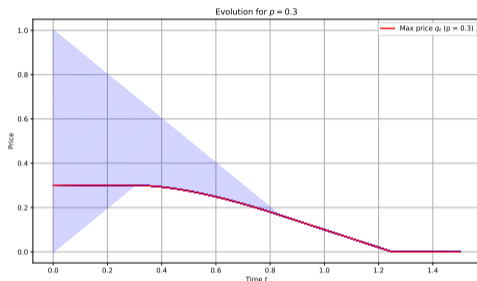


Figure: Continuous model, uniform initial measure, $p = 0.3$.

Theorem 5 (Evolution of delta measures)

For $\mu = \frac{1}{n} \sum_{b \in B} \delta_b$, the system has a *unique solution* $(B_t, q_t)_{t \geq 0}$:

- ▶ Each bid travels ballistically until hitting q_t
- ▶ $\dot{q}_t = u_t$: piecewise constant, right-continuous at collisions
- ▶ q_t : continuous and piecewise linear
- ▶ Finite total number of collisions

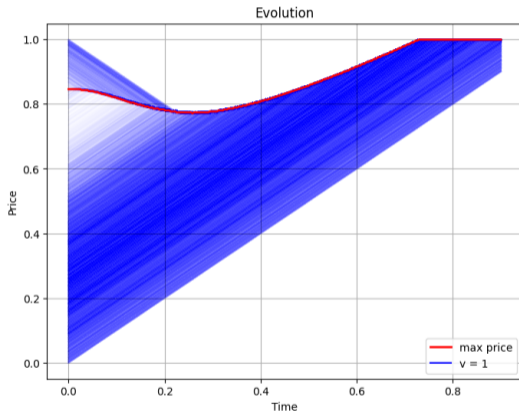


Figure: Sinusoidal initial density, $p=0.85$. Red: max-price q_t . Bids coalesce upon collision.

Learning parameter $\gamma = 1/m$, $m \in \mathbb{N}$; bids in $I^\gamma := [0, 1] \cap \gamma\mathbb{N}$; $w = np$ winners, $\ell = n(1-p)$ losers. **One-step operator:**

$$T^\gamma B := [b'_i(B) + \gamma(\mathbb{1}\{0 < i \leq np\} - \mathbb{1}\{1 > i > np\}) : i = 1, \dots, n]$$

$$T_t^\gamma B := (T^\gamma)^{\lfloor t/\gamma \rfloor} B, \quad q_t^\gamma := b'_{np}(T_t^\gamma B)$$

Key quantities

$$F(x) = \frac{1}{n} \sum_i \mathbb{1}\{b_i \leq x\}, \quad G_t(x) = \int_{x-t}^{x+t} F(z) dz, \quad b[\alpha, \beta] = \frac{\sum_i b_i \mathbb{1}_{b_i \in (\alpha, \beta]}}{\sum_i \mathbb{1}_{b_i \in (\alpha, \beta]}}$$

Proposition 6 (Max-price evolution, discrete)

At time $t \in \gamma\mathbb{N}$

$$\left| q_t^\gamma - b[q_t^\gamma - t, q_t^\gamma + t] - \frac{(p - F((q_t^\gamma - t)^-)) - (F(q_t^\gamma + t) - p)}{F(q_t^\gamma + t) - F((q_t^\gamma - t)^-)} \cdot t \right| \leq 2\gamma$$
$$|\mathcal{G}_t(q_t^\gamma) - 2pt| \leq 2\gamma$$

Idea of the proof: Once a bid's distance from q_t^γ falls below 2γ , it stays within 2γ of the max price forever. Hence the baricenter of bids near q_t^γ lies within 2γ of q_t^γ .

Why exact identity fails in the discrete model

- ▶ **Continuous (Thm. 2):** bids coalesce *exactly* at $q_t \Rightarrow \mathcal{G}_t(q_t) = 2pt$ holds *exactly*
- ▶ **Discrete:** bids cluster within 2γ of q_t^γ , but exact coalescence is impossible on a lattice of step γ

Convergence Theorem

Let $B^\gamma := [\gamma \lfloor b_i / \gamma \rfloor : b_i \in B]$ be the γ -discretization of B ($n =$ number of bids).

Theorem 7

Let $\mathcal{T}_t B$ (continuous) and $\mathcal{T}_t^\gamma B^\gamma$ (discrete) be the respective evolutions, with max prices q_t and q_t^γ . Then:

$$\sup_{t \geq 0} |q_t^\gamma - q_t| \leq (3n + 2)\gamma$$

$$\sup_{b \in B} \sup_{t \geq 0} |b_t^\gamma - b_t| \leq (3n + 2)\gamma$$

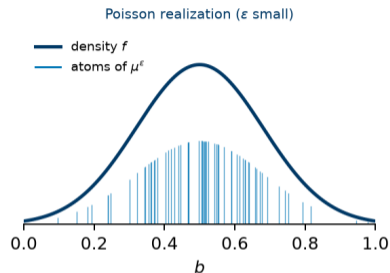
- ▶ Each of the n bids contributes $O(\gamma)$ error per collision event, there are at most n collisions, from which follows the bound $3n + 2$.
- ▶ The theorem provides a rigorous justification of the continuous model as scaling limit

Setup: Poisson Initial Bids

- ▶ $\mu(db) = f(b) db$: probability measure on $[0, 1]$, CDF F
- ▶ N^ε : Poisson process on $[0, 1]$ with intensity $\varepsilon^{-1}\mu$
- ▶ Empirical measure and CDF:

$$\mu^\varepsilon := \varepsilon \sum_{b \in N^\varepsilon} \delta_b, \quad F^\varepsilon(x) = \varepsilon \sum_{b \in N^\varepsilon} \mathbb{1}\{b \leq x\}$$

- ▶ $(q_t^\varepsilon)_{t \geq 0}$: max-price evolution for μ^ε
- ▶ $(q_t)_{t \geq 0}$: deterministic limit for μ



Atoms of μ^ε concentrate where f is large. As $\varepsilon \rightarrow 0$, $\mu^\varepsilon \rightarrow \mu$ a.s.

Integrated functions

$$G^\varepsilon(x) = \int_0^x F^\varepsilon(y) dy, \quad G(x) = \int_0^x F(y) dy, \quad \mathcal{G}_t^\varepsilon(x) = G^\varepsilon(x+t) - G^\varepsilon(x-t)$$

Theorem 8 (LLN for the max price)

For all fixed $t \geq 0$:

$$\lim_{\varepsilon \rightarrow 0} q_t^\varepsilon = q_t \quad \text{almost surely}$$

Proof sketch:

1. LLN for Poisson processes: $F^\varepsilon(x) \rightarrow F(x)$ and $G^\varepsilon(x) \rightarrow G(x)$ a.s., uniformly in x
2. Hence $\sup_x |\mathcal{G}_t^\varepsilon(x) - \mathcal{G}_t(x)| \rightarrow 0$ a.s.
3. Both $q_t = \mathcal{G}_t^{-1}(2pt)$ and $q_t^\varepsilon = (\mathcal{G}_t^\varepsilon)^{-1}(2pt)$
4. Continuity of the inversion map:

$$|q_t^\varepsilon - q_t| = |(\mathcal{G}_t^\varepsilon)^{-1}(2pt) - \mathcal{G}_t^{-1}(2pt)| \rightarrow 0 \quad \text{a.s.}$$

Gaussian Fluctuations — Functional CLTs

Rescaled fluctuation processes:

$$\zeta^{F^\varepsilon}(x) := \varepsilon^{-1/2}(F^\varepsilon(x) - F(x)), \quad \zeta^{G^\varepsilon}(x) := \varepsilon^{-1/2}(G^\varepsilon(x) - G(x))$$

Proposition 9 (Functional CLT for F^ε)

$\zeta^{F^\varepsilon} \xrightarrow{\text{law}} \zeta^F$ in $D[0, 1]$ (Skorokhod J_1): ζ^F is a centred Gaussian process with covariances

$$\mathbb{E}[\zeta^F(x)\zeta^F(y)] = F(\min(x, y))$$

i.e. $\zeta^F(x) = B(F(x))$, $B =$ standard Brownian motion on $[0, \infty)$.

Proposition 10 (Functional CLT for G^ε)

$\zeta^{G^\varepsilon} \xrightarrow{\text{law}} \zeta^G$ in $C[0, 1]$ (sup norm): ζ^G is a centred Gaussian process with covariances

$$\text{Cov}(\zeta^G(x), \zeta^G(y)) = \int_0^{x \wedge y} (x - z)(y - z) f(z) dz.$$

follows from the continuous mapping Theorem, since $\zeta^{G^\varepsilon}(x) = \int_0^x \zeta^{F^\varepsilon}(y) dy$.

Corollary 11

Let $\zeta^{\mathcal{G}^\varepsilon}(x, t) := \varepsilon^{-1/2}(\mathcal{G}_t^\varepsilon(x) - \mathcal{G}_t(x))$. As $\varepsilon \rightarrow 0$:

$$\zeta^{\mathcal{G}^\varepsilon}(x, t) \xrightarrow{\text{law}} \zeta^{\mathcal{G}}(x, t) := \zeta^{\mathcal{G}}(x+t) - \zeta^{\mathcal{G}}(x-t)$$

Centered Gaussian with covariance:

$$\text{Cov}(\zeta^{\mathcal{G}}(x, s), \zeta^{\mathcal{G}}(y, t)) = \int_0^1 \Psi_{x,s}(z) \Psi_{y,t}(z) \mu(dz)$$

where $\Psi_{x,t}(z) := (x+t-z)_+ - (x-t-z)_+$.

Proof tool

$\Phi_t(\psi)(x) := \psi(x+t) - \psi(x-t)$ is Lipschitz on $C[0, 1]$.

Apply the *Continuous Mapping Theorem* to $\zeta^{\mathcal{G}^\varepsilon} \xrightarrow{\text{law}} \zeta^{\mathcal{G}}$.

Theorem 12 (Gaussian fluctuations of q_t^ε)

Let $c_t := F(q_t + t) - F(q_t - t)$. Then:

$$\varepsilon^{-1/2}(q_t^\varepsilon - q_t) \xrightarrow{\text{law}} \zeta_t^q := -\frac{\zeta^G(q_t, t)}{c_t}$$

The process $\{\zeta_t^q\}_{t \geq 0}$ is *centered Gaussian* with

$$\text{Cov}(\zeta_t^q, \zeta_s^q) = \frac{1}{c_t c_s} \int_0^1 \Psi_{q_t, t}(z) \Psi_{q_s, s}(z) \mu(dz)$$

Proof: Functional Delta Method

Apply the Functional Delta Method to the map $\phi : g \mapsto g^{-1}(2pt)$.

Step 1. We have already established:

$$\varepsilon^{-1/2}(\mathcal{G}_t^\varepsilon(\cdot) - \mathcal{G}_t(\cdot)) \xrightarrow{\text{law}} \zeta^{\mathcal{G}}(\cdot, t) \quad \text{in } C[0, 1].$$

Step 2. The map ϕ is Hadamard differentiable at \mathcal{G}_t , with derivative in the direction $h \in C[0, 1]$:

$$\phi'_{\mathcal{G}_t}(h) = -\frac{h(\mathcal{G}_t^{-1}(2pt))}{\frac{d}{dx}\mathcal{G}_t(\mathcal{G}_t^{-1}(2pt))} = -\frac{h(q_t)}{c_t}, \quad c_t := \frac{d}{dx}\mathcal{G}_t(q_t) = F(q_t + t) - F(q_t - t).$$

Step 3. The Functional Delta Method gives:

$$\begin{aligned} \varepsilon^{-1/2}(q_t^\varepsilon - q_t) &= \\ &= \varepsilon^{-1/2}((\mathcal{G}_t^\varepsilon)^{-1}(2pt) - (\mathcal{G}_t)^{-1}(2pt)) \xrightarrow{\text{law}} \phi'_{\mathcal{G}_t}(\zeta^{\mathcal{G}}(\cdot, t)) = -\frac{\zeta^{\mathcal{G}}(q_t, t)}{c_t}. \end{aligned}$$

[van der Vaart & Wellner 1996]

Multi-Velocity Model — Setup

Generalization: bidder i has individual velocity $v_i \in \mathbb{N}$.

- ▶ Winner: $b_i \mapsto b_i + \gamma v_i$
- ▶ Loser: $b_i \mapsto b_i - \gamma v_i$
- ▶ $v_i = 1 \Rightarrow$ original PSK model

Main result

Under periodicity, the max-price velocity equals

$$u = \frac{w - \ell}{w + \ell} \cdot h$$

where $h := \frac{n}{\sum_{i=1}^n \frac{1}{v_i}}$

is the **harmonic mean** of $\{v_i\}$ at q_t .

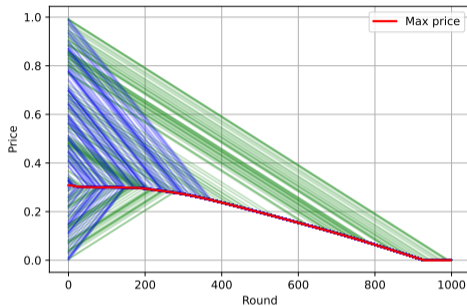


Figure: Two-velocity model.

Proposition 13

If $(T^\gamma)^\kappa B = B + N\gamma$ for some $N \in \mathbb{Z}$, $\kappa \in \mathbb{N}$ (periodic motion), then:

$$u = \frac{N}{\kappa} = \frac{w - \ell}{\sum_{i=1}^n \frac{1}{v_i}} = \frac{w - \ell}{w + \ell} \cdot h, \quad h := \frac{n}{\sum_{i=1}^n \frac{1}{v_i}}$$

$h =$ *harmonic mean* of the velocities $\{v_i\}$.

Key idea: bidder i must satisfy $(w_i(\kappa) - \ell_i(\kappa))v_i = N$ and $w_i(\kappa) + \ell_i(\kappa) = \kappa$, giving $w_i = \frac{1}{2}(\kappa + N/v_i)$ and $\ell_i = \frac{1}{2}(\kappa - N/v_i)$. Summing over i :

$$(w - \ell)\kappa = \sum_{i=1}^n (w_i(\kappa) - \ell_i(\kappa)) = N \sum_{i=1}^n \frac{1}{v_i} \implies \frac{N}{\kappa} = \frac{w - \ell}{\sum_i \frac{1}{v_i}}. \quad \square$$

Two Velocities: Periodicity vs. Separation

$v_i \in \{\nu_-, \nu_+\}$, $\nu_- < \nu_+$; n_-, n_+ bidders of each type; $h = n/(n_-/\nu_- + n_+/\nu_+)$.

Theorem 14 (Periodicity \Leftrightarrow bounded velocity)

The following are *equivalent*:

1. Motion attains a periodic increment state

2. $\frac{|w - \ell|}{n} \cdot h \leq \nu_-$

- ▶ **Condition holds:** slow and fast bids coexist near q_t
- ▶ **Condition fails:** slow bids *separate* from q_t ; drift ballistically (baricenter of fast bids grows faster than ν_-)

Theorem 15

Assume B is in the bulk regime: $|b_i - b_j| = O(\gamma)$ for all i, j . Then, uniformly in γ and $t \geq 0$:

$$\sup_{t \geq 0} \frac{1}{\gamma} |q_t^\gamma - q_t| < \infty, \quad \sup_t \frac{1}{\gamma} |b_i^\gamma(t) - b_i(t)| < \infty$$

where

$$q_t = \begin{cases} q_0 + \frac{w - \ell}{n} h t & \text{if } \frac{|w - \ell|}{n} h \leq \nu_- \\ q_0 + \frac{w - \ell - n_- \text{sign}(w - \ell)}{n} \nu_+ t & \text{otherwise} \end{cases}$$

$$b_i(t) := \begin{cases} q_t & \text{if } \frac{|w - \ell|}{n} h \leq \nu_- \text{ or } v_i = \nu_+ \\ q^0 + \text{sign}(w - \ell) \nu_- t & \text{otherwise} \end{cases}$$

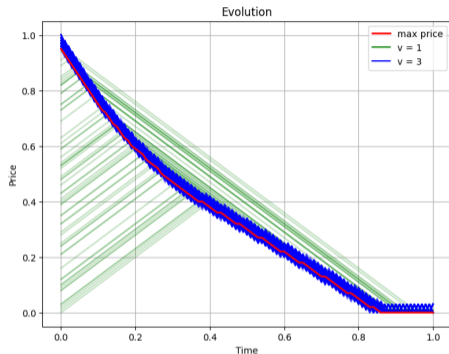


Figure: Two velocities with separation.

Continuous Multi-Velocity Model

ODE system for $B = [(b_1, v_1), \dots, (b_n, v_n)]$, $v_i \in \{\nu_-, \nu_+\}$:

$$h_t = \frac{w_t + \ell_t}{\frac{1}{n} \sum_i \frac{1}{v_i} \mathbb{1}\{b_i(t) = q_t\}}, \quad \frac{dq_t}{dt} = u_t$$
$$u_t = \begin{cases} \frac{w_t - \ell_t}{w_t + \ell_t} \cdot h_t, & \left| \frac{w_t - \ell_t}{w_t + \ell_t} \cdot h_t \right| \leq \nu_- \\ \frac{w_t - \ell_t - \text{sign}(w_t - \ell_t) m_t^-}{m_t^+} \cdot \nu_+, & \left| \frac{w_t - \ell_t}{w_t + \ell_t} \cdot h_t \right| > \nu_- \end{cases}$$
$$m_t^\pm = \frac{1}{n} \sum_{i=1}^n \mathbb{1}\{b_i(t) = q_t\} \mathbb{1}\{v_i = \nu_\pm\}$$
$$\frac{db_i}{dt} = \begin{cases} +v_i & b_i(t) < q_t \\ u_t & b_i(t) = q_t, |u_t| \leq v_i \quad (\text{stays on } q_t) \\ \text{sign}(u_t)v_i & b_i(t) = q_t, |u_t| > v_i \quad (\text{separates}) \\ -v_i & b_i(t) > q_t \end{cases}$$

Theorem 16

Let B^γ be the γ -discretization of B (velocities preserved). Then, uniformly in $\gamma \in (0, 1)$ and $t \geq 0$:

$$\frac{1}{\gamma} |q_t^\gamma - q_t| < C_1, \quad \frac{1}{\gamma} |b_i^\gamma(t) - b_i(t)| < C_2 \quad \forall i$$

for finite constants $C_1, C_2 > 0$.

Proof: induction over inter-collision intervals $[t_l, t_{l+1}]$:

1. Apply ballistic evolution theorem on each interval
2. Timing error: $|t_1^\gamma - t_1| \leq C'\gamma$ (linear behaviour near collision)
3. Propagate bound across all finitely many collisions

Open problems and possible generalisations

- ▶ **More than two velocities:** periodicity and convergence for $k \geq 3$ distinct bidder velocities
- ▶ **Max-price fluctuations (multi-velocity):** CLT for q_t^ε in the multi-velocity Poisson model; covariance formula remains **open**
- ▶ **Time-varying winner fraction $p(t)$:** generalisation of the ODE and baricenter formula to a time-dependent competition level
- ▶ **Stochasticity in the interaction:** random winner fraction p , random bid updates, or random participation at each round
- ▶ **Asymmetric update:** winners increase by γ_+ , losers decrease by $\gamma_- \neq \gamma_+$
- ▶ **Non-myopic strategies:** sellers with memory; cooperative or competitive interaction between sellers

Thank you!