

PAPER: Interdisciplinary statistical mechanics

How does latent liquidity get revealed in the limit order book?

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Abstract. Latent order book models have allowed for significant progress in our understanding of price formation in financial markets. In particular they are able to reproduce a number of stylized facts, such as the square-root impact law. An important question that is raised—if one is to bring such models closer to real market data—is that of the connection between the latent (unobservable) order book and the real (observable) order book. Here we suggest a simple, consistent mechanism for the revelation of latent liquidity that allows for quantitative estimation of the latent order book from real market data. We successfully confront our results to real order book data for over a hundred assets and discuss market stability. One of our key theoretical results is the existence of a market instability threshold, where the conversion of latent order becomes too slow, inducing *liquidity crises*. Finally we compute the price impact of a metaorder in different parameter regimes.

Keywords: agent-based models, market microstructure, market impact, models of financial markets, quantitative finance

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Introduction

In the past few years, price formation in financial markets has attracted the interest of a broad community of academics and practitioners. This can partially be explained by the growing quality of market data which now allows to test theories with levels of precision that have nothing to envy to natural sciences, see e.g. [1]. The study of market impact is both of fundamental and practical relevance. Indeed, while it is an important component for understanding price dynamics, it is also the source of substantial trading costs. Despite a few dissenting voices, the empirical nonlinear impact of metaorders (often coined the *square root law*) is now indisputably among the most robust stylized facts of modern finance (see e.g. [2–9]). In particular, for a given asset the expected average price return I between the beginning and the end of a metaorder of size Q follows $I(Q) = Y\sigma_d(Q/V_d)^\theta$ where Y is a numerical factor of order one, σ_d and V_d denote daily volatility and daily traded volume respectively, and $\theta < 1$ (typically $0.4 < \theta < 0.6$) bears witness of the concave nature of price impact.

Recently, a new class of agent based models (ABM) has helped to gain insight into the origins of the square root law. Given that the instantaneous liquidity revealed in the limit order book is very small (less than 1% of daily traded volume), *latent* order book models [2, 6, 10] build on the idea that revealed liquidity chiefly reflects the activity of high frequency market makers that act as intermediaries between much larger *unrevealed* volume imbalances. The latter are not revealed in the limit order book in order to avoid giving away precious private information, until the probability to get executed is large enough to warrant posting the order close to the bid (or to the ask). Within this class of ABM, reaction-diffusion models (see e.g. [11–13]) have proved very successful at reproducing the square root law in a setup free of price manipulation, and more recently consistent with the so-called diffusivity puzzle—that is, reconciling persistent order flow with diffusive price dynamics [13, 14]. However, one very important question is yet to be addressed if one is to connect such models with real observable and quantifiable data: what is the relation between revealed and latent liquidity and what are the mechanisms through which latent liquidity becomes revealed? These issues are the subject of the present communication. We propose a simple dynamic model to account for liquidity flow between the latent and revealed order books. We introduce two ingredients that are to our eyes essential to reproduce realistic limit order book shapes: (i) the incentive to reveal one’s liquidity increases with decreasing distance to the trade price, and (ii) the process of revealing latent liquidity is not instantaneous and lag effects may be an important source of instability, as real liquidity is found to vanish in certain regions of parameters. In addition to providing an alternative scenario for liquidity crises, we show that our framework allows one to infer the shape of latent (unobservable) liquidity from real (observable) order book data. This is important because the concept of a latent order book is sometimes criticized as a figment of the theorist’s imagination. Having more direct indications of its existence is comforting. For an interesting discussion on the effects of liquidity timescales, see also [15]—where the authors find that the delayed revelation of latent liquidity breaks the dynamical equilibrium between the market order flow and the limit order flow, triggering large price jumps.

In section 1 we present the model and derive the governing equations. In section 2 we compute the stationary states of the limit order book both analytically and numerically. In section 3 we calibrate our model to the order books of over a hundred assets and discuss market stability as function of incentive to reveal and conversion rate. In section 4 we compute the price impact of a metaorder and discuss its behavior. In section 5 we conclude.

1. A mechanism for latent liquidity revealing

Our starting point is the reaction-diffusion latent order book model of Donier *et al* [12] (see also [16]). In their setup, the latent volume densities of limit orders in the order book $\rho_B(x, t)$ (bid side) and $\rho_A(x, t)$ (ask side) at price x and time t evolve according to the following rules. Latent orders diffuse with diffusivity constant D , are canceled with

rate ν , and new intentions are deposited with rate λ . When a buy intention meets a sell intention they are instantaneously matched: $A + B = \emptyset$. The trade price p_t is conventionally defined through the equation $\rho_B(p_t, t) = \rho_A(p_t, t)$. Donier *et al* showed that the resulting stationary order book is locally linear (around the price). In particular, in the infinite memory limit $\nu, \lambda \rightarrow 0$ while keeping $\mathcal{L} = \lambda/\sqrt{\nu D}$ constant, one obtains $\rho_A^{\text{st}}(\xi) = \rho_B^{\text{st}}(-\xi) = \mathcal{L}\xi$ for $\xi = x - p_t > 0$ ⁵. The parameter \mathcal{L} was coined the *latent liquidity*. Here we choose to work in such an infinite memory limit. For an analysis of finite memory effects $\nu, \lambda \neq 0$ see [13].

With the intention of building a mechanism for latent liquidity revealing we define the revealed and latent order books (see figure 1), together with the revealed and latent limit order densities for the bid and ask sides of the book $\rho_{A/B}^{(r)}(x, t)$ and $\rho_{A/B}^{(\ell)}(x, t)$. We denote D_ℓ and D_r the diffusion coefficients in the latent and revealed order books respectively. Here the latent liquidity parameter is naturally defined with respect to the latent diffusivity D_ℓ , $\mathcal{L} := \lambda/\sqrt{\nu D_\ell}$ with ν, λ the cancellation and deposition rates in the latent order book (and $\nu, \lambda \rightarrow 0$). While diffusion in the latent book signifies heterogeneous reassessments of agents intentions [12], the idea of diffusion in the revealed order book certainly deserves a discussion. Once a limit order is placed in the revealed order book, if one wants to change its position before it gets executed, one has to cancel it and place it somewhere else. So one may argue that there should be no diffusion in the revealed book such that revealed order reassessments must go through the latent book before they can be posted again. However, we believe that (i) unrevealing one's order because one is no longer confident about one's reservation price and waiting for an arbitrary amount of time (of order $(\Gamma\omega_r)^{-1}$, see below) to reveal it back, and (ii) canceling an order knowing that it will immediately be posted back at a revised price, are in fact two distinct processes. We thus leave the possibility for a nonzero diffusivity D_r in the revealed order book. In addition, note that trading fees and priority queues discourage traders from changing posted orders. Indeed, when canceling a revealed order traders lose their position in the priority queue which may result in harmful delays in the execution. One is thus tempted to surmise that $D_r < D_\ell$ in the general case, but one might as well argue that the presence of HFT can considerably increase the value of D_r thereby inverting such an inequality. Be as it may, the limit $D_r \rightarrow 0$ will very likely be an interesting one to address.

Furthermore we posit that latent orders are revealed at a position dependent rate $\omega_r\Gamma_r(k\xi)$ and unrevealed at rate $\omega_\ell\Gamma_\ell(k\xi)$, where $\xi = x - p_t$ denotes the distance from the trade price, k^{-1} is a characteristic price scale, and Γ_r and Γ_ℓ are functions taking values in $[0, 1]$ ⁶. Naturally, buy/sell order matching $A + B \rightarrow \emptyset$ (with rate κ) only takes place in the revealed order book. Assuming $\Gamma_{r/\ell}$ to be continuous and sufficiently regular on \mathbb{R}^* , one may write for the ask side:

$$\partial_t \rho_A^{(r)} = D_r \partial_{xx} \rho_A^{(r)} + \omega_r \Gamma_r(k\xi) \rho_A^{(\ell)} - \omega_\ell \Gamma_\ell(k\xi) \rho_A^{(r)} - \kappa \rho_A^{(r)} \rho_B^{(r)} \quad (1a)$$

⁵ Following [12] we are setting in the reference frame of the informational price component $\hat{p}_t = \int_0^t ds V_s$, where V_t is an exogenous term responsible for the price moves related to common information.

⁶ While it may be reasonable that, when unrevealed, orders do not land on the same price x they took off from (by that moving away the earlier decision, say $(x)_r \rightarrow (x + \Delta x)_\ell$, with $\text{sign}(\Delta x) = \text{sign}(x - p_t)$), we shall not consider such a possibility in the present study for reasons of analytical tractability (restrict to $\Delta x = 0$).

How does latent liquidity get revealed in the limit order book?

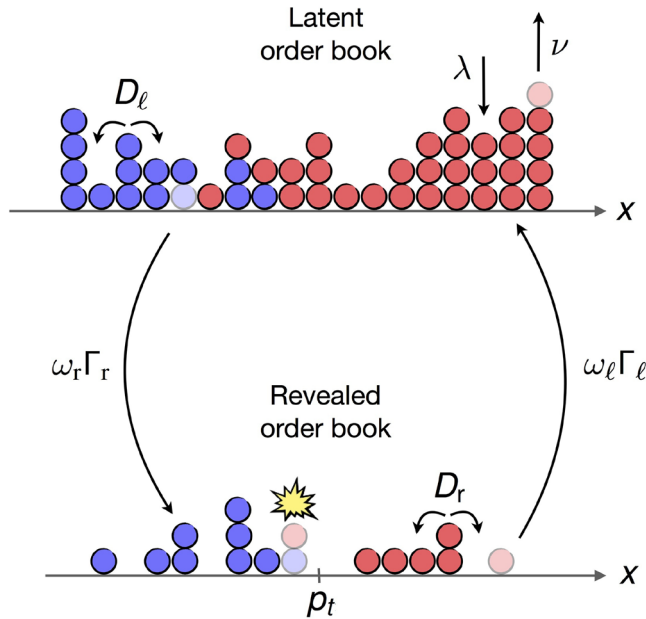


Figure 1. Schematics of the latent and revealed order books. Depositions, cancellations, diffusion and liquidity flow between the two books is signified with arrows (see section 1).

$$\partial_t \rho_A^{(\ell)} = D_\ell \partial_{xx} \rho_A^{(\ell)} - \omega_r \Gamma_r(k\xi) \rho_A^{(\ell)} + \omega_\ell \Gamma_\ell(k\xi) \rho_A^{(r)}, \quad (1b)$$

and for the bid side:

$$\partial_t \rho_B^{(r)} = D_r \partial_{xx} \rho_B^{(r)} + \omega_r \Gamma_r(-k\xi) \rho_B^{(\ell)} - \omega_\ell \Gamma_\ell(-k\xi) \rho_B^{(r)} - \kappa \rho_A^{(r)} \rho_B^{(r)} \quad (2a)$$

$$\partial_t \rho_B^{(\ell)} = D_\ell \partial_{xx} \rho_B^{(\ell)} - \omega_r \Gamma_r(-k\xi) \rho_B^{(\ell)} + \omega_\ell \Gamma_\ell(-k\xi) \rho_B^{(r)}. \quad (2b)$$

In the limit $\kappa \rightarrow \infty$, $\rho_A^{(r)}(x, t)$ and $\rho_B^{(r)}(x, t)$ do not overlap such that one may instead consider the difference function $\phi_r(x, t) := \rho_B^{(r)}(x, t) - \rho_A^{(r)}(x, t)$ and absorb the reaction terms without loss of information. Note however that the bid and ask sides of the latent order book are perfectly allowed to overlap. The trade price p_t is defined as the sign changing point of ϕ_r :

$$\lim_{\epsilon \rightarrow 0} [\phi_r(p_t + \epsilon, t) \phi_r(p_t - \epsilon, t)] < 0. \quad (3)$$

For the sake of simplicity we shall set $\omega_r = \omega_\ell = \omega$ ⁷. We also set $\Gamma_r = 1 - \Gamma_\ell := \Gamma$ where Γ is the conversion probability distribution function for which we impose $\Gamma(y \leq 0) = 1$. This implies that latent orders falling on the ‘wrong side’ reveal themselves with rate ω . One may rightfully argue that such orders should then be executed against the best quote, consistent with real market rules for which the real order book cannot be crossed. However this would prevent analytical progress. We ran numerical simulations

⁷ Note that relaxing this hypothesis is possible (see e.g. section 2.1 for the case $D_r = 0$) but at the cost of additional complexity. We checked that this point was without effect on our main qualitative results.

(see section 2.3) in which such revealed orders are properly located at the best quote and did not observe any significant impact on our main results.

Subtracting equation (1a) to (2a) and injecting $\rho_B^{(r)} = \phi_r \mathbb{1}_{\{x < p_t\}}$, $\rho_A^{(r)} = -\phi_r \mathbb{1}_{\{x > p_t\}}$, one obtains the following set of equations, central to our study:

$$\partial_t \rho_B^{(\ell)} = D_\ell \partial_{xx} \rho_B^{(\ell)} - \omega \left\{ \Gamma(-k\xi) \rho_B^{(\ell)} - [1 - \Gamma(-k\xi)] \mathbb{1}_{\{x < p_t\}} \phi_r \right\} \quad (4a)$$

$$\partial_t \rho_A^{(\ell)} = D_\ell \partial_{xx} \rho_A^{(\ell)} - \omega \left\{ \Gamma(k\xi) \rho_A^{(\ell)} + [1 - \Gamma(k\xi)] \mathbb{1}_{\{x > p_t\}} \phi_r \right\} \quad (4b)$$

$$\partial_t \phi_r = D_r \partial_{xx} \phi_r + \omega \left\{ \Gamma(-k\xi) \rho_B^{(\ell)} - \Gamma(k\xi) \rho_A^{(\ell)} - [1 - \Gamma(k|\xi|)] \phi_r \right\}. \quad (4c)$$

Equation (4) must be complemented with a set of boundary conditions. In particular we impose that $\lim_{x \rightarrow \infty} \partial_x \rho_A^{(\ell)} = -\lim_{x \rightarrow -\infty} \partial_x \rho_B^{(\ell)} = \mathcal{L}$ (see [12]), and that $\rho_A^{(\ell)}$ when $x \rightarrow -\infty$, respectively $\rho_B^{(\ell)}$ when $x \rightarrow \infty$, do not diverge. In addition, whenever $D_r \neq 0$ ⁸, one must impose that $\phi_r(0) = 0$ and $\phi_r(x)$ does not diverge when $|x| \rightarrow \infty$.

2. Stationary order books

In this section we compute analytically and numerically the stationary order books as function of the different parameters, and discuss interesting limit cases. Setting $\partial_t \rho_B^{(\ell)} = \partial_t \rho_A^{(\ell)} = \partial_t \phi_r = 0$ in equation (4) one obtains for all $\xi \in \mathbb{R}^*$, $\rho_B^{(\ell)}(\xi) = \rho_A^{(\ell)}(-\xi)$ and $\phi_r(\xi) = -\phi_r(-\xi)$. This allows one to solve the problem on \mathbb{R}^{+*} , with boundary conditions $\rho_B^{(\ell)}(0^+) = \rho_A^{(\ell)}(0^+)$, $\partial_\xi \rho_B^{(\ell)}(0^+) = -\partial_\xi \rho_A^{(\ell)}(0^+)$. The system one must solve for $\xi > 0$ reduces to:

$$0 = D_\ell \partial_{\xi\xi} \rho_B^{(\ell)} - \omega \rho_B^{(\ell)} \quad (5a)$$

$$0 = D_\ell \partial_{\xi\xi} \rho_A^{(\ell)} - \omega \left\{ \Gamma(k\xi) \rho_A^{(\ell)} + [1 - \Gamma(k\xi)] \phi_r \right\} \quad (5b)$$

$$0 = D_r \partial_{\xi\xi} \phi_r - \omega \left\{ \Gamma(k\xi) \rho_A^{(\ell)} + [1 - \Gamma(k\xi)] \phi_r - \rho_B^{(\ell)} \right\}. \quad (5c)$$

2.1. Analytical and numerical solutions

Here we provide a solution of equation (5) for three distinct cases of interest $D_r = 0$, $D_r = D_\ell$, and $D_r \neq D_\ell$. Note that for $D_\ell = 0$, one can show that $\rho_B^{(\ell)}(\xi > 0) = 0$, while $\phi_r(\xi > 0) = a\xi + b$ with a and b two constants, and $\rho_A^{(\ell)} = \phi_r[\Gamma(k\xi) - 1]/\Gamma(k\xi) \approx -\phi_r/\Gamma(k\xi)$ for $\xi \rightarrow \infty$, by assuming Γ to vanish at infinity. Such solutions are not compatible with the boundary conditions at infinity, and will therefore not be further inspected.

⁸ While $D_r = 0$ is an interesting limit, $D_\ell = 0$ does not seem to be particularly appealing on modelling grounds. In section 2.1 we give a more solid theoretical argument to support this claim.

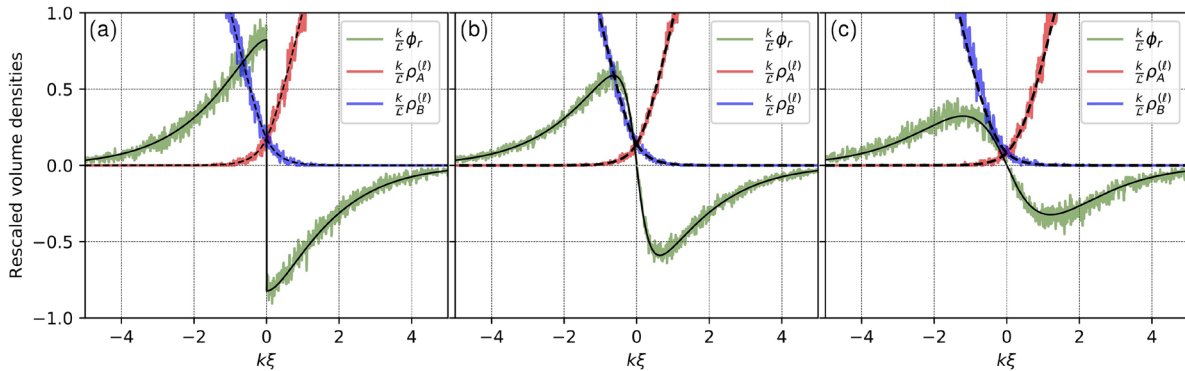


Figure 2. Rescaled stationary order books as function of rescaled price for (a) $D_r = 0$, (b) $D_r \neq D_\ell$ ($D_r < D_\ell$), and (c) $D_r = D_\ell$. Solid black lines indicate the rescaled theoretical revealed order density $k\phi_r/\mathcal{L}$ while dashed black lines signify the theoretical latent order densities $k\rho_{A/B}^{(\ell)}/\mathcal{L}$. The results of the numerical simulation presented in section 2.3 are plotted with color lines on top of the analytical curves, with $k\ell_\ell = 0.35$ for (a) and (c), and $\ell_r/\ell_\ell = 0.32$ for (b). Typical computational time: 1 min.

2.1.1. *The limit case $D_r = 0$.* Setting $D_r = 0$ in equation (5) and introducing $\ell_\ell := \sqrt{D_\ell/\omega}$, one finds that the stationary order book densities are given by (see figure 2(a)):

$$\rho_B^{(\ell)}(\xi) = \frac{\mathcal{L}\ell_\ell}{2} e^{-\xi/\ell_\ell} \tag{6a}$$

$$\rho_A^{(\ell)}(\xi) = \mathcal{L}\xi + \frac{\mathcal{L}\ell_\ell}{2} e^{-\xi/\ell_\ell} \tag{6b}$$

$$\phi_r(\xi) = \frac{\mathcal{L}\ell_\ell}{2} e^{-\xi/\ell_\ell} - \frac{\mathcal{L}\xi\Gamma(k\xi)}{1 - \Gamma(k\xi)}. \tag{6c}$$

Let us stress that the solution $\phi_r(\xi)$ is discontinuous in $\xi = 0$, consistent with no diffusion in the revealed order book. More precisely, provided $\Gamma'(0^+) \neq 0$, one has $\phi_r(0^-) - \phi_r(0^+) = -\mathcal{L}[\ell_\ell + 2/(\Gamma'(0^+)k)]$. While the solution for the case $D_r = 0$ can be expressed for an arbitrary function $\Gamma(y > 0)$, this is not the case for $D_r \neq 0$ and one must specify its shape. In the following we choose to work with:

$$\Gamma(y) = \begin{cases} 1 & \forall y \leq 0 \\ e^{-y} & \forall y > 0. \end{cases} \tag{7}$$

Note that studying the effect of a scale-invariant power law decaying Γ could also yield interesting results⁹.

⁹ Also note that in this particular case, the solution for $\omega_\ell \neq \omega_r$ can be easily expressed by substituting ℓ_ℓ by $\ell_\ell^{(r)} = \sqrt{D_\ell/\omega_r}$ into equations (6a) and (6b), and replacing equation (6c) by:

$$\phi_r(\xi) = \frac{\omega_r}{\omega_\ell} \left[\frac{\mathcal{L}\ell_\ell^{(r)}}{2} e^{-\xi/\ell_\ell^{(r)}} - \frac{\mathcal{L}\xi e^{-k\xi}}{1 - e^{-k\xi}} \right].$$

2.1.2. *The limit case $D_r = D_\ell$.* For $D_r = D_\ell$ and provided $k\ell_\ell \neq 1$ the stationary books are given by (see figure 2(c)):

$$\rho_B^{(\ell)}(\xi) = \frac{\mathcal{L}}{k} g(k\ell_\ell) e^{-\xi/\ell_\ell} \tag{8a}$$

$$\rho_A^{(\ell)}(\xi) = \mathcal{L}\xi + \frac{\mathcal{L}}{k} g(k\ell_\ell) e^{-\xi/\ell_\ell} + \phi_r(\xi) \tag{8b}$$

$$\begin{aligned} \phi_r(\xi) = \mathcal{L} & \left(\frac{1}{(k\ell_\ell)^2 - 1} \left[\xi + \frac{2k\ell_\ell^2}{(k\ell_\ell)^2 - 1} \right] e^{-k\xi} + \frac{g(k\ell_\ell)}{k^2\ell_\ell(k\ell_\ell + 2)} e^{-(1/\ell_\ell + k)\xi} \right. \\ & \left. + \left[\frac{g(k\ell_\ell)}{2k\ell_\ell} \xi - \frac{2k\ell_\ell^2}{[(k\ell_\ell)^2 - 1]^2} - \frac{g(k\ell_\ell)}{k^2\ell_\ell(k\ell_\ell + 2)} \right] e^{-\xi/\ell_\ell} \right), \end{aligned} \tag{8c}$$

where we introduced $g(\zeta) = 2\zeta^2(2 + \zeta)^2 / [(1 + \zeta)^2(8 + 3\zeta)]$. For $k\ell_\ell = 1$ the result can be obtained by Taylor expanding equation (8) about $k\ell_\ell = 1$. Note that in this case the function $\phi_r(\xi)$ is continuous in $\xi = 0$, consistent with nonzero diffusivity in the revealed order book.

2.1.3. *The general case $D_r \neq D_\ell$.* For $D_r \neq D_\ell$ the set of equation (5) must be solved numerically. Figure 2(b) displays a plot of the stationary order books computed using a finite difference method for $D_r/D_\ell \approx 0.1$, that is $\ell_\ell/\ell_r \approx 3$ where we introduced $\ell_r := \sqrt{D_r/\omega}$. As one can see, in the situation $D_r < D_\ell$ the revealed order book's shape is somewhat in between the cases $D_r = 0$ and $D_r = D_\ell$, that is continuous but with a steeper slope at $\xi = 0$. As expected, a little amount of diffusion in the revealed order book suffices to regularize the singularity at the trade price.

As it can be seen from equation (5a) (or from equations (6a) and (8a) in the particular cases $D_r = 0$ and $D_r = D_\ell$), in all cases ℓ_ℓ denotes the typical scale over which the latent books overlap. This is consistent with the idea that ℓ_ℓ is the typical displacement by diffusion of a latent order in the vicinity of the trade price during a time interval ω^{-1} , that is before it gets revealed. Also, it can be seen from equations (6c) and (8c) $D_r = 0$ and $D_r = D_\ell$ that the typical horizontal extension of the revealed order book, commonly called *order book depth*, is given by $\max(k^{-1}, \ell_\ell)$, consistent with the decay of the conversion probability function Γ and the horizontal extension of the latent books. Note however that, as shall be argued in section 3, k^{-1} must always be of order or larger than ℓ_ℓ for stability reasons, and therefore $\max(k^{-1}, \ell_\ell) \sim k^{-1}$. In the following we shall thus call k^{-1} the order book depth. Finally note that when ℓ_r (equivalently D_r) is decreased while keeping all other parameters constant, the slope of the revealed order book around the origin increases, by that concentrating further the available liquidity around the trade price.

2.2. The LLOB limit

Our model being built upon the locally linear order book model (LLOB) by Donier *et al* [12], we should be able to recover such a limit for certain values of the parameters. Since

there is only one diffusion coefficient in the LLOB model, the latter should correspond to the $D_r = D_\ell$ case. Then, the LLOB model assumes no lag effect, i.e. latent orders are immediately executed when at the trade price. This translates into $\omega \rightarrow \infty$ or equivalently $\ell_\ell \rightarrow 0$, that is no overlap of the latent books. More rigorously, nondimensionalizing equation (8) as $\{\tilde{\phi}_r, \tilde{\rho}^{(\ell)}\} = \{k\phi_r/\mathcal{L}, k\rho^{(\ell)}/\mathcal{L}\}$ and $\tilde{\xi} = k\xi$, one can see that taking the limit $k\ell_\ell \rightarrow 0$ yields for all $\tilde{\xi} > 0$, $\tilde{\rho}_B^{(\ell)}(\tilde{\xi}) \rightarrow 0$, and $\tilde{\rho}_A^{(\ell)}(\tilde{\xi}) + \tilde{\rho}_A^{(r)}(\tilde{\xi}) = \tilde{\xi} := \tilde{\rho}_A^{\text{LLOB}}(\tilde{\xi})$ that is precisely the LLOB result. To summarize, provided the latent order book of Donier *et al* is defined as the sum of the latent and revealed books, the LLOB limit is recovered for $k\ell_\ell \ll 1$. The latter condition indicates that the typical displacement ℓ_ℓ of latent orders in the vicinity of the price must remain small compared to the order book depth k^{-1} . Note that, despite what a first intuition might suggest, the condition $k \rightarrow \infty$, that is no incentive to give away information until absolutely necessary, is not required to recover the LLOB limit.

2.3. Numerical simulation

In order to test our results, we performed a numerical simulation of our model (see figure 2) which proceeds as follows. We define four vectors $\rho_A^{(\ell)}, \rho_A^{(r)}, \rho_B^{(\ell)}, \rho_B^{(r)}$ of size 2000 on the price axis (a good trade-off between the continuous approximation of the model and computational time cost). Each component of these vectors stores the number of orders contained at that price at a given time. At each cycle we draw the orders that shall diffuse from a binomial distribution of parameter $p_{\ell/r}$, directly related to the diffusion constant in the respective book by the relation $D_{\ell/r} = p_{\ell/r}/(2\tau)$ where τ denotes the time step. Some of the orders (drawn from a binomial of parameter 1/2) will move to the left and the remaining to the right. Reflecting boundary conditions are imposed for revealed orders; the slope of the latent book at the boundaries is ensured by an incoming current of particles $J = D_\ell \mathcal{L}$. Then, some orders in the latent book are drawn from a binomial distribution of parameter $\omega\tau\Gamma(k\xi)$ for the ask side (resp. $\omega\tau\Gamma(-k\xi)$ for the bid side) and moved to the revealed book. Here p_t denotes the mid-price. Equivalently revealed orders are moved to the latent book, only with parameter $\omega\tau(1 - \Gamma(\cdot))$. Whenever bid and ask orders are found at the same price, they are cleared from the book. Figure 2 shows that the results of the numerical simulations are in very good agreement with the analytical solutions.

3. Market stability and calibration to real data

In this section, we address the important question of market stability, as given by the amount of liquidity in the revealed order book. Clearly, when the conversion rate ω is low, the revealed liquidity is thin and prices can be prone to liquidity crises, even when the latent liquidity is large. We calibrate our model to real order book data and discuss the results in the light of the stability map provided by our model.

3.1. Market stability

Imposing that the order densities $\rho_A^{(\ell)}, \rho_A^{(r)}, \rho_B^{(\ell)}, \rho_B^{(r)}$ must be non negative, consistent with a physically meaningful solution, restricts the possible values of $\ell_\ell := \sqrt{D_\ell/\omega}$ for a given order book depth k^{-1} .

In the $D_r = 0$ case, combining equation (6c) with equation (7) yields $\phi_r(0^+) = \mathcal{L}[\ell_\ell/2 - 1/k]$. Restricting to $\phi_r(0^+) \leq 0$ (which is tantamount to $\rho_A^{(r)}(0^+), \rho_B^{(r)}(0^+) \geq 0$) gives $k\ell_\ell \leq 2$. Note however that this condition is not sufficient to say that the order densities are everywhere positive, but that they are only positive around the origin. The necessary and sufficient condition to ensure full positiveness reads $k\ell_\ell \leq 1$. For $1 \leq k\ell_\ell \leq 2$ the order book displays a ‘hole’ along the price axis, but is well defined around the origin. Since we are most interested in the revealed liquidity in the vicinity of the trade price we will choose $k\ell_\ell \leq 2$ (or, in terms of ω , $\omega \geq D_\ell k^2/4$) as our *stability condition*, with no qualitative and only little quantitative effect on our main conclusions. The maximum amplitude of the real order book density scales as:

$$\max_{\xi} |\phi_r(\xi)| = -\phi_r(0^+) = \frac{\mathcal{L}}{k} \left(1 - \frac{k\ell_\ell}{\zeta_c} \right), \quad \zeta_c = 2. \tag{9}$$

For $D_\ell = D_r$ the stability condition is imposed by the sign of the slope at $\xi = 0$. One finds that the critical value of $k\ell_\ell$ for which the liquidity around the origin vanishes is given by $\zeta_c = [-2 + (73 - 6\sqrt{87})^{1/3} + (73 + 6\sqrt{87})^{1/3}]/3 \approx 1.875^{10}$. Arguing that in this case the maximum of the density can be approximated by $\max_{\xi} |\phi_r'(0^+) \xi e^{-k\xi}| = |\phi_r'(0^+)|/(ek)$ yields:

$$\max_{\xi} |\phi_r(\xi)| \sim \frac{\mathcal{L}}{ek} \frac{\zeta_c(3\zeta_c^2 + 4\zeta_c - 3)}{(1 + \zeta_c)^2(8 + 3\zeta_c)} \left(1 - \frac{k\ell_\ell}{\zeta_c} \right), \quad \zeta_c \approx 1.875. \tag{10}$$

Figure 3(a) displays $-k\phi_r'(0^+)/\mathcal{L}$ as function of the dimensionless parameters $k\ell_\ell$ and $k\ell_r$. The dash-dotted line corresponding to $\phi_r'(0) = 0$ splits the parameter space into a stable region (green) and an unstable region (red). We naturally checked that the analytical values of ζ_c obtained above for $\ell_\ell = \ell_r$ and $\ell_r \rightarrow 0$ are recovered. As one can see, while the role played by ℓ_r with respect to the position of the critical line ζ_c is quite marginal (quasi-vertical dash dotted line), the slope of the order book around $\xi = 0$ increases with decreasing ℓ_r/ℓ_ℓ . More precisely, figure 3(c) displays the slope of the revealed book in the vicinity of the transition and shows that, at given ℓ_r/ℓ_ℓ , the slope indeed scales linearly with $|k\ell_\ell - \zeta_c|$. In addition the top right inset shows that the slope also scales as ℓ_ℓ/ℓ_r , which finally leads to $|\phi_r'(0^+)| \sim |k\ell_\ell - \zeta_c|(\ell_\ell/\ell_r)$. The other insets show ζ_c as function of ℓ_r/ℓ_ℓ , and the total revealed volume. For the sake of completeness, figure 3(b) displays a proxy of the overlap between the latent bid and ask books in the parameter space $(k\ell_\ell, k\ell_r)$. While vanishing in the region $k\ell_\ell \ll \zeta_c$, the overlap is quite large (of the order of k^{-1}) in the vicinity of the critical line $k\ell_\ell \lesssim \zeta_c$ indicating a large volume of latent orders in the vicinity of the price. Interestingly, combined with a vanishing level of liquidity, the increased level of activity around the origin induces important fluctuations of the trade price. To illustrate this, figure 3(d) displays the numerically determined squared volatility of the trade price p_t as function of $k\ell_\ell$ in the

¹⁰ Note that in this case the solution, provided $k\ell_\ell > 1$, is also asymptotically unstable: $\lim_{\xi \rightarrow \infty} \phi_r(\xi) = \mathcal{L}g(k\ell_\ell)\xi e^{-\xi/\ell_\ell} / (2k\ell_\ell) > 0$, while for $k\ell_\ell = 1$, $\lim_{\xi \rightarrow \infty} \phi_r < 0$.

How does latent liquidity get revealed in the limit order book?

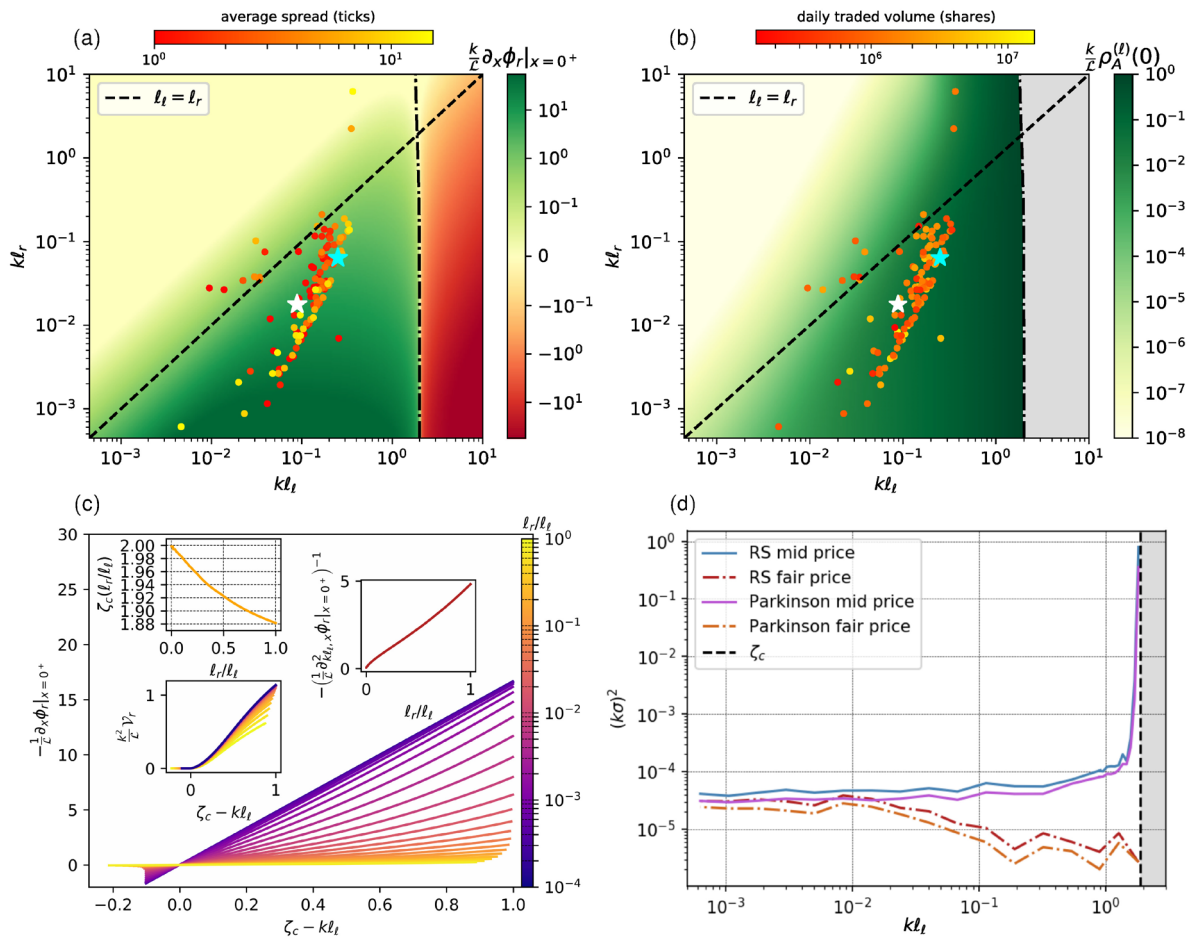


Figure 3. Parametric study of stationary order books. (a) Stability map. Density plot (symlog scale) of the rescaled slope of the revealed order density at the origin, as function of $k\ell_\ell$ and $k\ell_r$. The dashed line indicates $\ell_r = \ell_\ell$, the dash-dotted line indicates the critical line $\phi_r'(0) = 0$. (b) Overlap of the latent order books. Density plot (log scale) of the y -intercept of the latent order book densities at the origin. Note that such a quantity is a direct measure of the overlap. Typical computational time to draw the stability map: 1 h. The empirical order book data (US stocks, see table 1, and Euro Stoxx, white star) are reported with a colormap scaling with their average spread value (see (a)) and with their average daily traded volume (see (b)). The light blue star results from the calibration of Euro Stoxx data 2h around the flash crash of February 5th 2018 at 8:15pm (Paris time). Typical computational time: 1 min for each asset. (c) Plot of the slope of the revealed order book at the origin as function of $\zeta_c - k\ell_\ell$, for different values of ℓ_r/ℓ_ℓ . The top inset shows ζ_c as function of ℓ_r/ℓ_ℓ , and the center left inset displays the total volume in the revealed order book $\mathcal{V}_r = \left| \int_0^\infty d\xi \phi_r(\xi) \mathbb{1}_{\phi_r(\xi) < 0} \right|$, as function of $\zeta_c - k\ell_\ell$. The right inset shows the angular coefficient of the slope of the main plot at the origin. (d) Numerical volatility. Plot of rescaled numerical squared volatility of the trade price and the fair price (see equation (11)) in the $\ell_r = \ell_\ell$ case. We display two different estimations: Rogers–Satchell, $\sigma_{RS}^2 = \mathbb{E}[(p_H - p_O)(p_H - p_C) + (p_L - p_O)(p_L - p_C)]$, and Parkinson, $\sigma_P^2 = \frac{1}{4 \ln(2)} \mathbb{E}[(p_H - p_L)^2]$ where p_H, p_L, p_O, p_C denote the high, low, open and close prices respectively [17, 18].

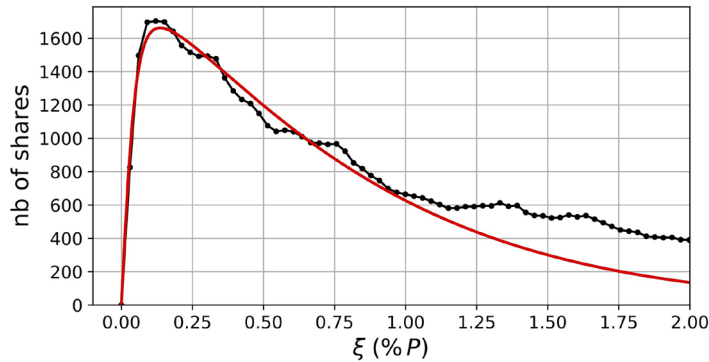


Figure 4. Fit of the stationary revealed order book ϕ_r to the average Euro Stoxx futures contract order book. The calibration yields $\mathcal{L} = 4599$, $k = 2.12$, $\ell_\ell = 0.042$ and $\ell_r = 0.0084$. Typical computational time: 1 min to obtain the average order book of the asset and apply the fitting procedure.

$\ell_r = \ell_\ell$ limit. For comparison, we also plotted the volatility of the *fair price* p_t^f , here defined as the value that equilibrates total (revealed and latent) supply and demand:

$$\int_0^{p_t^f} dx \left[\rho_A^{(r)}(x, t) + \rho_A^{(\ell)}(x, t) \right] = \int_{p_t^f}^{\infty} dx \left[\rho_B^{(r)}(x, t) + \rho_B^{(\ell)}(x, t) \right]. \quad (11)$$

As one can see, for $\omega \gg \omega_c := D_\ell k^2 / \zeta_c^2$ the volatility of the trade price coincides with its fair price counterpart, consistent with the idea that for high conversion rates the coupling between the revealed and latent books is almost instantaneous and therefore the mid-price tends to follow the fair price. In the vicinity of the critical line the volatility of the fair price slightly decreases. However, the volatility of the trade price strongly diverges as the vanishing liquidity limit is approached, i.e. when the conversion rate ω decreases to ω_c . Note that while the trade price can no longer be defined when a liquidity crisis arises, the fair price as defined in equation (11) remains well behaved. We also investigated the volatility in the $\ell_r = 0$ limit and obtained similar qualitative results, only with weaker overall volatility levels consistent with liquidity concentration around the trade price.

At this point, let us summarise our results. The market is most stable when the conversion rate is large and/or the latent orders diffusivity is small. In these cases, there is a good level of revealed liquidity and the trade price follows the fair price. However, when lag effects become important, and more particularly when the conversion time ω^{-1} becomes longer than the typical diffusion time of a latent order across the revealed order book depth k^{-1} , the order book empties out and the trade price undergoes large excursions away from the fair price, see also [15, 19]. As for the effect of diffusion in the real order book, a small ℓ_r concentrates the liquidity around the origin, therefore providing a wall to price fluctuations, while a large ℓ_r induces a weaker revealed order book slope around the origin, facilitating larger price excursions. Thus, our framework in the vicinity of the critical line could provide an interesting scenario to understand the nature of liquidity crises in financial markets. One indeed expects that in the presence of increased uncertainty, the conversion rate will decrease (as investors remain on the sidelines) and latent orders diffusivity will increase, both effects driving the market towards a liquidity crisis.

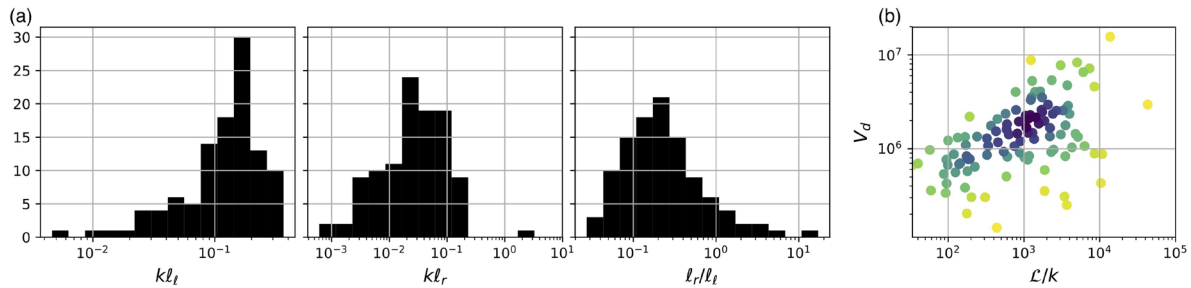


Figure 5. (a) Histograms of the fitted parameters kl_ℓ , kl_r and l_r/l_ℓ for the US stocks presented in table 1. (b) Average daily traded volume V_d as function of the liquidity parameter \mathcal{L}/k . The data points are coloured by density (from yellow to dark blue).

3.2. Order book data

To illustrate how our model allows one to calibrate latent parameters using visible order book data, we calibrate it to the real order book of the Euro Stoxx futures contract and of a set of over 100 large cap US stocks extracted from their primary market on the period August 2017 to April 2018, see table 1. For each asset, we take a snapshot of the order book every ten minutes (every minute for the Euro Stoxx contract), on regular trading hours. We then fold the bid side onto the ask side, the origin being taken at the opposite best [20], we average over the whole sample and rescale the x -axis by the average price taken over the sample period. Finally, fitting the stationary order book ϕ_r (see figures 4, 8–10) outputs four parameters: \mathcal{L} , k , l_ℓ and l_r (see table 1). Note that k^{-1} , l_ℓ and l_r have units of %op, and \mathcal{L} is in shares per %op.

The fitted values of kl_ℓ and kl_r for the Euro Stoxx contract and the US stocks presented in table 1 are reported in figures 3(a) and (b). Figure 5(a) displays the statistics of kl_ℓ , kl_r and l_r/l_ℓ for the stocks. As one can see the values of kl_ℓ stand below the critical line by typically one order of magnitude. Also most of the data is consistent with $l_r < l_\ell$. More precisely, the ratio l_r/l_ℓ is typically ≈ 0.1 – 0.3 , which is D_r/D_ℓ typically ≈ 0.01 – 0.1 , consistent with our initial intuition that diffusivity is much smaller in the revealed order book than in its latent counterpart. While not a strong effect, the l_r/l_ℓ ratios are on average slightly smaller for the small tick assets (figure 3(a)). The daily traded volume V_d does not seem to play a notable role in the distribution of the data in the (kl_r, kl_ℓ) map (figure 3(b)). However, \mathcal{L}/k shows positive correlation with V_d (see figure 5(b)), consistent with the interpretation given to the liquidity parameter \mathcal{L} in the latent order book models [12]. In order to evaluate the potential relevance of our model to quantitatively account for market stability, we have calibrated the Eurostoxx data on a particularly agitated subset, that is the two hours about the flash crash of February 5th 2018 at 8:15pm (Paris time), see light blue star on pannels (a) and (b) of figure 3. One finds, as expected, that the new data point is closer to the critical line than the reference point (white star), indicating lesser stability during the flash crash period.

4. Price impact

In this section we study how liquidity reacts, in our model, to the presence of a metaorder, namely a large trading order split into small orders executed incrementally. Following Donier *et al* [12] we introduce a metaorder as an additional current of buy/sell particles falling precisely at the trade price. The system of equations governing the system in the presence of a metaorder is left unchanged for the latent order book (equations (4a) and (4b)), while the RHS of equation (4c) must be complemented with the extra additive term $+ m_t \delta(x - p_t)$, representing the metaorder, with m_t the execution rate and δ the Dirac delta function. In the following we restrict to buy metaorders with constant execution rates $m_t = m_0 > 0$, without loss of generality since the sell metaorder $m_0 < 0$ is perfectly symmetric. In order to extract the dimensionless parameters governing the dynamic system, we introduce $\tilde{\xi} = k\xi$, $\tilde{t} = \omega t$, $\tilde{\rho} = k\rho/\mathcal{L}$, $\tilde{\phi}_r = k\phi_r/\mathcal{L}$ and write the equations in a dimensionless form:

$$\partial_{\tilde{t}} \tilde{\rho}_B^{(\ell)} = (k\ell_\ell)^2 \partial_{\tilde{x}\tilde{x}} \tilde{\rho}_B^{(\ell)} - \left\{ \Gamma(-\tilde{\xi}) \tilde{\rho}_B^{(\ell)} - [1 - \Gamma(-\tilde{\xi})] \mathbb{1}_{\{\tilde{x} < \tilde{p}_{\tilde{t}}\}} \tilde{\phi}_r \right\} \quad (12a)$$

$$\partial_{\tilde{t}} \tilde{\rho}_A^{(\ell)} = (k\ell_\ell)^2 \partial_{\tilde{x}\tilde{x}} \tilde{\rho}_A^{(\ell)} - \left\{ \Gamma(\tilde{\xi}) \tilde{\rho}_A^{(\ell)} + [1 - \Gamma(\tilde{\xi})] \mathbb{1}_{\{\tilde{x} > \tilde{p}_{\tilde{t}}\}} \tilde{\phi}_r \right\} \quad (12b)$$

$$\partial_{\tilde{t}} \tilde{\phi}_r = (k\ell_r)^2 \partial_{\tilde{x}\tilde{x}} \tilde{\phi}_r - \left\{ \Gamma(\tilde{\xi}) \tilde{\rho}_A^{(\ell)} - \Gamma(-\tilde{\xi}) \tilde{\rho}_B^{(\ell)} + [1 - \Gamma(|\tilde{\xi}|)] \tilde{\phi}_r \right\} + (m_0/\mathcal{J}) \delta(\tilde{\xi}), \quad (12c)$$

with $\mathcal{J} = \mathcal{L}\omega/k^2$ the typical scale of the rate at which latent orders are revealed (recall that \mathcal{L}/k^2 is the typical available volume in the latent order book that have a substantial probability to be revealed). Matching the first and third terms on the right hand side of equation (12c) yields a relevant dimensionless number $(m_0/\mathcal{J})/(k\ell_r)^2 = m_0/J_r$ with $J_r = D_r \mathcal{L}$. In the case $\ell_r = 0$ the relevant dimensionless number is simply given by m_0/\mathcal{J} .

In order to compute the price impact $I(Q_t) = \mathbb{E}[p_t - p_0 | Q_t = m_0 t]$, we performed numerical simulations of our model in the presence of a metaorder in several limit cases. More specifically, starting from the stationary order books, we implemented a buy/sell metaorder as an additional current of buy/sell particles falling precisely at the best ask/bid, thus acting as a sequence of market orders. We stored the order books as a function of time to be able to compute dynamically the relevant scalar quantities below. We explored in particular $\ell_r = \ell_\ell$ and $\ell_r = 0$, in both high and low participation rate regimes, for different values of $k\ell_\ell$ (see figures 6 and 7).

Before presenting the results of the numerical simulations, note that there exists a regime where the calculations can be brought a little bit further analytically, that is when we can give a geometrical interpretation to the problem. When the book is almost static on the time scale of the metaorder execution i.e. $m_0 \gg J_r$ (resp. $m_0 \gg \mathcal{J}$ for $\ell_r = 0$), one has $\int_0^t ds m_0 = - \int_0^{p_t - p_0} d\xi \phi_r^{st}(\xi)$ with $\xi = x - p_0$. In particular for $\ell_r = 0$ one obtains:

$$Q_t = -\frac{\mathcal{L}\ell_\ell^2}{2} (1 - e^{-p_t/\ell_\ell}) + \frac{\mathcal{L}p_t}{k} \log(1 - e^{-kp_t}) - \frac{\mathcal{L}}{k^2} [\text{Li}_2(e^{-kp_t}) - \text{Li}_2(1)], \quad (13)$$

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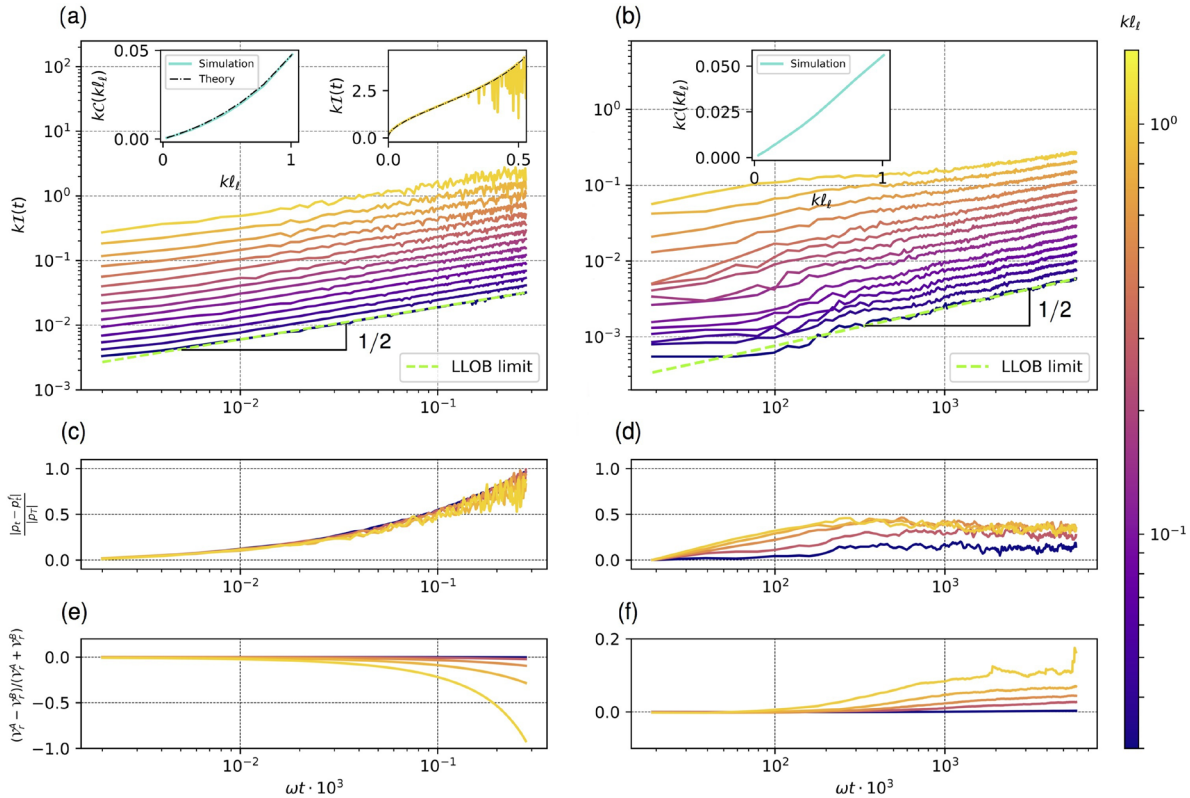


Figure 6. Price impact for $\ell_r = \ell_\ell$. (a) $m_0 \gg J_r$ (b) $m_0 \ll J_r$. The dashed green lines indicate the LLOB limits, $I_{\text{LLOB}}(t) = \sqrt{\alpha Q_t / (\pi \mathcal{L})}$ with $\alpha = m_0 / J_u$ for the slow regime and $I_{\text{LLOB}}(t) = \sqrt{2Q_t / \mathcal{L}}$ for the fast regime. The top left insets on each plot indicate the factor c as function of $k\ell_\ell$ defined as $I(t) = c I_{\text{LLOB}}(t)$. The top right inset of subplot (a) shows an extreme regime with very high execution rate, the dash-dotted line indicates the theoretical prediction as given by the numerical inversion of equation (14). Subplots (c) and (d) display relative price difference between the trade price and the fair price (see equation (11)). Subplots (e) and (f) display the relative revealed volume imbalance. Typical computational time: a few hours to obtain all the impact curves.

where $\text{Li}_2(y) = \sum_{k=1}^{\infty} y^k / k^2$ stands for the polylogarithm of order 2, and can be inverted numerically to obtain the price trajectory p_t . Similarly, for $\ell_r = \ell_\ell$ one has:

$$Q_t = \frac{\mathcal{L}\alpha}{k^2} \{ [k(p_t + \beta) + 1]e^{-kp_t} - (k\beta + 1) \} - \frac{\mathcal{L}\ell_\ell\gamma}{1 + k\ell_\ell} (1 - e^{-(k+1/\ell_\ell)p_t}) + \mathcal{L}\ell_\ell\eta p_t e^{-p_t/\ell_\ell} - \mathcal{L}[\eta\ell_\ell^2 - \ell_\ell(\alpha\beta + \gamma)] (1 - e^{-p_t/\ell_\ell}), \quad (14)$$

where $\alpha = [(k\ell_\ell)^2 - 1]^{-1}$, $\beta = 2\alpha k\ell_\ell^2$, $\gamma = g(k\ell_\ell)[k^2\ell_\ell(k\ell_\ell + 2)]^{-1}$, $\eta = g(k\ell_\ell)[2k\ell_\ell]^{-1}$. In the following we discuss the more general numerical results for both limit cases $\ell_\ell = \ell_r$ and $\ell_r = 0$.

4.1. The case $\ell_\ell = \ell_r$

The main plots in figures 6(a) and (b) display robust square root price trajectories, regardless of the values of $k\ell_\ell$. For $m_0 \gg J_r$ the price trajectory matches the theoretical

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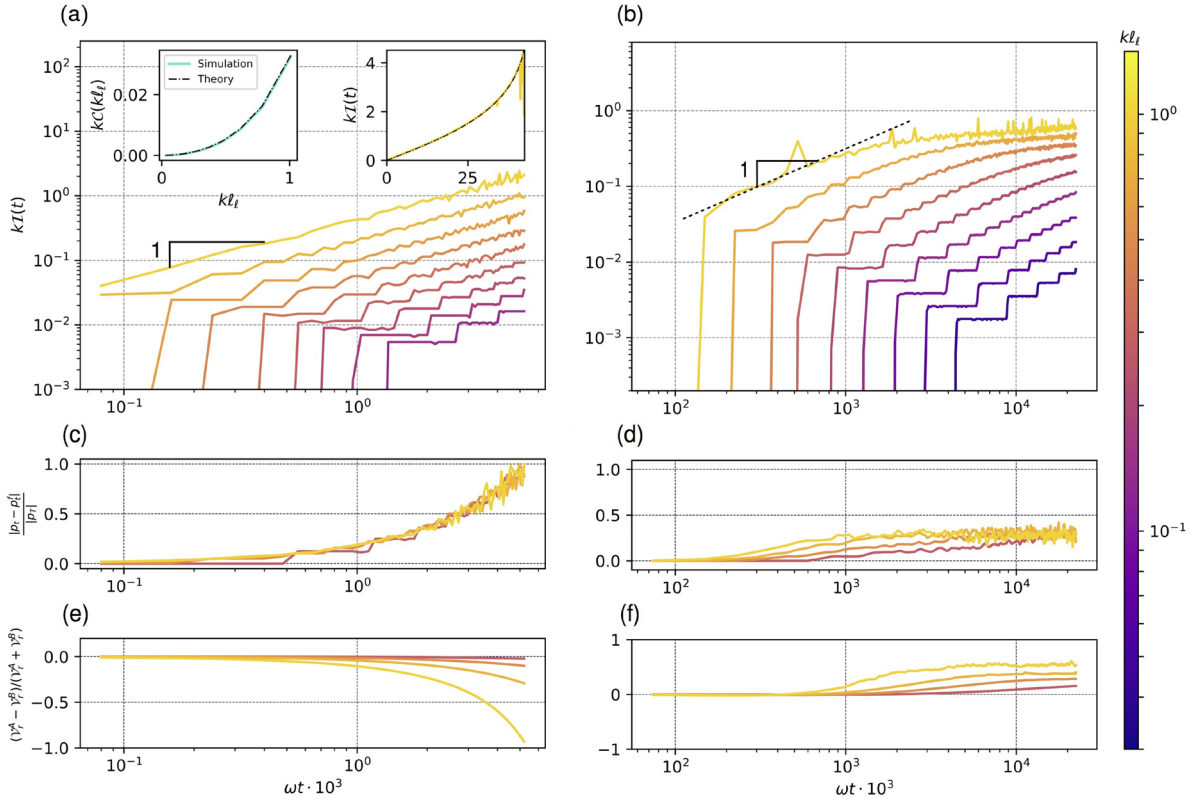


Figure 7. Price impact for $\ell_r = 0$. (a) and $m_0 \gg \mathcal{J}$ (b) $m_0 \ll \mathcal{J}$. The top left insets on subplot (a) indicates the factor c as function of kl_ℓ here defined as $I(t) = ct$ for $\ell_r = 0$. The top right insets show an extreme regime with very high execution rate for which the impact diverges. Subplots (c) and (d) display relative price difference between the trade price and the fair price, see equation (11). Subplots (e) and (f) display the relative revealed volume imbalance. Typical computational time: a few hours to obtain all the impact curves.

prediction given above inverting equation (14). As expected from the exponentially vanishing liquidity when $x - p_0 > k^{-1}$, impact eventually diverges for very extreme regimes, see top right inset in figure 6(a). For $kl_\ell \ll 1$, one recovers the LLOB limit in both fast and slow regimes, also as expected. For non vanishing values of kl_ℓ , the impact increases with increasing kl_ℓ . In particular for $m_0 \gg J_r$ one obtains at short times (equivalently small volumes):

$$p_t = \sqrt{\frac{2Q_t}{|\partial_x \phi_r(0^+)|}} \sim \sqrt{2Q_t} \left(1 - \frac{kl_\ell}{\zeta_c}\right)^{-1/2}. \quad (15)$$

As expected, impact diverges when at the incipient liquidity crisis point. Figures 6(c) and (d) display the relative distance between the trade price and the fair price as function of time. In the fast execution regime all curves fall on top of each other and $|p_t - p_t^f| \approx |p_t|$, consistent with the idea that the book (in particular latent) does not have time to reassess during the execution, and as a consequence the fair price varies at a much slower rate than the trade price. A different scenario takes place in the small execution rate regime. We observe that the relative distance between trade and

fair prices stabilizes. In other terms, the latent order book evolves at a speed that is comparable to that of the metaorder and the fair price follows the trade price quite accurately.

The evolution of relative volume imbalance (see figures 6(e) and (f)) allows one to draw similar conclusions. In the fast execution limit, the imbalance diverges as the latent order book has no time to refill the revealed order book (this effect is all the more evident as we approach the vanishing liquidity limit $k\ell_\ell \rightarrow \zeta_c$), while in the slow execution limit the imbalance is much smaller. Most importantly, note that in this limit the imbalance becomes positive. This is consistent with the fact that when the trade price moves slowly, the conversion probability Γ is shifted with it and new orders reveal on top of the existing ones to supply the metaorder, while the orders left behind progressively unreveal.

4.2. The case $\ell_r = 0$

The limit $\ell_r \rightarrow 0$ corresponds to $m_0/J_r \rightarrow \infty$, so in some sense one could say that we are always in a high participation rate regime. However the absence of diffusion means that the system can only evolve through the revealing and unrevealing currents. As mentioned above, in this case the relevant dimensionless number becomes m_0/\mathcal{J} , that shall be referred to as participation rate in the following.

Figure 7(a) displays price trajectories in the fast execution regime, here $m_0 \gg \mathcal{J}$. The metaorder is faster than the revealing current and the price trajectory is given by inverting equation (13). At short times, one obtains (see equation (9)):

$$p_t = \frac{kQ_t}{\mathcal{L}} \left(1 - \frac{k\ell_\ell}{\zeta_c} \right)^{-1}, \quad (16)$$

which is linear impact, consistent with $\phi_r(0) \neq 0$. Analogous to the case $\ell_\ell = \ell_r$, in extreme regimes the impact eventually diverges (see top right inset of figure 7(a)). Regarding the imbalance and the relative distance between the trade price and the fair price, the interpretation is analogous to the $\ell_r = \ell_\ell$ case.

Figure 7(b) displays the price trajectories in the low participation rate regime $m_0 \ll \mathcal{J}$. Here the impact is genuinely linear at short times but crossovers to concave after a typical time $t^* \sim \mathcal{V}_r/m_0$ with \mathcal{V}_r the typical volume available in the revealed order book. This interesting regime can be easily understood as follows. At short times the metaorder is executed against the orders present in the stationary locally constant revealed order book (linear impact), but after a while liquidity revealing from the linear latent order book takes over (implying concave impact)¹¹. More precisely, as $k\ell_\ell$ is increased, t^* is decreased consistent with the idea that: the larger $k\ell_\ell$, the smaller the revealed liquidity \mathcal{V}_r and thus the sooner it gets consumed. Note that linear impact at short times (equiv. small volumes) has been reported empirically in the literature [21]. An alternative theoretical framework to understand linear impact at short times is provided in [13] with the introduction of two types of agents, fast and slow. Our

¹¹ Note that recovering precisely the square root law in this regime is quite challenging because the numerical simulation, by essence discontinuous, induces artificial spread effects: the spread widens and does not get refilled as fast as it should, due to limited resolution and vanishing liquidity in the price region. Obtaining smooth results requires a lot of averaging.

model, in its final version, should also include the possibility for agents with different timescales, possibly with different conversion rates ω . In figure 7(d), we observe that the relative distance between trade and fair prices stabilizes after the typical time t^* . In other terms, for $t > t^*$ the latent order book evolves at a rate comparable to that of the metaorder, and the fair price follows the trade price with some constant lag. Figure 7(f) displays similar conclusions.

5. Concluding remarks

We have proposed a simple, consistent model describing how latent liquidity gets revealed in the real order book. As a first step in the rapprochement of latent order book models and real order book data, our study upgrades the former from a *toy model* status to an observable setup that can be quantitatively calibrated on real data. Although probably too simple in its present form, our main contribution is to show how the latent order book could be inferred from its revealed counterpart. One of our key theoretical result is the existence of a market instability threshold, where the conversion of the trading intentions of market participants into *bona fide* limit orders becomes too slow, inducing *liquidity crises*. From a regulatory perspective, our model indicates how assets can be sorted according to their position in the stability map, a proxy for their propensity to liquidity dry-ups and large price jumps. In particular, our setup could constitute an insightful alternative to relate stability and tick size, a subject that has raised the attention of many in the recent past (see e.g. [22]). To confirm potential perspectives for market regulation, we showed that our model exhibits lesser stability when calibrating the same asset during a period of higher volatility.

While providing quite satisfactory results on several grounds, our model lacks a number of features that must be addressed in future work. In particular, being an inheritance of the LLOB model [12], our model does not solve the so-called diffusivity puzzle (even persistent order flow lead to mean-reverting prices—see [6, 13, 14]). An interesting extension is suggested by the stocks data presented in figures 8–10. Indeed, some of the order books display a bimodal shape indicating that they would be better fitted with a model involving two conversion timescales (equivalently two price scales). The recent multi-timescale liquidity setup (see [13]) allows for both a *fast* liquidity of high frequency market makers, and a *slow* liquidity of directional traders, while resolving several difficulties of latent order book models. Such a framework should output bimodal distributions and yield even better fits of the real order book, allowing one to infer different liquidity timescales.

Acknowledgments

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Appendix

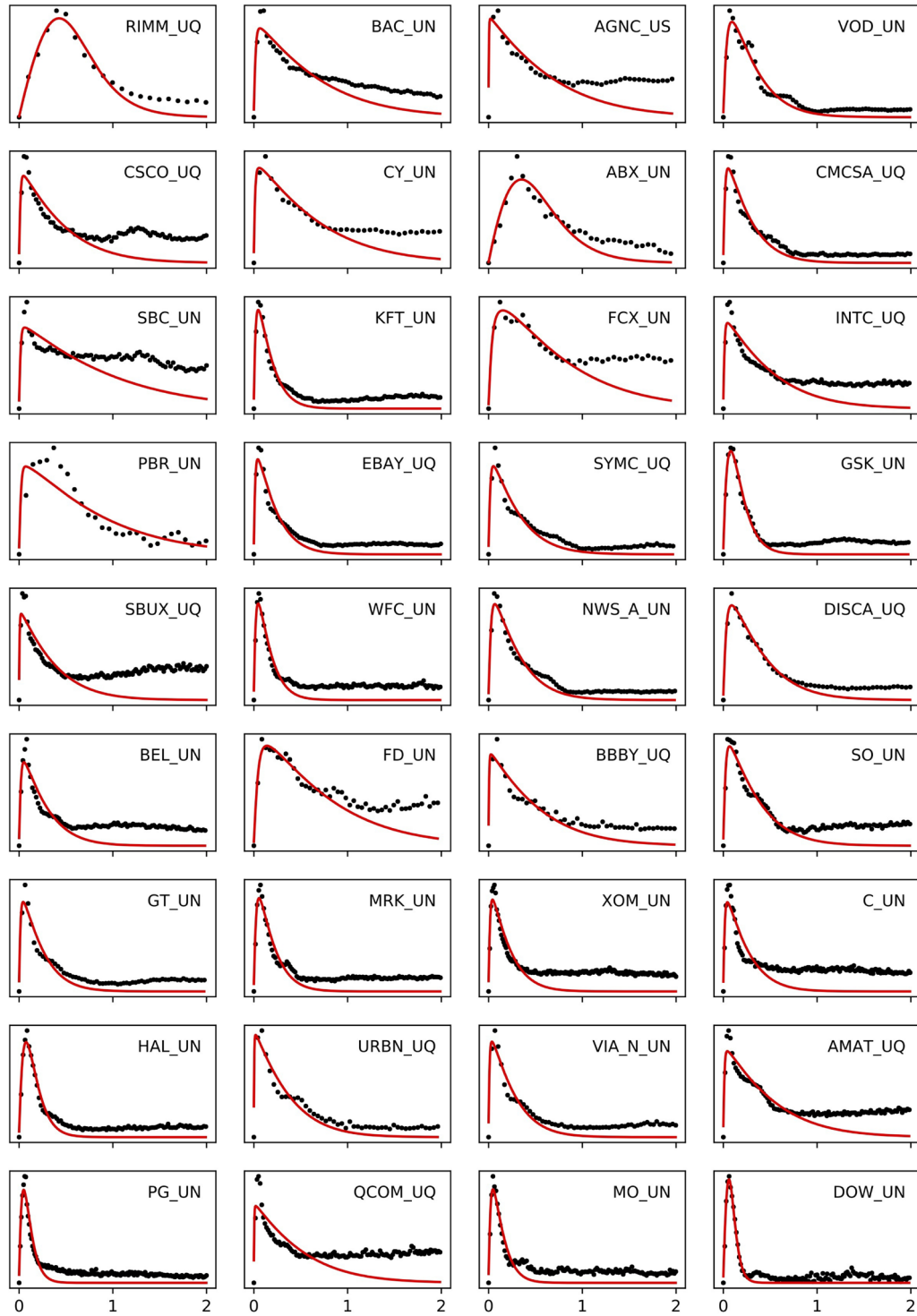
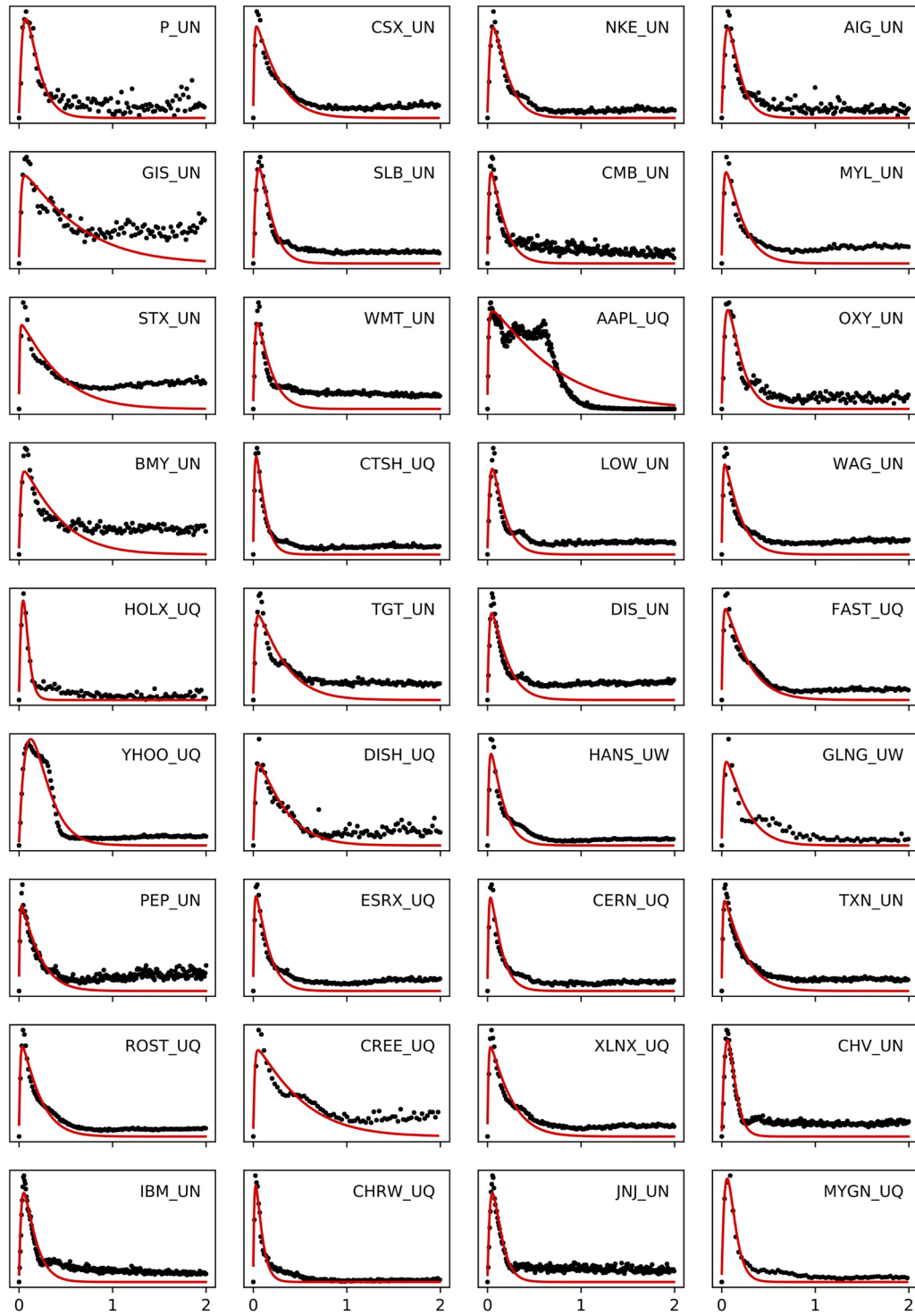


Figure 8. Fit of the stationary revealed order book ϕ_r to the average order books of over one hundred US stocks, see table 1.

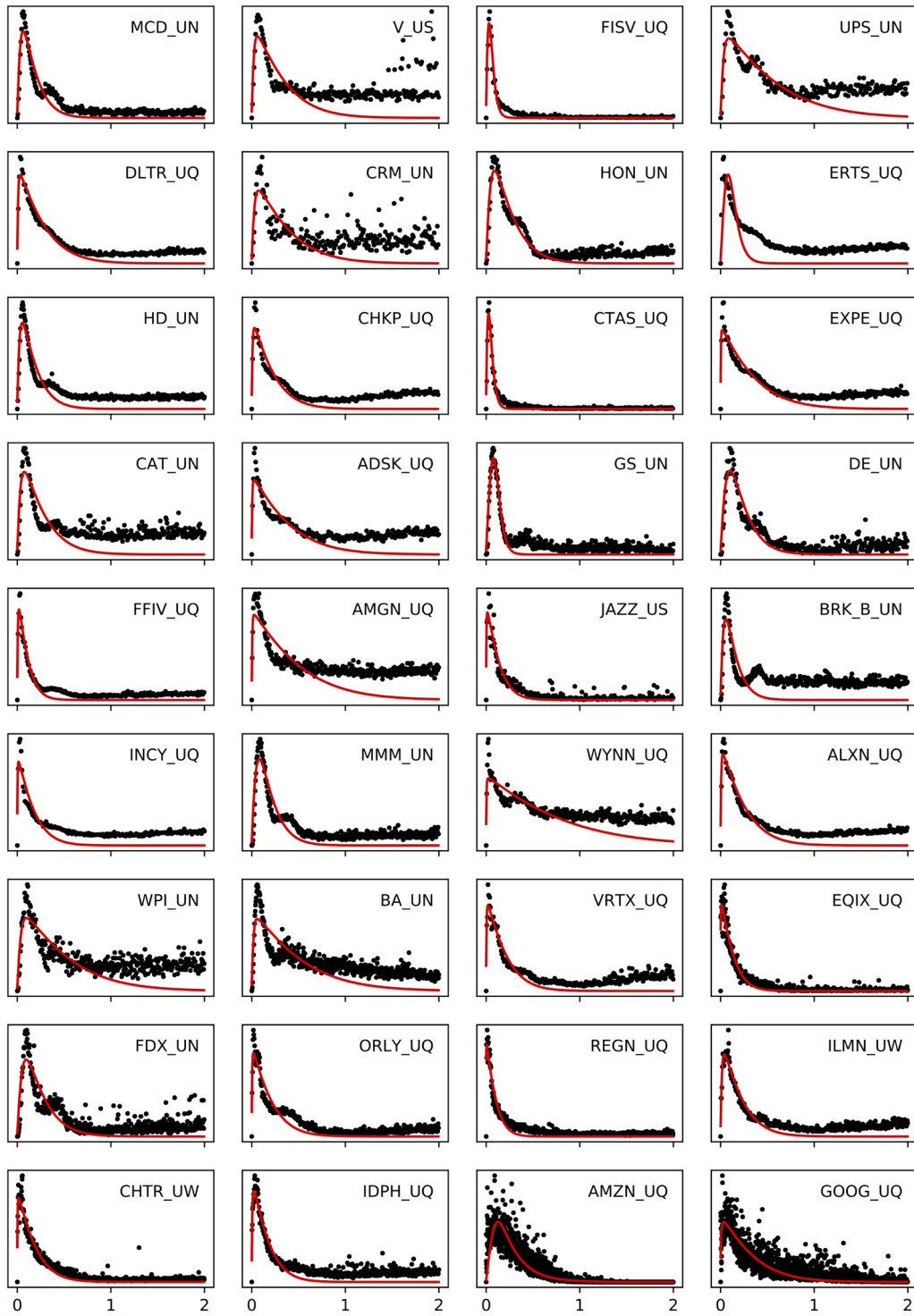
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Figure 9. Fit of the stationary revealed order book ϕ_r to the average order books of over one hundred US stocks, see table 1.

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Figure 10. Fit of the stationary revealed order book ϕ_r to the average order books of over one hundred US stocks, see table 1.

Table 1. US stocks (large cap.) used for the empirical analysis of section 3.2 (see also figures 8–10). The average spread S has tick units, V_d denotes the average daily traded volume (in shares), k^{-1} , ℓ_ℓ and ℓ_r are expressed in %price, and \mathcal{L} is shares per unit %price.

Stock	price \$	S	$10^{-6}V_d$	$10^{-3}\mathcal{L}$	k	ℓ_ℓ	ℓ_r
RIMM	10.03	1.003	0.87	602.0	4.71	0.0020	0.0059
BAC	28.66	1.007	15.62	34.3	2.50	0.0350	0.0033
AGNC	19.92	1.009	1.03	10.8	2.51	0.0165	0.0005
VOD	29.62	1.011	0.89	47.6	5.58	0.0230	0.0090
CSCO	38.57	1.012	7.17	26.7	3.64	0.0309	0.0027
CY	16.14	1.018	1.44	11.9	2.49	0.0195	0.0020
ABX	16.46	1.023	2.95	191.1	4.45	0.0031	0.0060
CMCSA	37.65	1.023	6.54	38.3	6.31	0.0230	0.0046
SBC	36.51	1.029	7.76	5.3	1.75	0.0417	0.0023
KFT	42.25	1.036	2.88	39.1	10.22	0.0131	0.0047
FCX	16.45	1.040	4.74	8.3	2.22	0.0201	0.0054
INTC	44.87	1.042	8.31	16.7	3.31	0.0318	0.0024
PBR	13.54	1.044	4.59	16.9	1.99	0.1284	0.0035
EBAY	39.12	1.052	2.73	28.1	7.71	0.0176	0.0037
SYMC	28.96	1.072	1.78	19.4	5.50	0.0227	0.0040
GSK	37.99	1.074	0.83	38.4	10.94	0.0084	0.0069
SBUX	57.01	1.076	2.95	9.4	4.53	0.0178	0.0009
WFC	56.31	1.082	5.37	25.6	11.02	0.0156	0.0057
NWS	33.21	1.100	2.55	18.7	5.74	0.0262	0.0065
DISCA	22.57	1.101	1.42	13.7	4.46	0.0247	0.0072
BEL	49.35	1.104	4.09	10.0	6.11	0.0278	0.0048
FD	23.44	1.117	2.46	3.9	2.26	0.0411	0.0093
BBBY	22.20	1.118	1.35	7.0	3.17	0.0178	0.0008
SO	44.21	1.143	1.69	5.2	4.87	0.0323	0.0060
GT	31.36	1.191	0.98	17.8	7.24	0.0169	0.0035
MRK	56.76	1.210	3.54	11.5	9.16	0.0209	0.0071
XOM	80.51	1.219	4.03	11.2	8.03	0.0231	0.0042
C	73.26	1.225	5.28	8.8	6.83	0.0261	0.0040
HAL	47.17	1.254	2.35	28.7	11.94	0.0117	0.0097
URBN	23.44	1.257	0.92	10.5	4.57	0.0123	0.0007
VIA	29.54	1.264	1.66	14.2	6.45	0.0127	0.0020
AMAT	53.43	1.315	3.53	5.6	3.23	0.0305	0.0021
PG	84.03	1.401	2.60	30.7	17.95	0.0098	0.0078
QCOM	59.38	1.416	3.30	4.0	3.25	0.0177	0.0006
MO	64.07	1.428	2.09	14.9	13.02	0.0143	0.0064
DOW	64.99	1.439	2.35	100.6	25.26	0.0015	0.0030
P	52.99	1.439	1.85	17.5	11.15	0.0149	0.0104
CSX	54.49	1.561	2.16	9.8	6.77	0.0199	0.0026
NKE	63.01	1.586	2.29	12.4	10.24	0.0191	0.0088
AIG	61.12	1.592	1.48	15.9	12.27	0.0168	0.0109
GIS	49.88	1.645	1.17	1.3	2.79	0.0431	0.0037
SLB	67.36	1.661	2.20	14.9	11.92	0.0167	0.0097
CMB	106.81	1.667	4.03	8.6	11.14	0.0199	0.0048
MYL	39.10	1.748	1.98	16.1	8.65	0.0184	0.0049
STX	45.28	1.822	1.48	4.6	4.09	0.0212	0.0013
WMT	90.73	1.884	2.59	8.3	10.61	0.0211	0.0055
AAPL	168.14	1.922	8.78	3.1	2.51	0.0107	0.0011
OXY	64.65	1.971	1.20	9.6	10.77	0.0191	0.0101
BMJ	61.12	2.032	1.76	1.5	4.21	0.0372	0.0044
CTSH	75.94	2.129	1.27	23.7	17.41	0.0095	0.0040
LOW	81.69	2.204	1.61	6.9	11.71	0.0194	0.0078

(Continued)

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Table 1. (Continued)

Stock	price \$	S	$10^{-6} V_d$	$10^{-3} \mathcal{L}$	k	ℓ_ℓ	ℓ_r
WAG	71.64	2.298	2.25	10.3	9.79	0.0176	0.0032
HOLX	42.09	2.314	0.88	317.2	29.35	0.0008	0.0012
TGT	66.76	2.409	1.53	2.4	5.62	0.0320	0.0050
DIS	103.38	2.440	2.36	5.4	9.51	0.0236	0.0057
FAST	50.13	2.495	0.91	5.9	6.68	0.0213	0.0032
YHOO	71.07	2.649	2.55	21.6	8.05	0.0042	0.0043
DISH	48.54	2.671	0.82	3.3	5.53	0.0309	0.0059
HANS	59.37	2.751	0.84	13.8	11.79	0.0163	0.0047
GLNG	27.57	2.925	0.35	14.1	7.52	0.0187	0.0048
PEP	112.52	3.009	1.42	5.3	8.42	0.0174	0.0020
ESRX	69.15	3.032	1.45	9.6	11.69	0.0137	0.0029
CERN	67.81	3.100	0.81	13.6	13.42	0.0134	0.0037
TXN	99.90	3.131	1.82	4.6	7.61	0.0200	0.0025
ROST	70.66	3.149	1.06	6.2	8.32	0.0198	0.0033
CREE	34.92	3.165	0.51	2.0	3.48	0.0322	0.0030
XLNX	70.21	3.199	0.93	4.0	7.34	0.0200	0.0026
CHV	116.12	3.867	1.95	16.8	19.58	0.0085	0.0108
IBM	152.11	3.917	1.29	3.9	11.96	0.0218	0.0083
CHRW	87.75	3.980	0.59	41.7	22.18	0.0075	0.0034
JNJ	131.86	4.007	2.07	6.9	15.56	0.0149	0.0098
MYGN	33.28	4.032	0.31	61.1	17.84	0.0019	0.0021
MCD	156.99	4.131	1.07	3.5	10.38	0.0276	0.0109
V	121.14	4.170	2.20	1.0	5.21	0.0381	0.0058
FISV	106.11	4.408	0.43	377.8	36.42	0.0008	0.0010
UPS	107.88	5.055	0.87	0.4	3.02	0.0566	0.0071
DLTR	97.27	5.124	0.86	2.0	6.07	0.0206	0.0019
CRM	116.98	5.563	1.22	0.5	4.61	0.0452	0.0074
HON	148.04	5.754	0.78	1.4	7.36	0.0365	0.0151
ERTS	116.94	5.872	1.07	220.6	34.96	0.0101	0.0641
HD	175.53	5.944	1.35	2.4	9.83	0.0275	0.0079
CHKP	104.95	6.161	0.38	1.3	7.57	0.0183	0.0020
CTAS	161.47	6.443	0.25	141.6	38.58	0.0038	0.0040
EXPE	124.49	6.716	0.83	0.8	4.82	0.0114	0.0006
CAT	153.78	6.934	1.10	1.2	6.78	0.0408	0.0106
ADSK	118.11	6.967	0.77	0.8	4.34	0.0181	0.0010
GS	230.09	6.976	0.76	30.5	27.22	0.0011	0.0038
DE	154.44	6.985	0.58	1.6	8.70	0.0338	0.0216
FFIV	128.76	7.457	0.30	4.9	15.98	0.0095	0.0016
AMGN	179.67	7.504	1.31	0.5	3.42	0.0187	0.0009
JAZZ	146.67	8.266	0.14	5.6	12.72	0.0064	0.0007
BRK	202.45	8.669	1.34	2.1	12.53	0.0257	0.0107
INCY	97.51	8.723	0.64	2.2	9.68	0.0095	0.0007
MMM	222.45	8.976	0.70	1.4	9.74	0.0334	0.0167
WYNN	162.67	9.496	0.76	0.2	2.17	0.0106	0.0004
ALXN	122.92	9.738	0.69	0.9	6.73	0.0154	0.0013
WPI	176.75	10.580	0.70	0.1	3.37	0.0575	0.0093
BA	297.60	10.630	0.97	0.2	3.31	0.0404	0.0037
VRTX	155.65	11.509	0.56	1.0	8.18	0.0105	0.0009
EQIX	433.69	11.889	0.20	2.3	12.98	0.0015	0.0002
FDX	246.89	12.048	0.36	0.4	7.53	0.0445	0.0183
ORLY	224.90	12.416	0.43	0.8	8.09	0.0119	0.0009
REGN	372.48	12.480	0.30	3.4	16.62	0.0028	0.0002
ILMN	217.03	12.989	0.34	0.7	7.93	0.0178	0.0033
CHTR	344.48	13.312	0.67	0.9	8.88	0.0060	0.0005
IDPH	307.01	13.479	0.54	0.8	9.68	0.0099	0.0014
AMZN	1097.54	14.080	1.34	160.6	31.19	0.0117	0.2002
GOOG	1009.80	14.339	0.66	0.2	5.03	0.0009	0.0001

References

- [1] Kanazawa K, Sueshige T, Takayasu H and Takayasu M 2018 *Phys. Rev. Lett.* **120** 138301
- [2] Tóth B, Lemperiere Y, Deremble C, De Lataillade J, Kockelkoren J and Bouchaud J P 2011 *Phys. Rev. X* **1** 021006
- [3] Torre N 1997 *Market Impact Model Handbook* (Berkeley: BARRA Inc.)
- [4] Almgren R, Thum C, Hauptmann E and Li H 2005 *Risk* **18** 5862
- [5] Engle R, Ferstenberg R and Russell J 2006 SSRN
- [6] Mastromatteo I, Tóth B and Bouchaud J P 2014 *Phys. Rev. E* **89** 042805
- [7] Brokmann X, Serie E, Kockelkoren J and Bouchaud J P 2015 *Mark. Microstruct. Liquidity* **1** 1550007
- [8] Bacry E, Iuga A, Lasnier M and Lehalle C A 2015 *Mark. Microstruct. Liquidity* **1** 1550009
- [9] Bershova N and Rakhlin D 2013 *Quant. Finance* **13** 1759–78
- [10] Lehalle C A, Guéant O and Razafinimanana J 2011 High-frequency simulations of an order book: a two-scale approach *Econophysics of Order-Driven Markets* (Berlin: Springer) pp 73–92
- [11] Bak P, Paczuski M and Shubik M 1997 *Phys. A: Stat. Mech. Appl.* **246** 430–53
- [12] Donier J, Bonart J F, Mastromatteo I and Bouchaud J P 2015 *Quant. Finance* **15** 1109–21
- [13] Benzaquen M and Bouchaud J P 2018 *Quant. Finance* **18** 1781–90
- [14] Benzaquen M and Bouchaud J P 2018 *Eur. Phys. J. B* **91** 23
- [15] Corradi F, Zaccaria A and Pietronero L 2015 *Phys. Rev. E* **92** 062802
- [16] Smith E *et al* 2003 *Quant. Finance* **3** 481–514
- [17] Rogers L C, Satchell S E and Yoon Y 1994 *Appl. Finance Econ.* **4** 241–7
- [18] Parkinson M 1980 *J. Bus.* **61**–5
- [19] Cristelli M, Alfì V, Pietronero L and Zaccaria A 2010 *Eur. Phys. J. B* **73** 41–9
- [20] Bouchaud J P, Bonart J, Donier J and Gould M 2018 *Trades, Quotes and Prices: Financial Markets Under the Microscope* (Cambridge: Cambridge University Press)
- [21] Zarinelli E, Treccani M, Farmer J D and Lillo F 2015 *Mark. Microstruct. Liquidity* **1** 1550004
- [22] Dayri K and Rosenbaum M 2015 *Mark. Microstruct. Liquidity* **1** 1550003