Hydrodynamic limit of move-to-front rules and LRU caching

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We study a hydrodynamic limit approach to move-to-front rules. Namely, we consider a stochastic process of particles aligned in a line, each of which jumps randomly to the top, and study a large particle number limit. A scaling limit of the joint empirical distribution of jump rate and position satisfies a system of Burgers type partial differential equations. Results are applied to the search cost probabilities for the least-recently-used caching in the data theory of computer sciences.

1. Move-to-front rules. Let N be a positive integer, and S_N a set of all permutations of $1, 2, \dots, N$, and define a Markov process $X^{(N)}(t) = (X_1^{(N)}(t), \dots, X_N^{(N)}(t)), t \ge 0,$

 $X^{(N)}(t) = (X_1^{(N)}(t), \dots, X_N^{(N)}(t)), \quad t \ge 0,$ with the state space \mathcal{S}_N , as follows. For each $i = 1, 2, \dots, N$, let $\tau_{i,j}^{(N)}$, $j = 1, 2, \dots,$ be an increasing sequence of random variables (jump times, to be specified shortly), and we define $X^{(N)}(t)$ to be constant in t for $t \notin \{\tau_{i,j}^{(N)} \mid i = 1, 2, \dots, N, j = 1, 2, \dots\}$. At a jump time $t = \tau_{i,j}$, we define $X_i^{(N)}(\tau_{i,j}) = 1$, and for $i' \neq i$,

$$X_{i'}^{(N)}(\tau_{i,j}) = X_{i'}^{(N)}(\tau_{i,j} - 0) + \begin{cases} 1, & \text{if } X_{i'}^{(N)}(\tau_{i,j} - 0) < X_i^{(N)}(\tau_{i,j} - 0), \\ 0, & \text{if } X_{i'}^{(N)}(\tau_{i,j} - 0) > X_i^{(N)}(\tau_{i,j} - 0). \end{cases}$$

For convenience, we put $\tau_{i,0}^{(N)} = 0$ ($\forall i$), and we define { $\tau_{i,j+1}^{(N)} - \tau_{i,j}^{(N)}$, $j = 0, 1, 2, \cdots$ } to be independent in *i* and *j*, identical in distribution for all *j*, whose distribution is the exponential distribution with parameter $w_i^{(N)} > 0$: P[$\tau_{i,1}^{(N)} > t$] = exp($-w_i^{(N)}t$). Note that as in the standard Poisson process, with probability 1 the jump times are different for different (i, j). This completes the definition of the process $X^{(N)}$.

In the following, we regard $X^{(N)}$ as an N particle system aligned on a single line, with the suffix i in $X_i^{(N)}(t)$ standing for the label of the particle and $X_i^{(N)}(t)$ denoting the position (rank) of the particle i at time t.

2. Hydrodynamic limit. We embed $S_N \subset \mathbb{R}_+$ and scale by N to consider a particle system in an interval [0,1); $Y_i^{(N)}(t) := \frac{1}{N}(X_i^{(N)}(t)-1)$. Note that $y_C^{(N)}(t) = \frac{1}{N} \sharp\{i \mid \tau_{i,1} \leq t\} \in [0,1)$ is the boundary of particles which jumped to the top position and those which has not jumped up to time t. In the following we denote by δ_a the unit distribution concentrated on a, and assume $\lambda^{(N)} := \frac{1}{N} \sum_{i=1}^{N} \delta_{w_i^{(N)}} \to \lambda$ weakly as $N \to \infty$, for a probability distribution λ . Proposition $\mathbf{1}([1])$. $y_C^{(N)}(t) \to y_C(t) := 1 - \int_0^{\infty} e^{-wt} \lambda(dw) \ (N \to \infty, \text{ in prob.})$. \diamond Consider a joint empirical distribution $\mu_t^{(N)} = \frac{1}{N} \sum_{i=1}^N \delta_{(w_i^{(N)}, Y_i^{(N)}(t))}$. Theorem $\mathbf{2}([1])$. Assume $\int_0^{\infty} w\lambda(dw) < \infty$ and $\lambda(\{0\}) = 0$, and assume that the initial distribution $\mu_0^{(N)}$ determined by the initial configuration $Y^{(N)}(0) = y^{(N)}$

converges weakly to a distribution μ_0 as $N \to \infty$. Then for each t > 0, there exists a deterministic distribution μ_t such that $\mu_t^{(N)} \to \mu_t$ as $N \to \infty$. μ_t is given by

$$U(dw, y, t) := \mu_t(dw, [y, 1)) = \begin{cases} \lambda(dw) e^{-wt_0(y)}, & y < y_C(t), \\ U(dw, \hat{y}(y, t), 0) e^{-wt}, & y > y_C(t), \end{cases}$$

where, $t = t_0(y)$ is the inverse function of $y = y_C(t)$, and $\hat{y}(y,t)$ is the inverse function in y of $y_C(y,t) = 1 - \int_y^1 \int_0^\infty e^{-wt} \mu_0(dw,dz)$. \diamond The assumption $\int_0^\infty w\lambda(dw) < \infty$ is unnecessary for the convergence at y > 0.

3. Burgers type equation. We succeeded in proving Theorem 2 by guessing the explicit formula for μ_t correctly, and then by proving the convergence. The explicit formula hence is of importance, which we found as a solution to a following system of PDEs. Consider the case where there are at most countable types of jump

rates; $\lambda = \sum_{\alpha} \rho_{\alpha} \delta_{f_{\alpha}}$, where f_{α} and ρ_{α} are positive constants, satisfying $\sum_{\alpha} \rho_{\alpha} = 1$. **Proposition 3**([2]) . $U_{\alpha}(y,t) := U(\{f_{\alpha}\}, y, t) = \mu_t(\{f_{\alpha}\}, [y, 1))$ is a unique classical time global solution to the following initial value problem:

$$\frac{\partial U_{\alpha}}{\partial t}(y,t) + \sum_{\beta} f_{\beta} U_{\beta}(y,t) \frac{\partial U_{\alpha}}{\partial y}(y,t) = -f_{\alpha} U_{\alpha}(y,t), \ (y,t) \in [0,1) \times [0,\infty), \ \alpha = 0$$

1, 2, ..., with boundary conditions $U_{\alpha}(0,t) = \rho_{\alpha}, t \geq 0$, $\alpha = 1, 2, \cdots$. For each α , the initial data $U_{\alpha}(y,0) = U_{\alpha}(y), 0 \leq y < 1$, are smooth, non-negative, non-decreasing, satisfying $\sum_{\beta} f_{\beta}U_{\beta}(0) < \infty$ and $\sum_{\beta} U_{\beta}(y) = 1 - y$.

This system is solved by a standard method of characteristic curves, with explicit formula containing inverse functions such as t_0 of the characteristic curve y_C . Hence the idea of hydrodynamic limit is of relevance to the results. (Incidentally, the method of characteristic curves gives time local solutions, while the assumptions on initial data satisfies no-shock wave condition, implying time global solution.)

4. Search cost. There is a large number of studies concerning move-to-front rule in the context of data theory in computer sciences. The LRU (least-recently-used) caching as a data allocation algorithm in computer memory or web page browsing is equivalent to move-to-front rule, with a data request corresponding to a particle jumping to the top position. The search cost C_N defined as the position of the first requested data just before the request is of interest. See [4] for references. We have $\frac{1}{N}C_N = Y_{Q^{(N)}(t)}^{(N)}(t)$, where $Q^{(N)}(t)$ is the label of the particle which jumped first after time t, to which we can apply **Theorem 2** to obtain, for example, $\lim_{N\to\infty} \Pr_t[\frac{1}{N}C_N(t) > x] = \frac{\int \int w \mu_{y,t}(dw) dy}{\int w \lambda(dw)}$.

Stationary distribution $\mathbf{E}_{\infty}[\cdot]$ for $N < \infty$ has naturally been studied since the model's first appearance in the literature [5]. By having the initial configuration $y^{(N)}$ distribute under $\mathbf{E}_{\infty}[\cdot]$, we can handle the stationary distribution $\mu_{\infty}^{(N)} = \mathbf{E}_{\infty}[\mu_{0}^{(N)}]$ for the joint jump rate and position distribution in our framework. **Theorem 2** then implies, for example, $\lim_{N\to\infty} \mathbf{P}_{\infty}[\frac{1}{N}C_{N} > x] = \frac{\int w e^{-wt_{0}(x)}\lambda(dw)}{\int w\lambda(dw)}$.

References.

- [1] K. Hattori, T. Hattori, Stochastic Processes & Applications 119 (2009) 966–979.
- [2] K. Hattori, T. Hattori, Funkcialaj Ekvacioj (2009), to appear.
- [3] K. Hattori, T. Hattori, preprint (2008).
- [4] K. Hattori, T. Hattori, preprint (2009).
- [1-4] available at http://web.econ.keio.ac.jp/staff/hattori/liamazn.htm.
- [5] M. L. Tsetlin, Russian Math. Surv. **18**(4) (1963) 1–27.