

# Scaling Limits for Large Stochastic Networks

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# Results.

- *Rate Control under Heavy Traffic with Strategic Servers* (Bayraktar, B. and Cohen (2016)).
  - $N$ -player game for single server queues.
  - Each server has a cost function it seeks to minimize.
  - **Objective:** Compute (near) Nash equilibria.
  - **Asymptotic Model:** Mean Field Game for reflected diffusions.

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  - **Objective:** Compute (near) Nash equilibria.
  - **Asymptotic Model:** Mean Field Game for reflected diffusions.
- *Controlled Weakly Interacting Large Finite State Systems with Simultaneous Jumps* (B. and Friedlander (2016)).
  - Rate Control for large finite state jump Markov processes.
  - Central Controller.
  - **Objective:** Optimize system performance.
  - **Asymptotic Model:** Drift Control for Degenerate Time - Inhomogeneous Diffusions.

# Results.

- *Coding and Load Balancing Mechanisms in Cloud Storage Systems* (B. and Friedlander (2017)).
  - Large number of file stored “in pieces” over a large number of servers.
  - Each file stored in equally sized pieces across  $L$  servers s.t. any  $k$  pieces recover the full file.
  - **Objective:** Model Simplification (LLN and CLT for fluctuations).
  - **Asymptotic Model:** SDE in  $\ell_2$  driven by cylindrical Brownian motion.
- *Power of  $d$  Schemes on Erdős-Rényi Graphs* (B., Mukherjee and Wu (2017)).
  - Each server has an associated queue in infinite capacity buffer.
  - An Erdős-Rényi graph (possibly time varying) describes the neighborhood of any server.
  - An arriving job chooses a server at random which then queries  $d - 1$  neighbors at random and sends the job to shortest queue.
  - **Objective:** LLN (Annealed and Quenched).
  - **Asymptotic Model:** Same infinite system of ODE as the ‘fully connected’ system. ( $np_n \rightarrow \infty$ )

# Rate Control with Strategic Servers.

- Sequence of  $d$  single server queues. (arrival rate  $\lambda^n$ , service rate  $\mu^n$ )
- Critically Loaded:  $n^{-1/2}(\lambda^n - \mu^n) \rightarrow c$ .
- Limit (under usual scaling) given by a reflected BM.
- I.e. if  $Q_i^n(t)$  is queue length of  $i$ -th queue then  $\tilde{Q}_i^n \doteq Q_i^n/\sqrt{n}$  converges to a BM with drift  $c$ , reflected at 0.
- Here consider a setting where each server can exercise control of arrival/service rates. [Arise in service networks, cloud computing, limit order books...]

# Rate Control with Strategic Servers.

- control can depend on 'everything' up to current time.
- ...also rates can depend on queue state and the empirical measure.
- Each server aims to minimize its individual cost.
- Interested in (near) Nash equilibria.
- For large  $d$  (even with diffusion approximations) computing Nash equilibria is computationally intractable.
- **Approach:** Heavy traffic + large  $d$  asymptotic regime.

# Problem Setting.

- Fix  $T$  (time horizon) and  $L$  (buffer size). **Control set:**  $U$  a compact set.

- **Controlled rates:**

$$\lambda^{N,i}(t) = aN + \lambda(t, \tilde{v}^N(t), \tilde{Q}_i^N(t), \alpha_i^N(t))\sqrt{N} + o(\sqrt{N}),$$

$$\mu^{N,i}(t) = aN + \mu(t, \tilde{v}^N(t), \tilde{Q}_i^N(t), \alpha_i^N(t))\sqrt{N} + o(\sqrt{N}).$$

- $\tilde{Q}^N = Q^N/\sqrt{N}$ ,  $\tilde{v}^N(t) = \frac{1}{N} \sum_{i=1}^N \delta_{\tilde{Q}_i^N(t)}$ .

- $V^{N,i}, Z^{N,i}$  unit rate independent Poisson processes.

- **State equation:**

$$Q^{N,i}(t) = Q^{N,i}(0) + V^{N,i} \left( \int_0^t 1_{\{\tilde{Q}^{N,i}(s) < L\}} \lambda^{N,i}(s) ds \right) - Z^{N,i} \left( \int_0^t 1_{\{\tilde{Q}^{N,i}(s) > 0\}} \mu^{N,i}(s) ds \right).$$

# Problem Setting.

- Skorohod map:

For  $\psi \in D([0, T] : \mathbb{R})$  with  $\psi(0) \in [0, L]$ , say  $(\varphi, \zeta_1, \zeta_2) \in \mathcal{D}([0, T] : \mathbb{R}^3)$  solves the Skorohod problem for  $\psi$  if:

- For every  $t \in [0, T]$ ,  $\varphi(t) = \psi(t) + \zeta_1(t) - \zeta_2(t) \in [0, L]$ .
- $\zeta_i$  are nonnegative and nondecreasing,  $\zeta_1(0) = \zeta_2(0) = 0$ , and

$$\int_{[0, T]} \mathbf{1}_{(0, L]}(\varphi(s)) d\zeta_1(s) = \int_{[0, T]} \mathbf{1}_{[0, L)}(\varphi(s)) d\zeta_2(s) = 0.$$

Write  $\Gamma(\psi) = (\varphi, \zeta_1, \zeta_2)$  and refer to  $\Gamma$  as the Skorohod map.



# Problem Setting.

- State evolution using the Skorokhod map:

$$\begin{aligned} & (\tilde{Q}_i^N, \tilde{Y}_i^N, \tilde{R}_i^N)(t) \\ &= \Gamma \left( \tilde{Q}_i^N(0) + \int_0^t \tilde{b}_i^N(s) ds + \tilde{A}_i^N(\cdot) - \tilde{D}_i^N(\cdot) + o(1) \right) (t), \quad t \in [0, T]. \end{aligned}$$

- $\tilde{b}_i^N(t) \doteq b(t, \tilde{v}^N(t), \tilde{Q}_i^N(t), \alpha_i^N(t))$ ,  $b \doteq \lambda - \mu$ ,

$$\tilde{Y}_i^N(t) \doteq \frac{1}{\sqrt{N}} \int_0^t 1_{\{\tilde{Q}_i^N(s)=0\}} \mu_i^N(s) ds, \quad \tilde{R}_i^N(t) \doteq \frac{1}{\sqrt{N}} \int_0^t 1_{\{\tilde{Q}_i^N(s)=L\}} \lambda_i^N(s) ds.$$

$$\langle \tilde{A}_i^N, \tilde{A}_j^N \rangle(t) = \delta_{ij} \frac{1}{N} \int_0^t 1_{\{\tilde{Q}_i^N(s) < L\}} \lambda_i^N(s) ds, \quad \langle \tilde{D}_i^N, \tilde{D}_j^N \rangle(t) = \dots$$

# Control Problem.

- $\mathcal{U}^N$  is the class of all admissible controls  $\alpha^N = (\alpha^{N,1}, \dots, \alpha^{N,N})$ .
- Cost for initial condition  $\tilde{Q}^N(0)$  and control  $\alpha^N$ :

$$J^{N,i}(\tilde{Q}^N(0); \alpha^N) \doteq E \left[ \int_0^T f(t, \tilde{v}^N(t), \tilde{Q}^{N,i}(t), \alpha^{N,i}(t)) dt + g(\tilde{v}^N(T), \tilde{Q}^{N,i}(T)) \right. \\ \left. - \int_0^T y(t, \tilde{v}^N(t)) d\tilde{Y}^{N,i}(t) + \int_0^T r(t, \tilde{v}^N(t)) d\tilde{R}^{N,i}(t) \right].$$

- **Asymptotic Nash Equilibrium:** Sequence of admissible controls  $\{\tilde{\alpha}^{N,i} : 1 \leq i \leq N\}_{N \in \mathbb{N}}$  is an **asymptotic Nash equilibrium** if for every  $j$ , and every sequence of admissible controls  $\{\beta^N\}_{N=1}^\infty$  for the  $j$ -th player,

$$\limsup_{N \rightarrow \infty} J^{N,j}(\tilde{Q}^N(0); \tilde{\alpha}^{N,1}, \dots, \tilde{\alpha}^{N,N}) \\ \leq \liminf_{N \rightarrow \infty} J^{N,j}(\tilde{Q}^N(0); \tilde{\alpha}^{N,1}, \dots, \tilde{\alpha}^{N,j-1}, \beta^N, \tilde{\alpha}^{N,j+1}, \dots, \tilde{\alpha}^{N,N}).$$

# Mean Field Game.

- (Lasry and Lions (2006), Huang, Malhamé and Caines (2006), Carmona and Delarue (2013), Carmona and Lacker (2015)...).
- ... a fixed point problem on  $\mathcal{P}_{T,L} = \mathcal{P}(C([0, T] : [0, L]))$ .
- For fixed  $x \in [0, L]$  and  $\nu \in \mathcal{P}_{T,L}$  consider a stochastic control problem:
- Filtered probability space:  $\Xi = (\Omega, \mathcal{F}, \{\mathcal{F}_t\}, P, B)$ .
- An admissible pair on  $\Xi$ : Stochastic processes  $(\alpha, Z)$ , such that
  - $\alpha = \{\alpha(s)\}_{0 \leq s \leq T}$  is a  $U$ -valued  $\mathcal{F}_s$ -progressively measurable process,
  - $Z = \{Z(s)\}_{0 \leq s \leq T}$  is a  $[0, L] \times \mathbb{R}_+ \times \mathbb{R}_+$  valued  $\mathcal{F}_s$ -adapted continuous process.such that  $(\alpha, Z)$  satisfy...

# Mean Field Game.

$$Z(t) = (X, Y, R)(t) = \Gamma \left( x + \int_0^t \bar{b}(s) ds + \sigma B(\cdot) \right) (t), \quad t \in [0, T],$$

where

$$\bar{b}(s) \doteq b(s, \nu(s), X(s), \alpha(s)), \quad s \in [0, T],$$

and  $\nu(s)$  is the marginal of  $\nu$  at time instant  $s$  and  $\sigma = \sqrt{2a}$ .

- Denote by  $\mathcal{A}(\Xi, x, \nu)$  the collection of all admissible pairs  $(\alpha, Z)$ .

# Mean Field Game.

- **Cost function in the MFG.** Given  $\nu \in \mathcal{P}_{T,L}$ ,  $x \in [0, L]$  and a system  $\Xi$ , let  $(\alpha, Z) \in \mathcal{A}(\Xi, x, \nu)$ . Define

$$J_\nu(x, \alpha, Z) \doteq E \left[ \int_0^T f(s, \nu(s), X(s), \alpha(s)) ds + g(\nu(T), X(T)) - \int_0^T y(s, \nu(s)) dY_s + \int_0^T r(s, \nu(s)) dR_s \right].$$

- **Value function:**

$$V_\nu(x) = \inf_{\Xi} \inf_{(\alpha, Z) \in \mathcal{A}(\Xi, x, \nu)} J_\nu(x, \alpha, Z).$$

- Denote by  $V_\nu(t, x)$  the value function for the control problem over  $[t, T]$ .

# Mean Field Game.

- A **solution to the MFG** with initial condition  $x \in [0, L]$  is defined to be a  $\nu \in \mathcal{P}_{T,L}$  such that there exist a system  $\Xi$  and an  $(\alpha, Z) \in \mathcal{A}(\Xi, x, \nu)$  such that  $Z = (X, Y, R)$  satisfies
  - $V_\nu(0, x) = J_\nu(0, x, \alpha, Z)$ .
  - $P \circ X^{-1} = \nu$
- If there exists a unique such  $\nu$ , we refer to  $V_\nu(0, x)$  as the **value of the MFG** with initial condition  $x$ .

# Solving the Mean Field Game.

- **Condition A:** Functions  $b, f, g, y, r$  are Lipschitz. For every  $(t, \eta, x, p) \in [0, T] \times \mathcal{P}([0, L]) \times [0, L] \times \mathbb{R}$ , there is a **unique**  $\hat{\alpha}(t, \eta, x, p) \in U$  such that

$$\hat{\alpha}(t, \eta, x, p) = \arg \min_{u \in U} h(t, \eta, x, u, p).$$

$$h(t, \eta, x, u, p) = f(t, \eta, x, u) + b(t, \eta, x, u)p.$$

- For  $c \in (0, \infty)$ , let  $\mathcal{M}_c$  be the collection of all  $\nu \in \mathcal{P}_{T,L}$  such that

$$\sup_{0 \leq s < t \leq T} \frac{W_1(\nu(t), \nu(s))}{(t-s)^{1/2}} \leq c$$

and let

$$\mathcal{M}_0 = \cup_{c>0} \mathcal{M}_c.$$

# Solving the Mean Field Game.

- Under **Condition A**, for  $\nu \in \mathcal{M}_0$ ,  $V_\nu$  is the unique  $H_{2+\frac{1}{2}}$  solution of:

$$-D_t\phi - H(t, \nu(t), x, D\phi) - \frac{1}{2}\sigma^2 D^2\phi = 0, \quad (t, x) \in [0, T] \times [0, L],$$

with **BC**:  $\phi(T, x) = g(\nu(T), x)$ ,

$$D\phi(t, 0) = y(t, \nu(t)), \text{ and } D\phi(t, L) = r(t, \nu(t)), \quad t \in [0, T],$$

where  $H(t, \eta, x, p) = \inf_{u \in U} h(t, \eta, x, u, p)$ .

- $\alpha(u, \omega) \doteq \hat{\alpha}(u, \nu(u), X(u, \omega), DV_\nu(u, X(u, \omega)))$  is the (essentially unique) optimal feedback control.



# Solving the Mean Field Game.

- Fix  $\nu$  and denote the state process under the optimal feedback control for the associated stochastic control problem by  $X^\nu$ .
- Define  $\Phi$  on  $\mathcal{M}_0$  as  $\Phi(\nu) \doteq P \circ (X^\nu)^{-1}$ .
- **Theorem.** Under **Condition A**  $\Phi$  has a fixed point  $\bar{\nu}$ . This fixed point is a solution of the MFG.
  - Proof idea: Schauder's fixed point theorem.
  - Main technical step: Proving the continuity of  $\Phi$ .
- Under an additional monotonicity condition the solution is unique.

# Asymptotic Nash Equilibrium.

- Given a solution  $\bar{v}$  of MFG, define 'feedback controls' for the  $N$ -player game as

$$\tilde{\alpha}^{N,i}(t) \doteq \hat{\alpha}(t, \bar{v}(t), \tilde{Q}^{N,i}(t), DV_{\bar{v}}(t, \tilde{Q}^{N,i}(t))).$$

- Condition B.** The drift  $b(t, \eta, x, \alpha) \equiv b(t, x, \alpha)$ . Initial conditions are exchangeable and  $\tilde{Q}^{N,i}(0) \rightarrow x$  for some  $x \in [0, L]$ .
- For a control  $\beta^N$ , let

$$\tilde{\alpha}_{-1}^N(t) \doteq (\beta^N, \tilde{\alpha}^{N,2}(t), \dots, \tilde{\alpha}^{N,N}(t)).$$

- Theorem.** Under Conditions A and B,  $\tilde{\alpha}^N$  is an asymptotic Nash equilibrium:

$$\limsup_{N \rightarrow \infty} J^{N,1}(\tilde{Q}^N(0), \tilde{\alpha}^N) = V_{\bar{v}}(0, x) \leq \liminf_{N \rightarrow \infty} J^{N,1}(\tilde{Q}^N(0), \tilde{\alpha}_{-1}^N).$$

# Proof Sketch.

- **Step 1** As  $N \rightarrow \infty$ ,  $\tilde{v}_{-1}^N \rightarrow \bar{v}$ , where  $\tilde{v}_{-1}^N = \frac{1}{N-1} \sum_{i=2}^{N-1} \delta_{\tilde{Q}^{N,i}}$ .

- **Step 2** When players uses  $\tilde{\alpha}^N$ ,  $(\tilde{Q}^{N,i}, \tilde{Y}^{N,i}, \tilde{R}^{N,i})$  converge to

$$(\tilde{Q}^i, \tilde{Y}^i, \tilde{R}^i) = \Gamma(x + \int_0^\cdot b(s, \tilde{Q}^i(s), \hat{\alpha}(s, \bar{v}(s), \tilde{Q}^i(s), DV_{\bar{v}}(s, \tilde{Q}^i(s)))) ds + \sigma B^i).$$

The costs converge as well.

- This proves:  $\limsup_{N \rightarrow \infty} J^{N,1}(\tilde{Q}^N(0), \tilde{\alpha}^N) = V_{\bar{v}}(0, x)$ .
- **Step 3** When for every  $i \neq 1$ , the  $i$ -th player uses  $\tilde{\alpha}^{N,i}$  the limit points of the cost (for the first player) are costs for the (relaxed version of) stochastic control problem for  $\bar{v}$  and so are bounded below by  $V_{\bar{v}}(0, x)$ .
- This proves:  $V_{\bar{v}}(0, x) \leq \liminf_{N \rightarrow \infty} J^{N,1}(\tilde{Q}^N(0), \tilde{\alpha}_{-1}^N)$ .

## Comments on Step 1: $\tilde{\nu}_{-1}^N \rightarrow \bar{\nu}$ .

- Follows Kotelenetz and Kurtz(2010).
- Define  $\tilde{G}^{N,i} \doteq (\tilde{Q}^{N,i}, \tilde{Y}^{N,i}, \tilde{R}^{N,i})$  and

$$\Xi^N \doteq \frac{1}{N-1} \sum_{i=2}^{N-1} \delta_{\tilde{G}^{N,i}}.$$

- $\{\tilde{G}^{N,i}\}_{N \in \mathbb{N}}$  is  $\mathcal{C}$ -tight for each  $i$ .
- If  $\tilde{G}^N = (\tilde{G}^{N,i})_{i=2}^{\infty}$  converges along a subsequence to  $\tilde{G}$ , then

$$(\tilde{G}^N, \Xi^N) \rightarrow (\tilde{G}, \Xi), \quad \Xi = \lim_{m \rightarrow \infty} \frac{1}{m} \sum_{i=1}^m \delta_{\tilde{G}^i}.$$

## Comments on Step 1: $\tilde{\nu}_{-1}^N \rightarrow \bar{\nu}$ .

- **Characterization.** For  $i \geq 2$ , with  $\tilde{G}^i = (\tilde{Q}^i, \tilde{Y}^i, \tilde{R}^i)$ ,

$$(\tilde{Q}^i, \tilde{Y}^i, \tilde{R}^i) = \Gamma(x + \int_0^\cdot b(s, \tilde{Q}^i(s), \hat{\alpha}(s, \bar{\nu}(s), \tilde{Q}^i(s), DV_{\bar{\nu}}(s, \tilde{Q}^i(s)))) ds + \sigma B^i).$$

- This shows that  $\tilde{Q}^i$  are i.i.d.  $\bar{\nu}$ . Thus

$$\lim_{N \rightarrow \infty} \tilde{\nu}_{-1}^N = \lim_{N \rightarrow \infty} \Xi_{(1)}^N = \Xi_{(1)} = \lim_{m \rightarrow \infty} \frac{1}{m} \sum_{i=1}^m \delta_{\tilde{Q}^i} = \bar{\nu}.$$

# Numerical Approximations.

- Construction of an asymptotic Nash equilibrium requires the solution of the MFG.
- Tractable expressions for the MFG solution are not available in general.
- Suitable numerical approximations needed.
- We introduce a Markov chain approximation method to construct numerical solutions of the MFG.

# Outline of the Scheme.

- Discretize state space: For  $h > 0$ ,  $\mathbb{S}^h \doteq \{-h, 0, h, \dots, L + h\}$ .
- Introduce a sequence of controlled Markov chains  $\{X_n^k\}_{n \in \mathbb{N}_0}$ , one for each value of the discretization parameter  $h^k > 0$ , where  $h^k \rightarrow 0$  as  $k \rightarrow \infty$ .
- **Theorem.** Suppose Condition A holds and that the MFG has a unique solution. Then the law  $\nu^k \in \mathcal{P}_{T,L}$  of the continuous time interpolation of  $X^k$  converges to  $\bar{\nu}$  for small  $T$ .

# Construction of $X_n^k$ .

- The  $k$ -th chain  $X_n^k$  is obtained by solving an **approximate fixed point problem**.
- Fix  $\nu \in \mathcal{P}_{T,L}$ . Formulate a MDP with transition kernel and cost depending on  $\nu$ . The cost is the discretized analog of the cost in the MFG.
- Denote the law of the optimal state process (with piecewise linear interpolation) as  $\Phi_k(\nu)$ .
- **Approximate Contraction.** For some  $q \in (0, 1)$

$$W_1^2(\Phi_k(\nu), \Phi_k(\nu')) \leq q(h_k^2 + W_1^2(\nu, \nu')).$$



# Construction of $X_n^k$ .

- Let  $\nu^m \doteq (\Phi_k)^m(\nu)$ . Then

$$W_1^2(\Phi_k(\nu^m), \nu^m) \leq \frac{q}{1-q} h_k^2 + q^{m-1} W_1^2(\nu^2, \nu^1).$$

- Let

$$m(k) \doteq \min \left\{ m : q^{m-1} W_1^2(\nu^2, \nu^1) \leq \frac{q}{1-q} h_k^2 \right\}.$$

- Then  $\bar{\nu}^k \doteq (\Phi_k)^{m(k)}(\nu)$  converges to  $\bar{\nu}$  as  $k \rightarrow \infty$ .