## Ph.D. course on Network Dynamics Homework 5

## To be discussed on Tuesday, November 5, 2013

**Recap** In class, we have considered network dynamics of the following form. Given a graph  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$ , a finite state space  $\mathcal{X}$ , and local transition kernels  $\Psi_v(x'|x_v, x_{\mathcal{N}_v})$  for  $v \in \mathcal{V}$ , we consider a continuous-time Markov chain  $X(t) = \{X_v(t) : v \in \mathcal{V}\}$  on  $\mathcal{X}^{\mathcal{V}}$ , whereby every  $v \in \mathcal{V}$  gets activated at the ticking of an independent rate-1 Poisson clock, and, if activated at some time  $t \geq 0$ , it updates its state to  $X_v(t^+) = j \in \mathcal{X}$  with conditional probability  $\Psi_v(j|X_v(t), X_{\mathcal{N}_v}(t))$ .

Let  $\mathcal{P}(\mathcal{X})$  be the simplex of probability vectors over  $\mathcal{X}$ . For a length-k vector x with components in  $\mathcal{X}$ , define the type of x as

$$\theta(x) \in \mathcal{P}(\mathcal{X}), \qquad \theta_i(x) = \frac{1}{k} \left| \{l = 1, \dots, k : x_l = i\} \right|.$$

We have argued that, if the population is totally mixed (i.e.,  $\mathcal{G}$  is complete with n nodes), and if the transition kernels are homogeneous, i.e., if

$$\Psi_v\left(j|i, x_{\mathcal{N}_v}\right) = P_{ij}(\theta(x_{\{v\}\cup\mathcal{N}_v})), \qquad \forall v \in \mathcal{V}$$

for some Lipschitz-continuous functions  $P_{ij} : \mathcal{P}(\mathcal{X}) \to \mathbb{R}$  (adding or not v to  $\mathcal{N}_v$  is rather arbitrary, and irrelevant as n grows large), then one can give up on keeping track of the states of the single nodes and rather take a 'Eulerian' viewpoint, looking at the evolution of the empirical density of the agents' opinions, defined as

$$\rho^n(t) := \theta(X(t))$$

In fact, under such assumptions of total mixing of the population and homogeneity of agents' behavior,  $\rho^n(t)$  is a Markov chain itself on the space of types  $\mathcal{P}_n(\mathcal{X}) := \{ \rho \in \mathbb{R}^{\mathcal{X}} : n\rho_i = n_i \in \mathbb{Z}_+, \sum_i n_i = n \}.$  Moreover, Kurtz's theorem guarantees that there exist positive constants K, K' such that, for all  $T > 0, \varepsilon > 0$ , if  $\lim_{n} \rho^{n}(0) = \rho(0) \in \mathcal{P}(\mathcal{X})$ , then

$$\mathbb{P}\left(\sup_{t\in[0,T]}||\rho^n(t)-\rho(t)||\geq\varepsilon\right)\leq K'\exp(-Kn\varepsilon^2/T)\,,$$

where  $\rho(t)$  is the solution of the Cauchy problem associated to the mean-field ODE

$$\frac{\mathrm{d}}{\mathrm{d}t}\rho = P'(\rho)\rho - \rho\,,$$

with initial condition  $\rho(0)$ .

Exercise 1 (Cardinality of the space of types). Prove that

$$|\mathcal{P}_n(\mathcal{X})| = \binom{n+|\mathcal{X}|-1}{|\mathcal{X}|-1} \le (n+|\mathcal{X}|-1)^{|\mathcal{X}|-1}$$

(Hint: one has to count the different ways of assigning n identical balls to  $|\mathcal{X}|$  distinguished urns... )

**Exercise 2** (mean-field limit of the noisy majority-rule dynamics). The noisy majority-rule dynamics is characterized by state space  $\mathcal{X} = \{0, 1\}$  and transition kernel  $\Psi_v(1|i, x_{\mathcal{N}_v}) = \Phi_\beta(d_v^{-1} \sum_{w \in \mathcal{N}_v} x_w), i = 0, 1, where$ 

$$\Phi_{\beta}(\rho) := \frac{\exp(\beta\rho)}{\exp(\beta\rho) + \exp(\beta(1-\rho))}, \qquad \beta \ge 0.$$

(a) Prove that,

$$\lim_{\beta \to \infty} \Phi_{\beta}(\rho) = \begin{cases} 0 & \text{if} \quad \rho \in [0, 1/2) \\ 1/2 & \text{if} \quad \rho = 1/2 \\ 1 & \text{if} \quad \rho \in (1/2, 1] \end{cases}$$

(this justifies the name 'noisy majority rule', with  $1/\beta$  to be interpreted as a measure of the noise level)

(b) Prove that the mean-field limit ODE is given by

$$\frac{\mathrm{d}}{\mathrm{d}t}\rho_1 = \Phi_\beta(\rho_1) - \rho_1; \qquad (1)$$

(c) For arbitrary  $\beta \ge 0$ , find the equilibria of (1), discuss their stability and characterize their region of attraction.

**Exercise 3** (k-majority rule dynamics). Let k be a positive odd integer and  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$  a graph such that  $d_v \geq k$  for all  $v \in \mathcal{V}$ . Consider the k-majority rule dynamics whereby each agent, when activated, selects k distinct neighbors uniformly at random and moves towards the opinion held by the majority of them. Find the mean-field ODE and study the asymptotic behavior (as t grows large) of the associated initial value problem.

**Exercise 4** (mean-field limit of the SIS epidemics). Let  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$  be a connected graph. Consider the SIS epidemics with rate- $(\gamma/(n-1))$  transmission time on every link with one infected and one susceptible end-node, and rate- $(1 - \gamma)$  exponential recovery time, where  $\gamma \in (0, 1)$ . Let  $\mathcal{X} = \{S, I\}$ .

- (a) Write down the transition kernel  $\Psi_v(j|i, x_{\mathcal{N}_v})$  for  $v \in \mathcal{V}$  (hint: assume that a node gets activated whenever it has a potential recovery or gets a potential decease transmission from one of the other n-1 nodes)
- (b) What are all the absorbing states in  $\mathcal{X}^{\mathcal{V}}$  of the Markov chain X(t)?
- (c) Prove that the mean-field limit ODE is

$$\begin{cases} \frac{\mathrm{d}}{\mathrm{d}t}\rho_S = (1-2\gamma)\rho_I + \gamma\rho_I^2 \\ \frac{\mathrm{d}}{\mathrm{d}t}\rho_I = -(1-2\gamma)\rho_I - \gamma\rho_I^2 \end{cases}$$
(2)

(d) What are the equilibria of (2)? Discuss their stability as γ varies in (0,1).

**Exercise 5** (mean-field limit of the SIR epidemics). Let  $\mathcal{G} = (\mathcal{V}, \mathcal{E})$  be a connected graph and  $\gamma \in (0, 1)$ . Consider the SIR epidemics with rate- $\frac{\gamma}{n-1}$  transmission time on every link with one infected and one susceptible end-node, and rate- $(1 - \gamma)$  exponential recovery time. Let  $\mathcal{X} := \{S, I, R\}$ .

- (a) Write down the transition kernel  $\Psi_v(j|i, x_{\mathcal{N}_v})$  for  $v \in \mathcal{V}$  (hint: assume that a node gets activated whenever it has a potential recovery or gets a potential decease transmission from one of the other n-1 nodes)
- (b) What are all the absorbing states in  $\mathcal{X}^{\mathcal{V}}$  of the Markov chain X(t)? And what are they if  $\mathcal{G}$  is the complete graph?

(c) Prove that the mean-field limit ODE is

$$\begin{cases} \frac{\mathrm{d}}{\mathrm{d}t}\rho_{S} = -\gamma\rho_{S}\rho_{I} \\ \frac{\mathrm{d}}{\mathrm{d}t}\rho_{I} = \gamma\rho_{S}\rho_{I} - (1-\gamma)\rho_{I} \\ \frac{\mathrm{d}}{\mathrm{d}t}\rho_{R} = (1-\gamma)\rho_{I} \end{cases}$$
(3)

- (d) Plot (or sketch) the phase portrait of (3);
- (e) What are the equilibria of (3)?
- (f) Prove that every trajectory of (3) is convergent to some limit  $\rho^* \in \mathcal{P}(\mathcal{X})$ , which depends on the initial condition  $\rho(0)$ ; (hint:  $\rho_S(t)$  is non-increasing,  $\rho_R(t)$  non-decreasing, and the trajectoris belong to the compact  $\mathcal{P}(\mathcal{X})$ )
- (g) Let  $\mathcal{R} := \{ \rho \in \mathcal{P}(\mathcal{X}) : \rho_I = 0 \}$ ,  $\mathcal{R}^S_{\gamma} := \{ \rho \in \mathcal{R} : \rho_S \leq 1/\gamma 1 \}$ , and  $\mathcal{R}^U_{\gamma} := \mathcal{R} \setminus \mathcal{R}^S_{\gamma}$ . Prove that, for every initial condition  $\rho(0) \in \mathcal{P}(\mathcal{X}) \setminus \mathcal{R}^U_{\gamma}$ , the solution of (3) satisfies

$$\lim_{t \to +\infty} \rho(t) \in \mathcal{R}^S_\gamma;$$

(h) Conclude that, for  $\gamma > 1/2$ ,

$$\rho_I(0) > 0 \implies \lim_{t \to \infty} \rho_R(t) \ge 2 - 1/\gamma > 0.$$