# Supplement to "Liquidity based modeling of asset price bubbles via random matching"

Francesca Biagini \* Andrea Mazzon\* Thilo Meyer-Brandis\* Katharina Oberpriller<sup>†</sup>

November 2, 2022

This is a supplement to the paper [1]. The supplement is organized as follows. First, we prove Theorem 3.13 in [1] which provides the existence of the dynamical system  $\mathbb{D}$  introduced in Definition 3.6 in [1]. Second, we show some properties of  $\mathbb{D}$  which are summarized in Theorem 3.14 in [1].

In the following, we only state the basic setting and refer to [1] for definitions.

### 1 Setting

Let  $(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{P})$  be a probability space and  $(\hat{\Omega}, \hat{\mathcal{F}})$  another measurable space. We define the product space

$$(\Omega, \mathcal{F}) := (\tilde{\Omega} \times \hat{\Omega}, \tilde{\mathcal{F}} \otimes \hat{\mathcal{F}}).$$
(1.1)

Let  $\hat{P}$  be a Markov kernel (or stochastic kernel) from  $\tilde{\Omega}$  to  $\hat{\Omega}$ . Given  $\tilde{\omega} \in \tilde{\Omega}$ , we set  $\hat{P}^{\tilde{\omega}} := \hat{P}(\tilde{\omega})$  with a slight notational abuse. We then introduce a probability measure P on  $(\Omega, \mathcal{F})$  as the semidirect product of  $\tilde{P}$  and  $\hat{P}$ , that is,

$$P(\tilde{A} \times \hat{A}) := (\tilde{P} \ltimes \hat{P})(\tilde{A} \times \hat{A}) = \int_{\tilde{A}} \hat{P}^{\tilde{\omega}}(\hat{A}) d\tilde{P}(\tilde{\omega}).$$
(1.2)

We fix an atomless probability space  $(I, \mathcal{I}, \lambda)$  representing the space of agents and let  $(I \times \Omega, \mathcal{I} \boxtimes \mathcal{F}, \lambda \boxtimes P)$ be a rich Fubini extension of  $(I \times \Omega, \mathcal{I} \otimes \mathcal{F}, \lambda \otimes P)$ . All agents in I can be classified according to their type. In particular, we let  $S = \{1, 2, ..., K\}$  be a finite space of types and say that an agent has type J if he is not matched. We denote by  $\hat{S} := S \times (S \cup \{J\})$  the extended type space. Moreover, we call  $\hat{\Delta}$  the space of extended type distributions, which is the set of probability distributions p on  $\hat{S}$  satisfying p(k,l) = p(l,k) for any k and l in S. This space is endowed with the topology  $\mathcal{T}^{\Delta}$  induced by the topology of the space of matrices with |S| rows and |S| + 1 columns. We consider  $(n)_{n\geq 1}$  time periods and denote by  $(\eta^n, \theta^n, \xi^n, \sigma^n, \varsigma^n)$  the matrix valued processes, with  $(\eta^n, \theta^n, \xi^n, \sigma^n, \varsigma^n) = (\eta^n_{kl}, \theta^n_{kl}, \xi^n_{kl}, \sigma^n_{kl}[r, s], \varsigma^n_{kl}[r])_{k,l,r,s\in S\times S\times S\times S}$  for  $n \geq 1$ , on  $(\Omega, \mathcal{F}, P)$ . For a detailed introduction of these processes we refer to Section 3 in [1]. Moreover, let  $\hat{p} = (\hat{p}^n)_{n\geq 1}$ be a stochastic process on  $(\Omega, \mathcal{F}, P)$  with values in  $\hat{\Delta}$ , representing the evolution of the underlying extended type distribution. We assume that  $\hat{p}^0$  is deterministic.

Given the input processes  $(\eta, \theta, \xi, \sigma, \varsigma)$  we denote by  $\mathbb{D}$  a dynamical system on  $(I \times \Omega, \mathcal{I} \boxtimes \mathcal{F}, \lambda \boxtimes P)$  and

<sup>\*</sup>Workgroup Financial Mathematics, Department of Mathematics, Ludwig-Maximilians-Universität München, Theresienstr. 39, 80333 Munich, Germany.

<sup>&</sup>lt;sup>†</sup>Department of Mathematics, University of Freiburg, Ernst-Zermelo-Str. 1, 79104 Freiburg im Breisgau, Germany.

by  $\Pi = (\alpha, \pi, g) = (\alpha^n, \pi^n, g^n)_{n \in \mathbb{N} \setminus \{0\}}$  the agent-type function, the random matching and the partner-type function, respectively, as introduced in Definition 3.6 in [1], which we recall in the following.

**Definition 1.1.** A dynamical system  $\mathbb{D}$  defined on  $(I \times \Omega, \mathcal{I} \boxtimes \mathcal{F}, \lambda \boxtimes P)$  is a triple  $\Pi = (\alpha, \pi, g) = (\alpha^n, \pi^n, g^n)_{n \in \mathbb{N} \setminus \{0\}}$  such that for each integer period  $n \ge 1$  we have:

- 1.  $\alpha^n : I \times \Omega \to S$  is the  $\mathcal{I} \boxtimes \mathcal{F}$ -measurable agent type function. The corresponding end-of-period type of agent *i* under the realization  $\omega \in \Omega$  is given by  $\alpha^n(i, \omega) \in S$ .
- 2. A random matching  $\pi^n : I \times \Omega \to I$ , describing the end-of-period agent  $\pi^n(i)$  to whom agent i is currently matched, if agent i is currently matched. If agent i is not matched, then  $\pi^n(i) = i$ . The associated  $\mathcal{I} \boxtimes \mathcal{F}$ -measurable partner-type function  $g^n : I \times \Omega \to S \cup \{J\}$  is given by

$$g^{n}(i,\omega) = \begin{cases} \alpha^{n}(\pi^{n}(i,\omega),\omega) & \text{if } \pi^{n}(i,\omega) \neq i \\ J & \text{if } \pi^{n}(i,\omega) = i \end{cases}$$

providing the type of the agent to whom agent i is matched, if agent i is matched, or J if agent i is not matched.

Let the initial condition  $\Pi^0 = (\alpha^0, \beta^0)$  of  $\mathbb{D}$  be given. We now construct a dynamical system  $\mathbb{D}$  defined on  $(I \times \Omega, \mathcal{I} \boxtimes \mathcal{F}, \lambda \boxtimes P)$  with input processes  $(\eta^n, \theta^n, \xi^n, \sigma^n, \varsigma^n)_{n \ge 1}$ . We assume that  $\Pi^{n-1} = (\alpha^{n-1}, \pi^{n-1}, g^{n-1})$  is given for some  $n \ge 1$ , and define  $\Pi^n = (\alpha^n, \pi^n, g^n)$  by characterizing the three sub-steps of random change of types of agents, random matchings, break-ups and possible type changes after matchings and break-ups as follows.

**Mutation:** For  $n \ge 1$  consider an  $\mathcal{I} \boxtimes \mathcal{F}$ -measurable post mutation function

$$\bar{\alpha}^n: I \times \Omega \to S.$$

In particular,  $\bar{\alpha}_i^n(\omega) := \bar{\alpha}^n(i,\omega)$  is the type of agent *i* after the random mutation under the scenario  $\omega \in \Omega$ . The type of the agent to whom an agent is matched is identified by a  $\mathcal{I} \boxtimes \mathcal{F}$ -measurable function

$$\bar{g}^n: I \times \Omega \to S \cup \{J\},\$$

given by

$$\bar{g}^n(i,\omega) = \bar{\alpha}^n(\pi^{n-1}(i,\omega),\omega)$$

for any  $\omega \in \Omega$ . In particular,  $\bar{g}_i^n(\omega) := \bar{g}^n(i,\omega)$  is the type of the agent to whom an agent is matched under the scenario  $\omega \in \Omega$ . Given  $\hat{p}^{n-1}$  and  $\tilde{\omega} \in \tilde{\Omega}$ , for any  $k_1, k_2, l_1$  and  $l_2$  in S, for any  $r \in S \cup \{J\}$ , for  $\lambda$ -almost every agent i, we set

$$\hat{P}^{\tilde{\omega}}\left(\bar{\alpha}_{i}^{n}(\tilde{\omega},\cdot)=k_{2},\bar{g}_{i}^{n}(\tilde{\omega},\cdot)=l_{2}|\alpha_{i}^{n-1}(\tilde{\omega},\cdot)=k_{1},g_{i}^{n-1}(\tilde{\omega},\cdot)=l_{1},\hat{p}^{n-1}(\tilde{\omega},\cdot)\right)(\hat{\omega})$$
$$=\eta_{k_{1},k_{2}}\left(\tilde{\omega},n,\hat{p}^{n-1}(\tilde{\omega},\hat{\omega})\right)\eta_{l_{1},l_{2}}\left(\tilde{\omega},n,\hat{p}^{n-1}(\tilde{\omega},\hat{\omega})\right),$$
(1.3)

$$\hat{P}^{\tilde{\omega}}\left(\bar{\alpha}_{i}^{n}(\tilde{\omega},\cdot)=k_{2},\bar{g}_{i}^{n}(\tilde{\omega},\cdot)=r|\alpha_{i}^{n-1}(\tilde{\omega},\cdot)=k_{1},g_{i}^{n-1}(\tilde{\omega},\cdot)=J,\hat{p}^{n-1}(\tilde{\omega},\cdot)\right)(\hat{\omega})$$
$$=\eta_{k_{1},k_{2}}\left(\tilde{\omega},n,\hat{p}^{n-1}(\tilde{\omega},\hat{\omega})\right)\delta_{J}(r),$$
(1.4)

We then set

$$\bar{\beta}^n(\omega) = (\bar{\alpha}^n(\omega), \bar{g}^n(\omega)), \quad n \ge 1.$$

The post-mutation extended type distribution realized in the state of the world  $\omega \in \Omega$  is denoted by  $\check{p}(\omega) = (\check{p}^n(\omega)[k,l])_{k \in S, l \in S \cup J}$ , where

$$\check{p}^{n}(\omega)[k,l] := \lambda(\{i \in I : \bar{\alpha}^{n}(i,\omega) = k, \bar{g}^{n}(i,\omega) = l\}).$$

$$(1.5)$$

**Matching:** We introduce a random matching  $\bar{\pi}^n : I \times \Omega \to I$  and the associated post-matching partner type function  $\bar{g}^n$  given by

$$\bar{\bar{g}}^{n}(i,\omega) = \begin{cases} \bar{\alpha}^{n}(\bar{\pi}^{n}(i,\omega),\omega) & \text{if } \bar{\pi}^{n}(i,\omega) \neq i \\ J & \text{if } \bar{\pi}^{n}(i,\omega) = i \end{cases}$$

satisfying the following properties:

- 1.  $\overline{\overline{g}}^n$  is  $\mathcal{I} \boxtimes \mathcal{F}$ -measurable.
- 2. For any  $\tilde{\omega} \in \tilde{\Omega}$ , any  $k, l \in S$  and any  $r \in S \cup \{J\}$ , it holds

$$\hat{P}^{\omega}(\bar{\bar{g}}^n(\tilde{\omega},\cdot)=r|\bar{\alpha}_i^n(\tilde{\omega},\cdot)=k,\bar{g}_i^n(\tilde{\omega},\cdot)=l)(\hat{\omega})=\delta_l(r).$$

This means that

$$\bar{\pi}^n_{\omega}(i) = \pi^{n-1}_{\omega}(i) \quad \text{ for any } i \in \{i : \pi^{n-1}(i,\omega) \neq i\}.$$

3. Given  $\tilde{\omega} \in \tilde{\Omega}$  and the post-mutation extended type distribution  $\check{p}^n$  in (1.5), an unmatched agent of type k is matched to a unmatched agent of type l with conditional probability  $\theta_{kl}(\tilde{\omega}, n, \check{p}^n)$ , that is for  $\lambda$ -almost every agent i and  $\hat{P}^{\tilde{\omega}}$ -almost every  $\hat{\omega}$ , we define

$$\hat{P}^{\bar{\omega}}(\bar{\bar{g}}^n(\tilde{\omega},\cdot) = l|\bar{\alpha}^n_i(\tilde{\omega},\cdot) = k, \bar{g}^n_i(\tilde{\omega},\cdot) = J, \check{p}^n(\tilde{\omega},\cdot))(\hat{\omega}) = \theta^n_{kl}(\tilde{\omega},\check{p}^n(\tilde{\omega},\hat{\omega})).$$
(1.6)

This also implies that

$$\hat{P}^{\tilde{\omega}}(\bar{g}^{n}(\tilde{\omega},\cdot)=J|\bar{\alpha}^{n}_{i}(\tilde{\omega},\cdot)=k, \bar{g}^{n}_{i}(\tilde{\omega},\cdot)=J, \check{p}^{n}(\tilde{\omega},\cdot))(\hat{\omega})=1-\sum_{l\in S}\theta^{n}_{kl}(\tilde{\omega},\check{p}^{n}(\tilde{\omega},\hat{\omega}))=b^{k}(\tilde{\omega},\check{p}^{n}(\tilde{\omega},\hat{\omega})).$$
(1.7)

The extended type of agent i after the random matching step is

$$\bar{\bar{\beta}}_i^n(\omega) = (\bar{\alpha}_i^n(\omega), \bar{\bar{g}}_i^n(\omega)), \quad n \ge 1$$

We denote the post-matching extended type distribution realized in  $\omega \in \Omega$  by  $\check{p}^n(\omega) = (\check{p}^n(\omega)[k, l])_{k \in S, l \in S \cup J}$ , where

$$\check{p}^{n}(\omega)[k,l] := \lambda(\{i \in I : \bar{\alpha}^{n}(i,\omega) = k, \bar{g}^{n}(i,\omega) = l\}).$$

$$(1.8)$$

Type changes of matched agents with break-up: We now define a random matching  $\pi^n$  by

$$\pi^{n}(i) = \begin{cases} \bar{\pi}^{n}(i) & \text{if } \bar{\pi}^{n}(i) \neq i \\ i & \text{if } \bar{\pi}^{n}(i) = i. \end{cases}$$
(1.9)

We then introduce an  $(\mathcal{I} \boxtimes \mathcal{F})$ -measurable agent type function  $\alpha^n$  and an  $(\mathcal{I} \boxtimes \mathcal{F})$ -measurable partner function  $g^n$  with

$$g^{n}(i,\omega) = \alpha^{n}(\pi^{n}(i,\omega),\omega), \quad n \ge 1,$$

for all  $(i, \omega) \in I \times \Omega$ . Given  $\tilde{\omega} \in \tilde{\Omega}$ ,  $\check{p}^n \in \hat{\Delta}$ , for any  $k_1, k_2, l_1, l_2 \in S$  and  $r \in S \cup \{J\}$ , for  $\lambda$ -almost every agent i, and for  $\hat{P}^{\tilde{\omega}}$ -almost every  $\hat{\omega}$ , we set

$$\hat{P}^{\tilde{\omega}}\left(\alpha_{i}^{n}(\tilde{\omega},\cdot)=l_{1},g_{i}^{n}(\tilde{\omega},\cdot)=r|\bar{\alpha}_{i}^{n}(\tilde{\omega},\cdot)=k_{1},\bar{g}_{i}^{n}(\tilde{\omega},\cdot)=J\right)(\hat{\omega})=\delta_{k_{1}}(l_{1})\delta_{J}(r),$$

$$\hat{P}^{\tilde{\omega}}\left(\alpha_{i}^{n}(\tilde{\omega},\cdot)=l_{1},g_{i}^{n}(\tilde{\omega},\cdot)=l_{2}|\bar{\alpha}_{i}^{n}(\tilde{\omega},\cdot)=k_{1},\bar{g}_{i}^{n}(\tilde{\omega},\cdot)=k_{2},\check{p}^{\tilde{n}}(\tilde{\omega},\cdot)\right)(\hat{\omega})$$
(1.10)

$$= \left(1 - \xi_{k_1 k_2}(\tilde{\omega}, n, \check{p}^n(\tilde{\omega}, \hat{\omega}))\right) \sigma_{k_1 k_2}[l_1, l_2](\tilde{\omega}, n, \check{p}^n(\tilde{\omega}, \hat{\omega})),$$
(1.11)

$$\hat{P}^{\tilde{\omega}}\left(\alpha_{i}^{n}(\tilde{\omega},\cdot)=l_{1},g_{i}^{n}(\tilde{\omega},\cdot)=J|\bar{\alpha}_{i}^{n}(\tilde{\omega},\cdot)=k_{1},\bar{g}_{i}^{n}(\tilde{\omega},\cdot)=k_{2},\check{p}^{n}(\tilde{\omega},\cdot)\right)(\hat{\omega})$$
$$=\xi_{k_{1}k_{2}}(\tilde{\omega},n,\check{p}^{n}(\tilde{\omega},\hat{\omega}))\varsigma_{k_{1}k_{2}}^{n}[l_{1}](\tilde{\omega},n,\check{p}^{n}(\tilde{\omega},\hat{\omega})).$$
(1.12)

The extended-type function at the end of the period is

$$\beta^{n}(\omega) = (\alpha^{n}(\omega), g^{n}(\omega)), \quad n \ge 1$$

We denote the extended type distribution at the end of period n realized in  $\omega \in \Omega$  by  $\hat{p}^n(\omega) = (\hat{p}^n(\omega)[k,l])_{k \in S, l \in S \cup J}$ , where

$$\hat{p}^{n}(\omega)[k,l] := \lambda(\{i \in I : \alpha^{n}(i,\omega) = k, g^{n}(i,\omega) = l\}).$$
(1.13)

Furthermore, the definition of Markov conditionally independent (MCI) dynamical system is provided in Definition 3.8 in [1]. We work under the following assumption, which is Assumption 3.9 in [1].

**Assumption 1.2.** Let  $(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{P})$  be the probability space introduced. We assume that there exists its corresponding hyperfinite internal probability space, which we denote from now on also by  $(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{P})$  by a slight notational abuse.

As already pointed out in [1], the proofs of the results below follow by analogous arguments as in [2] which is possible due to the product structure of the space  $\Omega$  in (1.1) and the Markov kernel P in (1.2). As in [2] we use some concepts and notations from nonstandard analysis. Note here that an object with an upper left star means the transfer of a standard object to the nonstandard universe. For a detailed overview of the necessary tools of nonstandard analysis, we refer to Appendix D.2. in [2].

# 2 Proof of Theorem 3.13 in [1]

From now on, we fix the hyperfinite internal space  $(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{P})$ , along with the input functions  $(\eta_{kl}, \theta_{kl}, \xi_{kl}, \sigma_{kl}[r, s], \xi_{kl}[r])_{k,l,r,s \in S \times S \times S \times S}$  from  $\tilde{\Omega} \times \mathbb{N} \times \Delta$  to [0, 1] introduced above. Given this framework we prove the existence of a rich Fubini extension  $(I \times \Omega, \mathcal{I} \boxtimes \mathcal{F}, \lambda \boxtimes P)$ , on which a dynamical system  $\mathbb{D}$  described in Definition 1.1 for such input probabilities is defined. More specifically, we are going to construct the space  $\hat{\Omega}$  and the probability measure  $\hat{P}$  such that  $\Omega = \tilde{\Omega} \times \hat{\Omega}$  and  $P = \tilde{P} \ltimes \hat{P}$  is a Markov kernel from  $\tilde{\Omega}$  to  $\hat{\Omega}$ .

We now present and prove Theorem 3.13 in [1]. The proof is based on Proposition 3.12 in [1], which focuses on the random matching step and shows the existence of a suitable hyperfinite probability space and partial matching, generalizing Lemma 7 in [2].

**Theorem 2.1.** Let Assumption 3.9 in [1] hold and  $(\eta_{kl}, \theta_{kl}, \xi_{kl}, \sigma_{kl}[r, s], \varsigma_{kl}[r])_{k,l,r,s\in S\times S\times S\times S}$  be the input functions from  $\tilde{\Omega} \times \mathbb{N} \times \hat{\Delta}$  defined in Section 3 in [1]. Then for any extended type distribution  $\ddot{p} \in \hat{\Delta}$  and any deterministic initial condition  $\Pi^0 = (\alpha^0, \pi^0)$  there exists a rich Fubini extension  $(I \times \Omega, \mathcal{I} \boxtimes \mathcal{F}, \lambda \boxtimes P)$  on which a discrete dynamical system  $\mathbb{D} = (\Pi^n)_{n=0}^{\infty}$  as in Definition 3.6 in [1] can be constructed with discrete time input processes  $(\eta^n, \theta^n, \xi^n, \sigma^n, \varsigma^n)_{n\geq 1}$  coming from  $(\eta_{kl}, \theta_{kl}, \xi_{kl}, \sigma_{kl}[r, s], \varsigma_{kl}[r])_{k,l,r,s\in S\times S\times S\times S}$  as stated in Section 2 in [1]. In particular,

$$\Omega = \tilde{\Omega} \times \hat{\Omega}, \quad \mathcal{F} = \tilde{\mathcal{F}} \otimes \hat{\mathcal{F}}, \quad P = \tilde{P} \ltimes \hat{P},$$

where  $(\hat{\Omega}, \hat{\mathcal{F}})$  is a measurable space and  $\hat{P}$  a Markov kernel from  $\tilde{\Omega}$  to  $\hat{\Omega}$ . The dynamical system  $\mathbb{D}$  is also MCI according to Definition 3.8 in [1] and with initial cross-sectional extended type distribution  $\hat{p}^0$  equal to  $\tilde{p}^0$  with probability one.

*Proof.* At each time period we construct three internal measurable spaces with internal transition probabilities taking into account the following steps:

- 1. random mutation
- 2. random matching
- 3. random type changing with break-up.

Let M be a limited hyperfinite number in  ${}^*\mathbb{N}_{\infty}$ . Let  $\{n\}_{n=0}^M$  be the hyperfinite discrete time line and  $(I, \mathcal{I}_0, \lambda_0)$  the agent space, where  $I = \{1, ..., \hat{M}\}, \mathcal{I}_0$  is the internal power set on  $I, \lambda_0$  is the internal counting probability measure on  $\mathcal{I}_0$ , and  $\hat{M}$  is an unlimited hyperfinite number in  ${}^*\mathbb{N}_{\infty}$ .

We start by transferring the deterministic functions  $\eta(0, \cdot), \theta(0, \cdot), \xi(0, \cdot), \sigma(0, \cdot), \varsigma(0, \cdot) : \hat{\Delta} \to [0, 1]$  to the nonstandard universe. In particular, we denote by  ${}^*\theta^0_{kl}$  for any  $k, l \in S$  and by  ${}^*f^0$  for  $f = \eta, \xi, \sigma, \varsigma$  the internal functions from  ${}^*\hat{\Delta}$  to [0, 1]. We also let  $\hat{\theta}^0_{kl}(\hat{\rho}) = {}^*\hat{\theta}^0_{kl}(\hat{\rho})$  and  $\hat{b}^0_k = 1 - \sum_{l \in S} \hat{\theta}^0_{kl}(\hat{\rho})$  for any  $k, l \in S$  and  $\hat{\rho} \in {}^*\hat{\Delta}$ , with  $1 \in {}^*\mathbb{N}$ .

We start at n = 0. To do so, we introduce the trivial probability space over the single set  $\{0\}$  denoted by  $(\bar{\Omega}_0, \bar{\mathcal{F}}_0, \bar{Q}_0)$ . Let  $\{A_{kl}\}_{(k,l)\in \hat{S}}$  be an internal partition of I such that  $\frac{|A_{kl}|}{\hat{M}} \simeq \ddot{p}_{kl}$  for any  $k \in S$  and  $l \in S \cup \{J\}$ , such that  $|A_{kk}|$  is even for any  $k, l \in S$  and  $|A_{kl}| = |A_{lk}|$  for any  $k, l \in S$ . Let  $\alpha^0$  be an internal function from  $(I, \mathcal{I}_0, \lambda_0)$  to S such that  $\alpha^0(i) = k$  if  $i \in \bigcup_{l \in S \cup \{J\}} A_{kl}$ . Let  $\pi^0$  be an internal partial matching from I to I such that  $\pi^0(i) = i$  on  $\bigcup_{k \in S} A_{kJ}$ , and the restriction  $\pi^0|_{A_{kl}}$  is an internal bijection from  $A_{kl}$  to  $A_{lk}$  for any  $k, l \in S$ . Let

$$g^{0}(i) = \begin{cases} \alpha^{0}(\pi^{0}(i)) & \text{if } \pi^{0}(i) \neq i \\ J & \text{if } \pi^{0}(i) = i. \end{cases}$$

It is clear that  $\lambda_0(\{i: \alpha^0(i) = k, g^0(i) = l\}) \simeq \ddot{p}_{kl}^0$  for any  $k \in S$  and  $l \in S \cup \{J\}$ .

Let  $(\tilde{\Omega}, \tilde{\mathcal{F}}, \tilde{P})$  be the hyperfinite internal space. Since the intensities are supposed to be deterministic at initial time, the Markov kernel from  $\tilde{\Omega}$  is trivial and we define the initial internal product probability space as

$$(\Omega_0, \mathcal{F}_0, Q_0) := (\tilde{\Omega} \times \bar{\Omega}_0, \tilde{\mathcal{F}} \otimes \bar{\mathcal{F}}_0, \tilde{P} \otimes \bar{Q}_0).$$

Suppose now that the dynamical system  $\mathbb{D}$  has been constructed up to time  $n-1 \in {}^*\mathbb{N}$  for  $n \geq 1$ , i.e., that the sequences  $\{(\Omega_m, \mathcal{F}_m, Q_m)\}_{m=0}^{3n-3}$  and  $\{\alpha^l, \pi^l\}_{l=0}^{n-1}$  have been constructed. In particular, we assume to have introduced the spaces  $(\hat{\Omega}_m, \hat{\mathcal{F}}_m)$  and the Markov kernel  $\hat{P}_m$  from  $\tilde{\Omega}$  to  $\hat{\Omega}_m$  for any  $m = 1, \ldots, n-3$ , so that we can define  $\Omega_m := \tilde{\Omega} \times \hat{\Omega}_m$  as a hyperfinite internal set with internal power set  $\mathcal{F}_m := \tilde{\mathcal{F}} \otimes \hat{\mathcal{F}}_m$  and  $Q_m := \tilde{P} \ltimes \hat{P}_m$  as an internal transition probability from  $\Omega^{m-1}$  to  $(\Omega_m, \mathcal{F}_m)$ , where

$$\Omega^m := \tilde{\Omega} \times \hat{\Omega}^m, \quad \hat{\Omega}^m := \bar{\Omega}_0 \times \prod_{j=1}^m \hat{\Omega}_j, \quad \hat{\mathcal{F}}^m := \bar{\mathcal{F}}_0 \otimes \left( \otimes_{j=1}^m \hat{\mathcal{F}}_j \right) \quad \text{and} \quad \mathcal{F}^m = \tilde{\mathcal{F}} \otimes \hat{\mathcal{F}}^m.$$
(2.1)

<sup>&</sup>lt;sup>1</sup>Note that at initial time, the functions are supposed to be deterministic and in particular independent of  $\tilde{\Omega}$ .

In this setting,  $\alpha^l$  is an internal type function from  $I \times \Omega^{3l-1}$  to the space S, and  $\pi^l$  an internal random matching from  $I \times \Omega^{3l}$  to I, such that

$$\alpha^l(i,(\tilde{\omega},\hat{\omega}^{3l-1})) = \alpha^l(i,\hat{\omega}^{3l-1}), \quad \text{for any } (\tilde{\omega},\hat{\omega}^{3l-1}) \in \Omega^{3l-1}$$

and

 $\pi^l(i,(\tilde{\omega},\hat{\omega}^{3l}))=\pi^l(i,\hat{\omega}^{3l}), \quad \text{for any } (\tilde{\omega},\hat{\omega}^{3l})\in \Omega^{3l}.$ 

Given  $\omega^{3l} \in \Omega^{3l}$  we denote by  $\pi^l_{\hat{\omega}^{3l}}: I \to I$  the function given by

$$\pi^{l}_{\hat{\omega}^{3l}}(i) := \pi^{l}(i, (\tilde{\omega}, \hat{\omega}^{3l})) = \pi^{l}(i, \hat{\omega}^{3l}).$$

A similar notation will be used for  $\alpha_{\hat{\omega}^{3l}}^l: I \to S$ . We now have the following.

#### (i) Random mutation step:

We let  $\hat{\Omega}_{3n-2} := S^I$ , which is the space of all internal functions from I to S, and denote its internal power set by  $\hat{\mathcal{F}}_{3n-2}$ . For each  $i \in I$  and  $\omega^{3n-3} = (\tilde{\omega}, \hat{\omega}^{3n-3}) \in \Omega^{3n-3}$ , if  $\alpha^{n-1}(i, \omega^{3n-3}) = \alpha^{n-1}(i, \hat{\omega}^{3n-3}) = k$ , define a probability measure  $\gamma_i^{\tilde{\omega}, \hat{\omega}^{3n-3}}$  on S by letting  $\gamma_i^{\tilde{\omega}, \hat{\omega}^{3n-3}}(l) := \theta_{kl}(\tilde{\omega}, n, \hat{\rho}_{\hat{\omega}^{3n-3}}^{n-1})$  for each  $l \in S$  with

$$\hat{\rho}_{\hat{\omega}^{3n-3}}^{n-1}[k,r] := \lambda_0(\{i \in I : \alpha_{\hat{\omega}^{3n-3}}^n(i) = k, \alpha_{\hat{\omega}^{3n-3}}^n(\pi_{\hat{\omega}^{3n-3}}^n(i)) = r\}), \quad k, r \in S$$

and

$$\hat{\rho}_{\hat{\omega}^{3n-3}}^{n-1}[k,J] := \lambda_0(\{i \in I : \alpha_{\hat{\omega}^{3n-3}}^n(i) = k, \pi_{\hat{\omega}^{3n-3}}^n(i) = i\}), \quad k \in S.$$

Define a Markov kernel  $\hat{P}_{3n-2}^{\hat{\omega}^{3n-3}}$  from  $\tilde{\Omega}$  to  $\hat{\Omega}_{3n-2}$  by letting  $\hat{P}_{3n-2}^{\hat{\omega}^{3n-3}}(\tilde{\omega})$  be the internal product measure  $\prod_{i \in I} \gamma_i^{\tilde{\omega}, \hat{\omega}^{3n-3}}$ . Define  $\bar{\alpha}^n : (I \times \Omega^{3n-2}) \to S$  by

$$\bar{\alpha}^n(i,(\tilde{\omega},\hat{\omega}^{3n-2})) := \bar{\alpha}^n(i,\hat{\omega}^{3n-2}) =: \hat{\omega}_{3n-2}(i)$$

and  $\bar{g}^n: (I \times \Omega^{3n-2}) \to S \cup \{J\}$  by

$$\bar{g}^{n}(i,(\tilde{\omega},\hat{\omega}^{3n-2})) := \bar{g}^{n}(i,\hat{\omega}^{3n-2}) := \begin{cases} \bar{\alpha}^{n}(\pi^{n-1}(i,\hat{\omega}^{3n-3}),\hat{\omega}^{3n-2}) & \text{if } \pi^{n-1}(i,\hat{\omega}^{3n-3}) \neq i \\ J & \text{if } \pi^{n-1}(i,\hat{\omega}^{3n-3})) = i. \end{cases}$$

Moreover, we introduce the notation

$$\bar{\alpha}^n_{\hat{\omega}^{3n-2}}(\cdot): I \to S, \quad \bar{\alpha}^n_{\hat{\omega}^{3n-2}}(i) := \bar{\alpha}^n(i, (\tilde{\omega}, \hat{\omega}^{3n-2})) := \bar{\alpha}^n(i, \hat{\omega}^{3n-2})$$

for the type function. We then define  $\pi_{\hat{\omega}^{3n-3}}^{n-1}(\cdot): I \to I$  and  $g_{\hat{\omega}^{3n-2}}^n: I \to S \cup \{J\}$  analogously. Finally, we define the cross-internal extended type distribution after random mutation  $\check{\rho}_{\hat{\omega}^{3n-2}}^n$  by

$$\check{\rho}^{n}_{\hat{\omega}^{3n-2}}[k,l] := \lambda_0(\{\in I : \bar{\alpha}^{n}_{\hat{\omega}^{3n-2}}(i) = k, \bar{g}^{n}_{\hat{\omega}^{3n-2}}(i) = l\}), \quad k, l \in S.$$

#### (ii) Directed random matching:

Let  $(\hat{\Omega}_{3n-1}, \hat{\mathcal{F}}_{3n-1})$  and  $\hat{P}_{3n-1}^{\hat{\omega}^{3n-2}}$  be the measurable space and the Markov kernel, respectively, provided by Proposition 3.12 in [1], with type function  $\bar{\alpha}_{\hat{\omega}^{3n-2}}^n(\cdot)$  and partial matching function  $\pi_{\hat{\omega}^{3n-3}}^{n-1}(\cdot)$ , for fixed matching probability function  $\theta(\cdot, n, \check{\rho}_{\hat{\omega}^{3n-2}}^n)$ . Proposition 3.12 in [1] also provides the directed random matching

$$\pi_{\theta^n\left(\cdot,\check{\rho}^n_{\hat{\omega}^{3n-2}}\right),\bar{\alpha}^n_{\hat{\omega}^{3n-2}},\pi^{n-1}_{\hat{\omega}^{3n-3}}},$$

which is a function defined on  $(\Omega_{3n-1}, \mathcal{F}_{3n-1})$  by

$$\pi_{\theta^{n}(\cdot,\check{\rho}_{\hat{\omega}^{3n-2}}^{n}),\bar{\alpha}_{\hat{\omega}^{3n-2}}^{n},\pi_{\hat{\omega}^{3n-3}}^{n-1}(i,(\tilde{\omega},\hat{\omega}_{3n-1})):=\pi_{\theta^{n}(\cdot,\check{\rho}_{\hat{\omega}^{3n-2}}^{n}),\bar{\alpha}_{\hat{\omega}^{3n-2}}^{n},\pi_{\hat{\omega}^{3n-3}}^{n-1}(i,\hat{\omega}_{3n-1}).$$

We then define  $\bar{\pi}^n : (I \times \Omega^{3n-1}) \to I$  by

$$\bar{\pi}^n(i,(\tilde{\omega},\hat{\omega}_{3n-1})) := \bar{\pi}^n(i,\hat{\omega}^{3n-1}) := \pi_{\theta^n(\cdot,\check{\rho}^n_{\hat{\omega}^{3n-2}}),\bar{\alpha}^n_{\hat{\omega}^{3n-2}},\pi^{n-1}_{\hat{\omega}^{3n-3}}}(i,\hat{\omega}_{3n-1})$$

and

$$\bar{\bar{g}}^n(i,(\tilde{\omega},\hat{\omega}^{3n-1})) = \bar{\bar{g}}^n(i,\hat{\omega}^{3n-1}) := \begin{cases} \bar{\alpha}^n(\bar{\pi}^n(i,\hat{\omega}^{3n-1}),\hat{\omega}^{3n-2}) & \text{if } \bar{\pi}^n(i,\hat{\omega}^{3n-1}) \neq i \\ J & \text{if } \bar{\pi}^n(i,\hat{\omega}^{3n-1}) = i. \end{cases}$$

Define now the cross-internal extended type distribution after the random matching  $\check{\check{\rho}}^n_{\hat{\omega}^{3n-1}}$  by

$$\check{p}^{n}_{\hat{\omega}^{3n-1}}[k,l] := \lambda_0(\{\in I : \bar{\alpha}^{n}_{\hat{\omega}^{3n-1}}(i) = k, \bar{\bar{g}}^{n}_{\hat{\omega}^{3n-1}}(i) = l\}).$$

#### (iii) Random type changing with break-up for matched agents:

Introduce  $\hat{\Omega}_{3n} := (S \times \{0, 1\})^I$  with internal power set  $\hat{\mathcal{F}}_{3n}$ , where 0 represents "unmatched" and 1 represents "paired"; each point  $\hat{\omega}_{3n} = (\hat{\omega}_{3n}^1, \hat{\omega}_{3n}^2) \in \hat{\Omega}_{3n}$  represents an internal function from I to  $S \times \{0, 1\}$ . Define a new type function  $\alpha^n : (I \times \Omega^{3n}) \to S$  by letting  $\alpha^n(i, (\tilde{\omega}, \hat{\omega}^{3n})) := \alpha^n(i, \hat{\omega}^{3n}) = \hat{\omega}_{3n}^1(i)$ . Fix  $(\tilde{\omega}, \hat{\omega}^{3n-1}) \in \Omega^{3n-1}$ . For each  $i \in I$ , we proceed in the following way.

- 1. If  $\bar{\pi}^n(i,\hat{\omega}^{3n-1}) = i$  (*i* is not paired after the matching step at time *n*), let  $\tau_i^{\tilde{\omega},\hat{\omega}^{3n-1}}$  be the probability measure on the type space  $S \times \{0,1\}$  that gives probability one to the type  $(\bar{\alpha}^n(i,(\tilde{\omega},\hat{\omega}^{3n-2})),0) = (\bar{\alpha}^n(i,\hat{\omega}^{3n-2}),0)$  and zero to the rest
- 2. If  $\bar{\pi}^n(i, (\tilde{\omega}, \hat{\omega}^{3n-1})) = \bar{\pi}^n(i, \hat{\omega}^{3n-1}) = j \neq i$  (*i* is paired after the matching step at time *n*),  $\bar{\alpha}^n(i, (\tilde{\omega}, \hat{\omega}^{3n-2})) = \bar{\alpha}^n(i, \hat{\omega}^{3n-2}) = k, \bar{\pi}^n(i, (\tilde{\omega}, \hat{\omega}^{3n-1})) = \bar{\pi}^n(i, \hat{\omega}^{3n-1}) = j$  and  $\bar{\alpha}^n(j, (\tilde{\omega}, \hat{\omega}^{3n-1})) = \bar{\alpha}^n(j, \hat{\omega}^{3n-1}) = l$ , define a probability measure  $\tau_{ij}^{\tilde{\omega}, \hat{\omega}^{3n-1}}$  on  $(S \times \{0, 1\}) \times (S \times \{0, 1\})$  as

$$\tau_{ij}^{\tilde{\omega},\tilde{\omega}^{3n-1}}((k',0),(l',0)) := \left(1 - \xi_{kl}(\tilde{\omega},n,\check{\rho}_{\hat{\omega}^{3n-1}}^n)\right)\varsigma_{kl}[k']\left(\tilde{\omega},n,\check{\rho}_{\hat{\omega}^{3n-1}}^n\right)\varsigma_{lk}[l']\left(\tilde{\omega},n,\check{\rho}_{\hat{\omega}^{3n-1}}^n\right)$$

and

$$\tau_{ij}^{\tilde{\omega},\tilde{\omega}^{3n-1}}\left((k',1),(l',1)\right) := \xi_{kl}\left(\tilde{\omega},n,\check{p}_{\hat{\omega}^{3n-1}}^{n}\right)\sigma_{kl}[k',l']\left(\tilde{\omega},n,\check{p}_{\hat{\omega}^{3n-1}}^{n}\right)$$

for  $k', l' \in S$ , and zero for the rest.

Let  $A^n_{\hat{\omega}^{3n-1}} = \{(i,j) \in I \times I : i < j, \bar{\pi}^n(i, (\tilde{\omega}, \hat{\omega}^{3n-1})) = \bar{\pi}^n(i, \hat{\omega}^{3n-1}) = j\}$  and  $B^n_{\hat{\omega}^{3n-1}} = \{i \in I : \bar{\pi}^n(i, (\tilde{\omega}, \hat{\omega}^{3n-1})) = \bar{\pi}^n(i, \hat{\omega}^{3n-1}) = i\}$ . Define a Markov kernel  $\hat{P}^{\hat{\omega}^{3n-1}}_{3n}$  from  $\tilde{\Omega}$  to  $\hat{\Omega}^{3n}$  by

$$\hat{P}_{3n}^{\hat{\omega}^{3n-1}}(\tilde{\omega}) := \prod_{i \in B^n_{\tilde{\omega},\hat{\omega}^{3n-1}}} \tau_i^{\tilde{\omega},\hat{\omega}^{3n-1}} \otimes \prod_{(i,j) \in A^n_{\hat{\omega}^{3n-1}}} \tau_{ij}^{\tilde{\omega},\hat{\omega}^{3n-1}}.$$

Let

$$\begin{aligned} \pi^{n}(i,(\tilde{\omega},\hat{\omega}^{3n})) &= \pi^{n}(i,\hat{\omega}^{3n}) \\ &:= \begin{cases} J & \text{if } \bar{\pi}^{n}(i,\hat{\omega}^{3n-1}) = J \text{ or } \hat{\omega}_{3n}^{2}(i) = 0 \text{ or } \hat{\omega}_{3n}^{2}(\bar{\pi}^{n}(i,\hat{\omega}^{3n-1})) = 0 \\ \bar{\pi}^{n}(i,\hat{\omega}^{3n-1}) & \text{otherwise,} \end{cases} \end{aligned}$$

and

$$g^{n}(i,(\hat{\omega},\hat{\omega}^{3n})) = g^{n}(i,\hat{\omega}^{3n}) := \begin{cases} \alpha^{n}(\pi^{n}(i,\hat{\omega}^{3n}),\hat{\omega}^{3n}) & \text{if } \pi^{n}(i,\hat{\omega}^{3n}) \neq i \\ J & \text{if } \pi^{n}(i,\hat{\omega}^{3n}) = i. \end{cases}$$

Define  $\hat{\rho}^n_{\hat{\omega}^{3n}} = \lambda_0(\alpha^n_{\hat{\omega}^{3n}}, \pi^n_{\hat{\omega}^{3n}})^{-1}.$ 

By repeating this procedure, we construct a hyperfinite sequence of internal transition probability spaces  $\{(\Omega_m, \mathcal{F}_m, Q_m)\}_{m=0}^{3M}$  and a hyperfinite sequence of internal type functions and internal random matchings  $\{(\alpha^n, \pi^n)\}_{n=0}^M$ . Moreover, define  $(\Omega^m, \mathcal{F}^m)$  as in (2.1), and

$$\hat{P}^m := \prod_{i=1}^m \hat{P}_i, \qquad Q^m := \tilde{P} \ltimes \hat{P}^m$$

where the product of the Markov kernels is  $\tilde{\omega}$ -wise.

Let  $(I \times \Omega^{3M}, \mathcal{I}_0 \otimes \mathcal{F}^{3M}, \lambda_0 \otimes Q^{3M})$  be the internal product probability space of  $(I, \mathcal{I}_0, \lambda_0)$  and  $(\Omega^{3M}, \mathcal{F}^{3M}, Q^{3M})$ . Denote the Loeb spaces of  $(\Omega^{3M}, \mathcal{F}^{3M}, Q^{3M})$  and the internal product  $(I \times \Omega^{3M}, \mathcal{I}_0 \otimes \mathcal{F}^{3M}, \lambda_0 \otimes Q^{3M})$  by  $(\Omega^{3M}, \mathcal{F}, P)$  and  $(I \times \Omega^{3M}, \mathcal{I} \boxtimes \mathcal{F}, \lambda \boxtimes P)$ , respectively. For simplicity, let  $\Omega^{3M}$  be denoted by  $\Omega$  and  $\hat{\Omega}^{3M}$  by  $\hat{\Omega}$ . Denote now  $Q^{3M}$  by P and the Markov kernel  $\hat{P}^{3M}$  by  $\hat{P}$ .

The properties of a dynamical system as well as the independence conditions follow now by applying similar arguments as in the proof of Theorem 5 in [2] for any fixed  $\tilde{\omega} \in \tilde{\Omega}$ . The only difference is that in our setting the input processes for the random mutation step and the break-up step also depend on the extended type distribution. Furthermore, these arguments are similar to the ones in the proof of Lemma 3.2 and can be found there with all details.

# 3 Proof of Theorem 3.14 in [1]

We now prove Theorem 3.14 in [1] which is a generalization of the results in Appendix C in [2]. For  $n \ge 1$ we define the mapping  $\Gamma^n$  from  $\tilde{\Omega} \times \hat{\Delta}$  to  $\hat{\Delta}$  by

$$\Gamma_{kl}^{n}(\tilde{\omega}, \hat{p}) = \sum_{k_{1}, l_{1} \in S} (1 - \xi_{k_{1}l_{1}}(\tilde{\omega}, n, \tilde{\tilde{p}}^{n}))\sigma_{k_{1}l_{1}}[k, l] \left(\tilde{\omega}, n, \tilde{\tilde{p}}^{n}\right) \tilde{p}_{k_{1}l_{1}}^{n} \\
+ \sum_{k_{1}, l_{1} \in S} (1 - \xi_{k_{1}l_{1}}(\tilde{\omega}, n, \tilde{\tilde{p}}^{n}))\sigma_{k_{1}l_{1}}[k, l] \left(\tilde{\omega}, n, \tilde{\tilde{p}}^{n}\right) \theta_{k_{1}l_{1}}(\tilde{\omega}, n, \tilde{p}^{n}) \tilde{p}_{k_{1}J}^{n},$$
(3.1)

and

$$\Gamma_{kJ}^{n}(\tilde{\omega}, \hat{p}) = b_{k}(\tilde{\omega}, n, \tilde{p}^{n})\tilde{p}_{kJ}^{n} + \sum_{k_{1}, l_{1} \in S} \xi_{k_{1}l_{1}}(\tilde{\omega}, n, \tilde{\tilde{p}}^{n})\varsigma_{k_{1}l_{1}}[k](\tilde{\omega}, \tilde{\tilde{p}}^{n})\tilde{p}_{k_{1}l_{1}}^{n} + \sum_{k_{1}, l_{1} \in S} \xi_{k_{1}l_{1}}(\tilde{\omega}, n, \tilde{\tilde{p}}^{n})\varsigma_{k_{1}l_{1}}[k](\tilde{\omega}, n, \tilde{\tilde{p}}^{n})\theta_{k_{1}l_{1}}(\tilde{\omega}, n, \tilde{p}^{n})\tilde{p}_{k_{1}J}^{n}$$
(3.2)

with

$$\tilde{p}_{kl}^n = \sum_{k_1, l_1 \in S} \eta_{k_1 k}(\tilde{\omega}, n, \hat{p}) \eta_{l_1 l}(\tilde{\omega}, n, \hat{p}) \hat{p}_{k_1 l_1}$$
$$\tilde{p}_{kJ}^n = \sum_{l \in S} \hat{p}_{lJ} \eta_{lk}(\tilde{\omega}, n, \hat{p})$$

and

$$\begin{split} \tilde{\tilde{p}}_{kl}^n &= \tilde{p}_{kl}^n + \theta_{kl}(\tilde{\omega}, n, \tilde{p}^n) \tilde{p}_{kJ}^n \\ \tilde{\tilde{p}}_{kJ}^n &= b_k(\tilde{\omega}, n, \tilde{p}^n) \tilde{p}_{kJ}^n. \end{split}$$

Theorem 3.14 in [1] is proven with the help of the following lemmas.

**Lemma 3.1.** Assume that the discrete dynamical system  $\mathbb{D}$  defined in Definition 3.6 in [1] is Markov conditionally independent given  $\tilde{\omega}$  as defined in Definition 3.8 in [1]. Then given  $\tilde{\omega} \in \tilde{\Omega}$ , the discrete time processes  $\{\beta_i^n\}_{n=0}^{\infty}, i \in I$ , are essentially pairwise independent on  $(I \times \hat{\Omega}, \mathcal{I} \boxtimes \hat{\mathcal{F}}, \lambda \boxtimes \hat{P}^{\tilde{\omega}})$ . Moreover, for fixed n = 1, ..., M also  $(\bar{\beta}_i^n)_{n=0}^{\infty}$  and  $(\bar{\beta}_i^n)_{n=0}^{\infty}, i \in I$ , are essentially pairwise independent on  $(I \times \hat{\Omega}, \mathcal{I} \boxtimes \hat{\mathcal{F}}, \lambda \boxtimes \hat{\mathcal{P}}^{\tilde{\omega}})$ .

*Proof.* This can be proven by the same arguments used in the proof of Lemma 3 in [2].  $\Box$ 

We now derive a result which shows how to compute for a fixed  $\tilde{\omega} \in \tilde{\Omega}$  the expected cross-sectional distributions  $\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^n]$ ,  $\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^n]$  and  $\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^n]$ .

**Lemma 3.2.** The following holds for any fixed  $\tilde{\omega} \in \tilde{\Omega}$ .

- 1. For each  $n \geq 1$ ,  $\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^n] = \Gamma^n(\tilde{\omega}, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}])$ , with  $\Gamma$  defined in (3.1).
- 2. For each  $n \ge 1$ , we have

$$\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{kl}^{n}] = \sum_{k_{1}, l_{1} \in S} \eta_{k_{1}, k}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}])\eta_{l_{1}, l}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}])\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}_{k_{1}, l_{1}}^{n-1}]$$

and

$$\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{kJ}^{n}] = \sum_{k_1 \in S} \eta_{k_1,k}(\tilde{\omega},n,\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}])\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}_{k_1,J}^{n-1}].$$

3. For each  $n \ge 1$ , we have

$$\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{kl}^{n}] = \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{kl}^{n}] + \theta_{kl}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}])\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{kJ}^{n}]$$

and

$$\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{kJ}^{n}] = b_{k}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}])\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{kJ}^{n}].$$

Proof. Fix  $\tilde{\omega} \in \tilde{\Omega}$  and  $k, l \in S$ . By Lemma 3.1 we know that the processes  $(\beta_i^n)_{n=0}^{\infty}, i \in I$ , are essentially pairwise independent. Then the exact law of large numbers in Lemma 1 in [2] implies that  $\hat{p}^{n-1}(\hat{\omega}) = \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\lambda(\beta^{n-1})^{-1}]$  for  $\hat{P}$ -almost all  $\hat{\omega} \in \hat{\Omega}$ . Thus equations (1.3) and (1.4) are equivalent to

$$\hat{P}^{\tilde{\omega}}\left(\bar{\alpha}_{i}^{n}=k_{2},\bar{g}_{i}^{n}=l_{2}|\alpha_{i}^{n-1}=k_{1},g_{i}^{n-1}=l_{1}\right)=\eta_{k_{1},k_{2}}\left(\tilde{\omega},n,\mathbb{E}^{P^{\tilde{\omega}}}[\hat{p}^{n-1}]\right)\eta_{l_{1},l_{2}}\left(\tilde{\omega},n,\mathbb{E}^{P^{\tilde{\omega}}}[\hat{p}^{n-1}]\right)$$
(3.3)

$$\hat{P}^{\tilde{\omega}}\left(\bar{\alpha}_{i}^{n}=k_{2},\bar{g}_{i}^{n}=r|\alpha_{i}^{n-1}=k_{1},g_{i}^{n-1}=J\right)=\eta_{k_{1},k_{2}}\left(\tilde{\omega},n,\mathbb{E}^{P^{\tilde{\omega}}}[\hat{p}^{n-1}]\right)\delta_{J}(r).$$
(3.4)

Therefore, for any  $k_1, l_1 \in S$  we have

$$\hat{P}^{\tilde{\omega}}\left(\bar{\beta}_{i}^{n}=(k,J)|\beta_{i}^{n-1}=(k_{1},l_{1})\right)=0$$
(3.5)

$$\hat{P}^{\tilde{\omega}}\left(\bar{\beta}_{i}^{n}=(k,l)|\beta_{i}^{n-1}=(k_{1},J)\right)=0.$$
(3.6)

Then with the same calculations as in the proof of Lemma 4 in [2] we get that

$$\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{kl}^{n}] = \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\lambda(i \in I : \bar{\beta}_{\omega}^{n}(i) = (k, l))]$$

$$= \int_{I} \hat{P}^{\tilde{\omega}}(\bar{\beta}_{i}^{n} = (k, l))d\lambda(i)$$

$$= \sum_{k_{1}, l_{1} \in S} \eta_{k_{1}, k}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}])\eta_{l_{1}, l}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}])\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}_{k_{1}l_{1}}^{n-1}]$$

$$(3.7)$$

and

$$\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{kJ}^{n}] = \sum_{k_{1} \in S} \eta_{k_{1},k}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}]) \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}_{kJ}^{n-1}].$$
(3.8)

By Lemma 3.1 we know that  $\bar{\beta}^n$  is essentially pairwise independent. Again it follows by the exact law of large numbers that  $\check{p}^n(\hat{\omega}) = \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^n]$  for  $\hat{P}^{\tilde{\omega}}$ -almost all  $\hat{\omega} \in \hat{\Omega}$ . Then (1.6) and (1.7) are equivalent to

$$\hat{P}^{\tilde{\omega}}(\bar{\bar{g}}^n = l | \bar{\alpha}^n_i = k, \bar{g}^n_i = J) = \theta_{kl}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^n])$$
(3.9)

$$\hat{P}^{\tilde{\omega}}(\bar{g}^n = J | \bar{\alpha}^n_i = k, \bar{g}^n_i = J) = b_k(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^n]).$$
(3.10)

By the same calculations as in the proof of Lemma 4 in [2] we have

$$\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{kl}^{n}] = \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{kl}^{n}] + \theta_{kl}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}])\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{kJ}^{n}]$$
(3.11)

and

$$\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}_{kJ}] = b_{k}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}]) \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}_{kJ}].$$
(3.12)

By Lemma 3.1,  $\bar{\beta}^n$  is essentially pairwise independent and thus  $\check{p}^n(\hat{\omega}) = \mathbb{E}^{\hat{P}^{\hat{\omega}}}[\check{p}^n]$  for  $\hat{P}$ -almost all  $\hat{\omega} \in \hat{\Omega}$ . Then (1.11) and (1.12) are equivalent to

$$\hat{P}^{\tilde{\omega}}(\alpha_{i}^{n}=l_{1},g_{i}^{n}=l_{2}|\alpha_{i}^{n}=k_{1},\bar{g}_{i}^{n}=k_{2})=(1-\xi_{k_{1}k_{2}}(\tilde{\omega},n,\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}]))\sigma_{k_{1}k_{2}}[l_{1},l_{2}]\left(\tilde{\omega},n,\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}]\right)$$

and

$$\hat{P}^{\tilde{\omega}}(\alpha_{i}^{n}=l_{1},g_{i}^{n}=J|\alpha_{i}^{n}=k_{1},\bar{g}_{i}^{n}=k_{2})=\xi_{k_{1}k_{2}}(\tilde{\omega},n,\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}])\varsigma_{k_{1}k_{2}}[l_{1},l_{2}]\left(\tilde{\omega},n,\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}]\right),$$

respectively. Thus

$$\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}_{kl}^{n}] = \sum_{k_{1},l_{1}\in S} (1 - \xi_{k_{1}l_{1}}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}]))\sigma_{k_{1}l_{1}}[k, l] \left(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}]\right) \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}_{k_{1}l_{1}}]$$
(3.13)

and

$$\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}_{kJ}^{n}] = \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{kJ}^{n}] + \sum_{k_{1},l_{1}\in S} \xi_{k_{1}l_{1}}(\tilde{\omega},n,\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}])\varsigma_{k_{1}l_{1}}[k]\left(\tilde{\omega},n,\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}]\right)\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{k_{1}l_{1}}^{n}].$$
(3.14)

By plugging (3.8) in (3.13) we get

 $\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^n_{kl}]$ 

$$= \sum_{k_{1},l_{1}\in S} (1 - \xi_{k_{1}l_{1}}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}]))\sigma_{k_{1}l_{1}}[k, l] \left(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}]\right) \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}_{k_{1}l_{1}}] \\ + \sum_{k_{1},l_{1}\in S} (1 - \xi_{k_{1}l_{1}}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}]))\sigma_{k_{1}l_{1}}[k, l] \left(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}]\right) \eta_{k_{1}l_{1}}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}])\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}_{k_{1}J}].$$
(3.15)

By using (3.12) and (3.13), it follows that

$$\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}_{kJ}^{n}] = b_{k}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}])\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{kJ}^{n}] \\
+ \sum_{k_{1}, l_{1} \in S} \xi_{k_{1}l_{1}}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}])\varsigma_{k_{1}l_{1}}[k] \left(\tilde{\omega}, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}]\right) \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{k_{1}l_{1}}^{n}] \\
+ \sum_{k_{1}, l_{1} \in S} \xi_{k_{1}l_{1}}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}])\varsigma_{k_{1}l_{1}}[k] \left(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}]\right) \theta_{k_{1}l_{1}}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}]) \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{k_{1}J}].$$

$$(3.16)$$

**Lemma 3.3.** Assume that the discrete dynamical system  $\mathbb{D}$  defined in Definition 3.6 in [1] is Markov conditionally independent given  $\tilde{\omega} \in \tilde{\Omega}$  according to Definition Definition 3.8 in [1]. Then for fixed  $\tilde{\omega} \in \tilde{\Omega}$  the following holds:

- 1. For  $\lambda$ -almost all  $i \in I$ , the extended type process  $\{\beta_i^n\}_{n=0}^{\infty}$  for agent i is a Markov chain on  $(I \times \hat{\Omega}, \mathcal{I} \boxtimes \hat{\mathcal{F}}, \lambda \boxtimes \hat{P}^{\tilde{\omega}})$  with transition matrix  $z^n$  after time n-1.
- 2.  $\{\beta^n\}_{n=0}^{\infty}$  is also a Markov chain with transition matrix  $z^n$  at time n-1.

*Proof.* Fix  $\tilde{\omega} \in \tilde{\Omega}$ .

1. The Markov property of  $\{\beta_i^n\}_{n=0}^{\infty}$  on  $(I \times \hat{\Omega}, \mathcal{I} \boxtimes \hat{\mathcal{F}}, \lambda \boxtimes \hat{P}^{\tilde{\omega}})$  follows by using the same arguments as in the proof of Lemma 5 in [2], for  $\lambda$ -almost all  $i \in I$ . We now derive the transition matrix with similar arguments as in [2]. By putting together (3.7), (3.8) and (3.15), we get

$$\begin{split} \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}_{kl}^{n}] \\ &= \sum_{k_{1},l_{1},k',l' \in S} (1 - \xi_{k_{1}l_{1}}(\tilde{\omega},n,\tilde{p}^{\tilde{\omega},n}))\sigma_{k_{1}l_{1}}[k,l] \left(\tilde{\omega},n,\tilde{p}^{\tilde{\omega},n}\right)\eta_{k'k_{1}}(\tilde{\omega},n,\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}]) \\ &\quad \cdot \eta_{l'l_{1}}(\tilde{\omega},n,\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}])\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}_{k'l'}^{n-1}] \\ &\quad + \sum_{k_{1},l_{1},k' \in S} (1 - \xi_{k_{1}l_{1}}(\tilde{\omega},n,\tilde{p}^{\tilde{\omega},n}))\sigma_{k_{1}l_{1}}[k,l] \left(\tilde{\omega},n,\tilde{p}^{\tilde{\omega},n}\right)\theta_{k_{1}l_{1}}(\tilde{\omega},n,\tilde{p}^{\tilde{\omega},n}) \\ &\quad \cdot \eta_{k'k_{1}}(\tilde{\omega},n,\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}])\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}_{k'J}^{n-1}]. \end{split}$$

Thus we have

$$z_{(k'J)(kl)}^{n}(\tilde{\omega}) = \sum_{k_{1},l_{1}\in S} (1 - \xi_{k_{1}l_{1}}(\tilde{\omega},n,\tilde{p}^{\tilde{\omega},n}))\sigma_{k_{1}l_{1}}[k,l]\left(\tilde{\omega},n,\tilde{p}^{\tilde{\omega},n}\right)\theta_{k_{1}l_{1}}(\tilde{\omega},n,\tilde{p}^{\tilde{\omega},n})$$
$$\cdot \eta_{k'k_{1}}(\tilde{\omega},n,\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}])$$
(3.17)

and

$$z_{(k'l')(kl)}^{n}(\tilde{\omega}) = \sum_{k_1, l_1 \in S} (1 - \xi_{k_1 l_1}(\tilde{\omega}, n, \tilde{\tilde{p}}^{\tilde{\omega}, n})) \sigma_{k_1 l_1}[k, l] \left(\tilde{\omega}, n, \tilde{\tilde{p}}^{\tilde{\omega}, n}\right) \eta_{k'k_1}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}])$$

$$\cdot \eta_{l'l_1}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}]).$$
(3.18)

Similarly, equations (3.7), (3.8) and (3.16) yield to

$$\begin{split} \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}_{kJ}^{n}] &= \sum_{k' \in S} b_{k}(\tilde{\omega}, n, \tilde{p}^{\tilde{\omega}, n}) \eta_{k'k}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}]) \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}_{k'J}] \\ &+ \sum_{k_{1}, l_{1}, k', l' \in S} \xi_{k_{1}l_{1}}(\tilde{\omega}, n, \tilde{p}^{\tilde{\omega}, n}) \varsigma_{k_{1}l_{1}}[k] \left[\tilde{\omega}, n, \tilde{p}^{\tilde{\omega}, n}\right] \\ &\cdot \eta_{k'k_{1}}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}]) \eta_{l'l_{1}}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}]) \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}] \\ &+ \sum_{k_{1}, l_{1}, k' \in S} \xi_{k_{1}l_{1}}(\tilde{\omega}, n, \tilde{p}^{\tilde{\omega}, n}) \varsigma_{k_{1}l_{1}}[k] \left(\tilde{\omega}, n, \tilde{p}^{\tilde{\omega}, n}\right) \theta_{k_{1}l_{1}}(\tilde{\omega}, n, \tilde{p}^{\tilde{\omega}, n}) \\ &\cdot \eta_{k'k}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}]) \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}]. \end{split}$$

Therefore, the transition probabilities from time n-1 to time n can be written as

$$z_{(k'l')(kJ)}^{n}(\tilde{\omega}) = \sum_{k_{1},l_{1}\in S} \xi_{k_{1}l_{1}}(\tilde{\omega},n,\tilde{p}^{\tilde{\omega},n})\varsigma_{k_{1}l_{1}}[k]\left(\tilde{\omega},n,\tilde{p}^{\tilde{\omega},n}\right)$$
$$\cdot \eta_{k'k_{1}}(\tilde{\omega},n,\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}])\eta_{l'l_{1}}(\tilde{\omega},n,\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}])$$
(3.19)

and

$$z_{(k'J)(kJ)}^{n}(\tilde{\omega}) = b_{k}(\tilde{\omega}, n, \tilde{p}^{\tilde{\omega}, n})\eta_{k'k}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}]) + \sum_{k_{1}, l_{1} \in S} \xi_{k_{1}l_{1}}(\tilde{\omega}, n, \tilde{\tilde{p}}^{\tilde{\omega}, n})\varsigma_{k_{1}l_{1}}[k] \left(\tilde{\omega}, n, \tilde{\tilde{p}}^{\tilde{\omega}, n}\right) \theta_{k_{1}l_{1}}(\tilde{\omega}, n, \tilde{p}^{\tilde{\omega}, n})\eta_{k'k}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}]).$$
(3.20)

2. The transition matrix of  $\{\beta^n\}_{n=0}^{\infty}$  at time n-1 can be derived by using (3.17)-(3.20) and the Fubini property applied to  $\lambda \boxtimes \hat{P}^{\tilde{\omega}}$  for every fixed  $\tilde{\omega} \in \tilde{\Omega}$  as in the proof of Lemma 6 in [2].

We are now able to prove Theorem 3.14 in [1], which we present here.

**Theorem 3.4.** Assume that the discrete dynamical system  $\mathbb{D}$  introduced in Definition 3.6 in [1] is Markov conditionally independent given  $\tilde{\omega} \in \tilde{\Omega}$  according to Definition 3.8 in [1]. Given  $\tilde{\omega} \in \tilde{\Omega}$ , the following holds:

- 1. For each  $n \geq 1$ ,  $\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^n] = \Gamma^n(\tilde{\omega}, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}]).$
- 2. For each  $n \ge 1$ , we have

Π

$$\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{kl}^{n}] = \sum_{k_{1},l_{1}\in S} \eta_{k_{1},k}(\tilde{\omega},n,\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}])\eta_{l_{1},l}(\tilde{\omega},n,\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}])\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}_{k_{1},l_{1}}^{n-1}]$$

and

$$\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{kJ}^{n}] = \sum_{k_1 \in S} \eta_{k_1,k}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}]) \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}_{k_1,J}^{n-1}].$$

3. For each  $n \ge 1$ , we have

$$\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}_{kl}] = \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}_{kl}] + \theta_{kl}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}])\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}_{kJ}]$$

and

$$\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{kJ}^{n}] = b_{k}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{kJ}^{n}])\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}_{kJ}^{n}].$$

4. For  $\lambda$ -almost every agent *i*, the extended-type process  $\{\beta_i^n\}_{n=0}^{\infty}$  is a Markov chain in  $\hat{S}$  on  $(I \times \hat{\Omega}, \mathcal{I} \boxtimes \hat{\mathcal{F}}, \lambda \boxtimes \hat{P}^{\tilde{\omega}})$ , whose transition matrix  $z^n$  at time n-1 is given by

$$z_{(k'J)(kl)}^{n}(\tilde{\omega}) = \sum_{k_{1},l_{1},k'\in S} (1 - \xi_{k_{1}l_{1}}(\tilde{\omega},n,\tilde{p}^{\tilde{\omega},n}))\sigma_{k_{1}l_{1}}[k,l]\left(\tilde{\omega},n,\tilde{p}^{\tilde{\omega},n}\right)\theta_{k_{1}l_{1}}(\tilde{\omega},n,\tilde{p}^{\tilde{\omega},n})$$
$$\cdot \eta_{k'k_{1}}(\tilde{\omega},n,\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}])$$
(3.21)

$$z_{(k'l')(kl)}^{n}(\tilde{\omega}) = \sum_{k_{1},l_{1},k',l'\in S} (1 - \xi_{k_{1}l_{1}}(\tilde{\omega},n,\tilde{p}^{\tilde{\omega},n}))\sigma_{k_{1}l_{1}}[k,l] \left(\tilde{\omega},n,\tilde{p}^{\tilde{\omega},n}\right)\eta_{k'k_{1}}(\tilde{\omega},n,\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}]) \\ \cdot \eta_{l'l_{1}}(\tilde{\omega},n,\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}])$$
(3.22)

$$z_{(k'l')(kJ)}^{n}(\tilde{\omega}) = \sum_{k_{1},l_{1}\in S} \xi_{k_{1}l_{1}}(\tilde{\omega},n,\tilde{p}^{\tilde{\omega},n})\varsigma_{k_{1}l_{1}}[k]\left(\tilde{\omega},n,\tilde{p}^{\tilde{\omega},n}\right)$$
$$\cdot \eta_{k'k_{1}}(\tilde{\omega},n,\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}])\eta_{l'l_{1}}(\tilde{\omega},n,\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}])$$
(3.23)

$$z_{(k'J)(kJ)}^{n}(\tilde{\omega}) = b_{k}(\tilde{\omega}, n, \tilde{p}^{\tilde{\omega}, n})\eta_{k'k}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}]) + \sum_{k_{1}, l_{1} \in S} \xi_{k_{1}l_{1}}(\tilde{\omega}, n, \tilde{\tilde{p}}^{\tilde{\omega}, n})\varsigma_{k_{1}l_{1}}[k]\left(\tilde{\omega}, n, \tilde{\tilde{p}}^{\tilde{\omega}, n}\right)\theta_{k_{1}l_{1}}(\tilde{\omega}, n, \tilde{p}^{\tilde{\omega}, n}) \cdot \eta_{k'k_{1}}(\tilde{\omega}, n, \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n-1}]).$$

$$(3.24)$$

- 5. For  $\lambda$ -almost every i and every  $\lambda$ -almost every j, the Markov chains  $\{\beta_i^n\}_{n=0}^{\infty}$  and  $\{\beta_j^n\}_{n=0}^{\infty}$  are independent on  $(\hat{\Omega}, \hat{\mathcal{F}}, \hat{P}^{\tilde{\omega}})$ .
- 6. For  $\hat{P}^{\tilde{\omega}}$ -almost every  $\hat{\omega} \in \hat{\Omega}$ , the cross sectional extended type process  $\{\beta^n_{\tilde{\omega}}\}_{n=0}^{\infty}$  is a Markov chain on  $(I, \mathcal{I}, \lambda)$  with transition matrix  $z^n$  at time n-1, which is defined in (3.21)- (3.24).
- 7. We have  $\hat{P}^{\tilde{\omega}}$ -a.s. that

$$\mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{p}^{n}_{kl}] = \check{p}^{n}_{kl} \quad and \quad \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\check{\check{p}}^{n}_{kl}] = \check{p}^{n}_{kl} \quad and \quad \mathbb{E}^{\hat{P}^{\tilde{\omega}}}[\hat{p}^{n}_{kl}] = \hat{p}^{n}_{kl}.$$

*Proof.* Fix  $\tilde{\omega} \in \tilde{\Omega}$ . Points 1. to 5. of Theorem 3.14 in [1] follow directly by Lemma 3.1, 3.2 and 3.3. Moreover, Points 6. and 7. can be proven by using the same arguments as in the proof of Theorem 4 in [2].

# References

- Francesca Biagini, Andrea Mazzon, Thilo Meyer-Brandis, and Katharina Oberpriller. Liquidity based modeling of asset price bubbles via random matching. *preprint*, 2022. URL http://arxiv.org/abs/ 2210.13804.
- [2] Darrell Duffie, Lei Qiao, and Yeneng Sun. Dynamic directed random matching. Journal of Economic Theory, 174:124–183, 2018.