

# Convergence results in multitype, multivariate, branching random walk

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## Abstract

A multitype branching random walk on  $d$ -dimensional Euclidian space is considered. The uniform convergence, as  $n$  goes to infinity, of a scaled version of the Laplace transform of the point process given by the  $n$ th generation particles of each type is obtained. Similar results in the one-type case, where the transform gives a martingale, have been obtained in Biggins (1992) and Barral (2001). This uniform convergence of transforms is then used to obtain limit results for numbers in the underlying point processes. Supporting results, which are of interest in their own right, are obtained on (i) ‘Perron-Frobenius theory’ for matrices that are smooth functions of a variable  $\lambda \in L$  and are non-negative when  $\lambda \in L_- \subset L$  and (ii) saddlepoint approximations of multivariate distributions. The saddlepoint approximations developed are strong enough to give a refined large deviation theorem of Chaganty and Sethuraman (1993) as a by-product.

# 1 Introduction

We consider the multitype branching random walk on the  $d$ -dimensional Euclidian space,  $\mathbb{R}^d$ . The process starts with a single particle located at the origin. This particle produces daughter particles, which are scattered in  $\mathbb{R}^d$ , to give the first generation. These first generation particles produce daughter particles to give the second generation, and so on. As usual in branching processes, the  $n$ th generation particles reproduce independently of each other. Each particle in this process has one of  $p$  types; the set of possible types is identified with  $\{1, \dots, p\}$ . For each  $i \in \{1, \dots, p\}$  there is a vector of point processes  $(Z_{i1}, Z_{i2}, \dots, Z_{ip})$ . Then, when a particle of type  $i$  reproduces the positions of its daughter particles of the various types, relative to the parent's position, are given by a copy of  $(Z_{i1}, Z_{i2}, \dots, Z_{ip})$ . The one-type branching random walk has an extensive literature. The multitype extension has received less attention, but discussion of it can be found in Mode (1971), Biggins (1976, 1996), Bramson *et al.* (1992) and Kyprianou and Rahimzadeh Sani (2001).

We keep  $i, j$  and  $k$ , for the types, drawn from  $\{1, \dots, p\}$ . Let  $\mu_{ij}$  be the intensity measure of  $Z_{ij}$ . We assume throughout that

$$\text{there is a } \vartheta \in \mathbb{R}^d \text{ such that } \max_{i,j} \int_{\mathbb{R}^d} e^{-\vartheta'x} \mu_{ij}(dx) < \infty, \quad (1.1)$$

where  $\vartheta'x = \sum_i \vartheta_i x_i$  is the usual inner product of vectors. This condition is enough to ensure that convolutions of the  $\mu_{ij}$  produce well-defined measures.

Technically, the description of the process outlines how to produce the probability measure on the space of trajectories given the type of the single initial ancestor. There are  $p$  such measures, one for each possible initial type. Suppose the initial ancestor has type  $i$  and write  $Z_{ij}^n$  be the point process giving the resulting positions of the type  $j$  particles in generation  $n$ ; then  $Z_{ij}^1$  is distributed like  $Z_{ij}$ . The first objective here is to obtain the asymptotic behaviour, as  $n \rightarrow \infty$ , of the Laplace transform of  $Z_{ij}^n$ . Note that results about  $Z_{ij}^n$  are, in fact, results for the measure on trajectories obtained when the single initial ancestor has type  $i$ . Incorporating the initial type into the notation in this way, rather than through the measure as  $P_i$  and  $E_i$ , is helpful in calculations.

Before the theorems can be stated, further notation about intensity measures and their transforms is needed. Define  $\mu_{ij}^{n\star}$  inductively by

$$\mu_{ij}^{(n+1)\star} = \sum_{k=1}^p \mu_{ik} * \mu_{kj}^{n\star}$$

where “ $*$ ” is ordinary convolution of measures. It is easy to confirm, by induction on  $n$ , that the point process  $Z_{ij}^n$  has the intensity measure  $\mu_{ij}^{n\star}$ . Furthermore the counts of the numbers of each type in each generation, given by  $(Z_{i1}^n(\mathbb{R}^d), \dots, Z_{ip}^n(\mathbb{R}^d))$ , is a multitype Galton-Watson process, which is discussed in Athreya and Ney (1972), for example. A matrix  $A$  of non-negative entries is called positive regular when for some positive integer  $n$ ,  $A^n$  has all its entries strictly positive. The multitype Galton-Watson process is positive regular when the matrix  $(P(Z_{ij}(\mathbb{R}^d) > 0))$  is positive regular. Throughout, the embedded Galton-Watson process is assumed to be positive regular. This Galton-Watson process is supercritical when the largest eigenvalue of its mean matrix,  $(EZ_{ij}(\mathbb{R}^d))$ , exceeds one. (For a positively regular process this eigenvalue will be infinite when the mean matrix

has any infinite entries.) A supercritical process survives with positive probability. One of the conditions of our theorems will imply that the process is supercritical.

The  $d$ -dimensional complex space  $\mathbb{C}^d$  is equipped with the maximum metric. Hence, for  $x = (x_1, \dots, x_d)$ ,  $y = (y_1, \dots, y_d) \in \mathbb{C}^d$ , we have  $d(x, y) = \max\{|x_i - y_i| : i = 1, \dots, d\}$  where  $|x_i - y_i|$  is the usual absolute value in  $\mathbb{C}$ . Let  $B(x, r)$  be the open ball centred at  $x$  of radius  $r$  using this metric and let  $\overline{B}(x, r)$  be its closure. (Later we will also need  $S(x, r)$  and  $\overline{S}(x, r)$  for the open and closed balls in  $\mathbb{R}^d$ .) In all that follows, we keep the letters  $\theta$  and  $\eta$  for the real and imaginary parts of  $\lambda \in \mathbb{C}^d$ , so that, with this convention,  $\lambda = \theta + \mathbf{i}\eta$ . For any  $A \subset \mathbb{C}^d$  let  $A_-$  be its intersection with  $\mathbb{R}^d$ ; thus

$$A_- = \{\lambda \in A : \lambda = \theta + \mathbf{i}\eta \text{ with } \eta = 0\}.$$

Define the Laplace transforms  $m_{ij}(\lambda)$ , for  $\lambda \in \mathbb{C}^d$ , by

$$m_{ij}(\lambda) = \int_{\mathbb{R}^d} e^{-\lambda x} \mu_{ij}(dx) \left( = E \int_{\mathbb{R}^d} e^{-\lambda x} Z_{ij}(dx) \right).$$

Then  $|m_{ij}(\lambda)| \leq m_{ij}(\theta)$  and, by Hölder's inequality,  $\{\theta \in \mathbb{R}^d : m_{ij}(\theta) < \infty\}$  is a convex set. Let  $\text{int}A$  be the interior of  $A$  and let

$$L = \bigcap_{i,j} \text{int}\{\lambda = \theta + \mathbf{i}\eta \in \mathbb{C}^d : m_{ij}(\theta) < \infty\}.$$

Assumption (1.1) is now strengthened to:

$L$  is non-empty.

Then  $L$  is a non-empty, open, convex subset of  $\mathbb{C}^d$  and each  $m_{ij}(\lambda)$  is analytic in  $\lambda \in L$ . Let  $M(\lambda)$  be the matrix given by  $M(\lambda) = (m_{ij}(\lambda))$  and let  $M^n(\lambda)$  be its  $n$ th power, with  $(i, j)$ th entry  $m_{ij}^n(\lambda)$ . Furthermore, let

$$\mathcal{M}_{ij}^n(\lambda) = \int_{\mathbb{R}^d} e^{-\lambda x} Z_{ij}^n(dx).$$

Then

$$m_{ij}^n(\lambda) = \int_{\mathbb{R}^d} e^{-\lambda x} \mu_{ij}^{n*}(dx) = E\mathcal{M}_{ij}^n(\lambda).$$

By analogy with the case when the entries are non-negative, the eigenvalue  $\rho$  of  $M$  is called the *maximum-modulus eigenvalue* if it is a simple root of  $\det[zI - M]$  and its modulus is strictly larger than that of all other roots. Note that when a maximum-modulus eigenvalue exists it is automatically unique.

When  $\vartheta \in L_- (= L \cap \mathbb{R}^d)$ , the entries of  $M(\vartheta)$  are finite non-negative real numbers. The positive regularity of the embedded Galton-Watson process easily implies that  $M(\vartheta)$  is then positively regular. This means that the following extension of the Perron-Frobenius, which is discussed in the final section, applies to  $M$ . In essence, in the present context, it says that Perron-Frobenius properties extend smoothly to complex arguments that are near to real ones. Part (iii) gives an asymptotic estimate, equation (1.2), of  $M^n(\lambda)$ , as  $n \rightarrow \infty$ . The key point about (1.2) is that it gives a uniform bound on the rate of convergence in a suitable neighbourhood. This uniformity has no direct parallel in the Perron-Frobenius Theorem and needed a little care to establish.

**Theorem 1** Suppose the  $p \times p$  matrix  $M = (m_{ij})$  of functions defined on the open set  $L \subset \mathbb{C}^d$  satisfies the following condition: for all  $i$  and  $j$ ,  $m_{ij}(\lambda)$  are analytic functions in  $\lambda \in L$ , and for all  $\tilde{\lambda} \in \tilde{L} \subset L$ ,  $M(\tilde{\lambda})$  is positive regular.

Then there is an open set  $\Omega \subset L$  containing  $\tilde{L}$  such that for  $\lambda \in \Omega$  the following hold.

(i)  $M(\lambda)$  has a unique maximum-modulus eigenvalue,  $\rho(\lambda)$ , that is analytic in  $\lambda$ .

(ii) The left and right eigenvectors associated with  $\rho(\lambda)$ ,  $u(\lambda)$  and  $v(\lambda)$ , normalised so that  $\sum_{i=1}^p u_i(\lambda) = 1$  and  $\sum_{i=1}^p u_i(\lambda)v_i(\lambda) = 1$ , are analytic in  $\lambda$  and, for all  $i$ ,  $u_i(\lambda) \neq 0$  and  $v_i(\lambda) \neq 0$ .

(iii) For any  $\vartheta \in \tilde{L}$  there is a neighbourhood,  $B$ , within  $\Omega$  and containing  $B(\vartheta, \delta)$  for some  $\delta > 0$ , and constants  $K < \infty$  and  $\gamma \in (0, 1)$  such that, for all  $n$ ,  $i$  and  $j$ ,

$$\sup_{\lambda \in B} |\rho(\lambda)^{-n} (M(\lambda)^n)_{ij} - v_i(\lambda)u_j(\lambda)| \leq K\gamma^n. \quad (1.2)$$

All the notation introduced in Theorem 1 is now applied to  $M$ , the matrix of Laplace transforms, with  $\tilde{L} = L_-$ . Thus  $M$  has maximum-modulus eigenvalue  $\rho$  and eigenvectors  $u$  and  $v$  defined on the set  $\Omega$ , with the properties described in Theorem 1. For all  $\lambda \in \Omega$ ,  $n = 1, 2, \dots$ , and  $i, j$ , define

$$\mathcal{W}_{ij}^n(\lambda) = \frac{v_j(\lambda)}{v_i(\lambda)} \rho(\lambda)^{-n} \mathcal{M}_{ij}^n(\lambda) = \frac{v_j(\lambda)}{v_i(\lambda)} \rho(\lambda)^{-n} \int_{\mathbb{R}^d} e^{-\lambda'x} Z_{ij}^n(dx), \quad (1.3)$$

which is a scaled version of the Laplace transform of the point process  $Z_{ij}^n$ . Establishing the uniform convergence of such transforms in the one-type case is an integral part of the approach to results on the distribution of the points described in Biggins (1992). For the multitype case, results on the distribution of points have been obtained in Bramson *et al.* (1992) when particles are confined to the integer lattice of  $\mathbb{R}$ . More precise results, of the kind given in Biggins (1992), should hold and should not be limited to the lattice case. One such result is given here in Theorem 7 below.

Let  $\mathcal{F}_n$  be the  $\sigma$ -algebra that contains all information on the multitype branching random walk up to generation  $n$ . In the one-type case, (1.3) defines a martingale with respect to  $\mathcal{F}_n$ , which makes aspects of the study simpler. Here, it is

$$\mathcal{W}_i^n(\lambda) = \sum_j \mathcal{W}_{ij}^n(\lambda) = \sum_j \frac{v_j(\lambda)}{v_i(\lambda)} \rho(\lambda)^{-n} \int_{\mathbb{R}^d} e^{-\lambda'x} Z_{ij}^n(dx) \quad (1.4)$$

that is a martingale with respect to  $\mathcal{F}_n$  (see Lemma 2.1). Our approach to the convergence of  $\{\mathcal{W}_{ij}^n(\lambda)\}$  involves considering the convergence of the martingale  $\{\mathcal{W}_i^n(\lambda)\}$  first. Note that, given any sample path of the process, each  $\mathcal{W}_{ij}^n(\lambda)$  and  $\mathcal{W}_i^n(\lambda)$  is analytic in  $\lambda \in \Omega$ .

We need to introduce certain sets that will be used to define when convergence occurs. For  $\alpha \in (1, 2]$ , let

$$\Omega_\alpha^2 = \left\{ \lambda \in \Omega : \alpha\lambda \in \Omega, \frac{\rho(\alpha\theta)}{|\rho(\lambda)|^\alpha} < 1 \right\}, \quad (1.5)$$

$$\Omega_\alpha^3 = \text{int} \left\{ \lambda = \theta + \mathbf{i}\eta \in \Omega : \max_i \{E[\mathcal{W}_i^1(\theta)^\alpha]\} < \infty \right\}, \quad (1.6)$$

$$\Lambda_\alpha = \Omega_\alpha^2 \cap \Omega_\alpha^3 \quad \text{and} \quad \Lambda = \bigcup_{1 < \alpha \leq 2} \Lambda_\alpha.$$

These are all open sets in  $\mathbb{C}^d$ . Error estimates we derive involve  $(\rho(\alpha\theta)/|\rho(\lambda)|^\alpha)^n$ . Hence  $\lambda \in \Omega_\alpha^2$  ensures that such bounds decay quickly with  $n$ . Obviously,  $\lambda \in \Omega_\alpha^3$  imposes a

moment condition. The approach requires  $\lambda$  to satisfy both these conditions for the same  $\alpha \in (1, 2]$ , which leads to the definition of  $\Lambda_\alpha$  and then  $\Lambda$ .

The process is supercritical, that is,  $\rho(0) > 1$ , whenever there is a  $\theta \in \Omega_\alpha^2$  for some  $\alpha \in (1, 2]$ . To check this, note first that  $\log \rho(\theta)$  is a convex function, see Kingman (1961), Miller (1961) or Seneta (1973, Theorem 3.7). Hence

$$\frac{(\alpha - 1)}{\alpha} \log \rho(0) + \frac{1}{\alpha} \log \rho(\alpha\theta) \geq \log \rho(\theta),$$

and  $\log(\rho(\alpha\theta)/\rho(\theta)^\alpha) < 0$  when  $\theta \in \Omega_\alpha^2$ . Then  $(\alpha - 1) \log \rho(0) > 0$  and so  $\rho(0) > 1$ .

**Theorem 2** *Let  $\alpha \in (1, 2]$  and  $\lambda = \theta + \mathbf{i}\eta \in \Lambda_\alpha$ . Then  $\{\mathcal{W}_i^n(\lambda)\}$  converges almost surely and in  $\alpha$ th mean, as  $n \rightarrow \infty$ , for each  $i$ .*

This gives the following result on mean convergence as an immediate consequence.

**Corollary 1** *Suppose  $\lambda \in \Lambda$ . Then  $\{\mathcal{W}_i^n(\lambda)\}$  converges almost surely and in mean, as  $n \rightarrow \infty$ , for each  $i$ .*

The mean convergence of  $\{\mathcal{W}_i^n(\lambda)\}$  for real  $\lambda$  is also discussed in Bramson *et al.* (1992), Kyprianou and Rahimzadeh Sani (2001) and, rather briefly, in Biggins and Kyprianou (2004).

**Theorem 3** *Assume  $\Lambda_- \neq \emptyset$ . Then there is an open set  $\Gamma$ , with  $\Lambda_- \subset \Gamma \subset \Lambda$ , such that the martingale  $\{\mathcal{W}_i^n(\lambda)\}$  converges uniformly in any compact subset of  $\Gamma$ , almost surely and in mean, as  $n \rightarrow \infty$ , for each  $i$ . Then the limit  $\mathcal{W}_i(\lambda)$  is analytic in  $\lambda \in \Gamma$ .*

Building on these convergence results for the martingale, analogous results will be obtained for  $\{\mathcal{W}_{ij}^n(\lambda)\}$ .

**Theorem 4** *Let  $\alpha \in (1, 2]$  and assume that  $(\Lambda_\alpha)_- \neq \emptyset$ . Then there is an open set  $\Gamma_1$  with  $(\Lambda_\alpha)_- \subset \Gamma_1 \subset \Lambda_\alpha$ , such that for all  $\lambda \in \Gamma_1$ ,  $\{\mathcal{W}_{ij}^n(\lambda)\}$  converges to  $u_j(\lambda)v_j(\lambda)\mathcal{W}_i(\lambda)$ , almost surely and in  $\alpha$ th mean, as  $n \rightarrow \infty$ , where  $\mathcal{W}_i(\lambda)$  is the limit of the martingale  $\{\mathcal{W}_i^n(\lambda)\}$  as  $n \rightarrow \infty$ .*

**Theorem 5** *Assume  $\Lambda_- \neq \emptyset$ . Then there is an open set  $\Gamma_2$ , satisfying  $\Lambda_- \subset \Gamma_2 \subset \Lambda$ , such that for all  $i, j$ ,  $\{\mathcal{W}_{ij}^n(\lambda)\}$  converges uniformly in any compact subset of  $\Gamma_2$ , almost surely and in mean, as  $n \rightarrow \infty$ , to the random variable  $u_j(\lambda)v_j(\lambda)\mathcal{W}_i(\lambda)$ , where  $\mathcal{W}_i(\lambda)$  is the limit of the martingale  $\{\mathcal{W}_i^n(\lambda)\}$  as  $n \rightarrow \infty$ .*

The martingales  $\mathcal{W}_i^n(\lambda)$  and their components  $\mathcal{W}_{ij}^n(\lambda)$  are only defined on the set  $\Omega$  introduced in Theorem 1. However, to move from information on transforms to information on the associated measures a result on the behaviour of  $\mathcal{M}_{ij}^n(\lambda)$  is needed for  $\theta \in L_-$  but where  $\lambda$  need not be in  $\Omega$ .

The branching random walk is *strongly non-lattice* when it is positively regular and, for some  $(k, l)$  and for some  $\theta \in \{\vartheta : m_{kl}(\vartheta) < \infty\}$

$$\left| \frac{m_{kl}(\theta + \mathbf{i}\eta)}{m_{kl}(\theta)} \right| = 1 \quad \text{only when } \eta = 0. \quad (1.7)$$

This follows the usage of ‘strongly non-lattice’ in (1.64) in Bhattacharya (1977) rather than that in Stone (1965).

**Theorem 6** For a strongly non-lattice branching random walk, for any set  $\mathcal{K} \subset \Lambda_-$  that is compact in  $\mathbb{R}^d$  and any  $\epsilon \in (0, 1)$  there is an  $\varepsilon < 1$  such that

$$\varepsilon^{-n} \sup_{i,j} \sup_{\theta \in \mathcal{K}} \sup_{\epsilon \leq |\eta| \leq \epsilon^{-1}} \left| \frac{\mathcal{M}_{ij}^n(\theta + \mathbf{i}\eta)}{\rho^n(\theta)} \right| \rightarrow 0 \quad \text{as } n \rightarrow \infty \quad (1.8)$$

almost surely.

Between them, Theorems 5 and 6 provide enough information on the behaviour of the transforms  $\mathcal{M}_{ij}^n(\lambda)$  to develop good estimates of the associated measures. In the one type case, in Biggins (1992), this step relies heavily on the corresponding results for sums of independent identically distributed random variables, obtained in Stone (1967). Unfortunately this approach does not extend directly, since the existing results for Markov additive processes, which are the analogue here of independent identically distributed random variables, are not suitable. Instead, the treatment has to go back to the general methods for saddlepoint approximations. The results needed, which give an approximation to a measure based only on a few of its characteristics, are contained in Theorems 9 and 10 in Section 5. That section is independent of the rest of the paper and of more general interest. We illustrate there how Theorems 9 and 10 connect with refined large deviation results in Chaganty and Sethuraman (1993) and Stone (1967).

It is worth noting that in this study the measures of interest,  $Z_{ij}^n$ , are atomic. Hence if (1.7) were replaced by the stronger, but easier to work with, Cramér's condition, (1.36) in Bhattacharya (1977), that condition could not transfer to the  $Z_{ij}^n$ .

The assumption that  $\log \rho$  is strictly convex will be imposed, which amounts to saying that the branching random walk is truly  $d$ -dimensional, but we do not demonstrate this. For  $\theta \in L_-$ , let  $(-1)^i v_i(\theta)$  be the array of  $i$ th derivatives of  $\log \rho(\theta)$ . Then  $v_2(\theta)$  is a positive-definite matrix when  $\log \rho$  is strictly convex. Let  $\mathfrak{D}$  be the continuously differentiable functions and if  $f \in \mathfrak{D}$  let its vector of derivatives be  $f'$ . We need functions that decay suitably at infinity and so we introduce

$$\mathfrak{G}(G) = \left\{ f \in \mathfrak{D} : \int_0^\infty \sup \{ |f(x)|, |f'(x)| : |x| \geq (r-1) \} (r+1)^{d+1} dr \leq G \right\}.$$

**Theorem 7** Assume  $\log \rho$  is strictly convex and the process is strongly non-lattice. Let  $\mathcal{K}$  be a compact subset of  $\Lambda_-$ . Let  $h$  be such that, for some finite  $G$ , for all  $\theta \in \mathcal{K}$  the function  $e^{\theta x} h(x)$  is in  $\mathfrak{G}(G)$ . Then, with  $\xi(\theta) = -\theta v_1(\theta) - \log \rho(\theta)$  and  $C$  any convex set in  $\mathbb{R}^d$ ,

$$n^{d/2} e^{n\xi(\theta)} \int_C h(x) Z_{ij}^n(dx + n v_1(\theta)) \rightarrow \frac{v_i(\theta) u_j(\theta) \mathcal{W}_i(\theta)}{\sqrt{(2\pi)^d \det[v_2(\theta)]}} \int_C e^{\theta x} h(x) dx$$

uniformly in  $\theta \in \mathcal{K}$  and  $C$ , almost surely on  $\mathcal{S}$ , where  $\mathcal{S}$  is the survival set of the underlying Galton-Watson process.

The condition that  $e^{\theta x} h(x)$  is in  $\mathfrak{G}(G)$  becomes more restrictive as  $\mathcal{K}$  becomes bigger. In particular, it forces  $h(x)$  to decay rapidly with  $|x|$  when the origin is in the interior of  $\mathcal{K}$ .

**Corollary 2** *Let  $b < \infty$ . Under the conditions of Theorem 7,*

$$n^{d/2} e^{n\xi(\theta)} Z_{ij}^n(C + nv_1(\theta)) \rightarrow \frac{v_i(\theta)u_j(\theta)\mathcal{W}_i(\theta)}{\sqrt{(2\pi)^d \det[v_2(\theta)]}} \int_C e^{\theta'x} dx$$

*uniformly in convex  $C \subset \{x : |x| \leq b\}$  and  $\theta \in \mathcal{K}$ , almost surely on  $\mathcal{S}$ .*

It is worth pointing out that if, for some  $y$ ,  $\mathcal{K}$  lies inside the half-space  $\{\theta : \theta'y > 0\}$  then the result in this corollary can also hold for some  $C$  that are not bounded.

The relationships between  $\rho$ ,  $v_1$  and  $\xi$  are the usual ones associated with large deviation and saddlepoint calculations; see Section 2.2 in Jensen (1995) for example. The assumption that  $\rho$  is strictly convex means that compact subsets of  $\Lambda_-$  translate, under  $y = v_1(\theta)$  to compact subsets of  $\text{int}\{v_1(\theta) : \theta \in L_-\}$ .

Theorem 7 has the following analogue for the measures  $\mu_{ij}^{n*}$ , with the proof requiring only obvious changes.

**Theorem 8** *Under the conditions of Theorem 7,*

$$n^{d/2} e^{n\xi(\theta)} \int_C h(x) \mu_{ij}^{n*}(dx + nv_1(\theta)) \rightarrow \frac{v_i(\theta)u_j(\theta)}{\sqrt{(2\pi)^d \det[v_2(\theta)]}} \int_C e^{\theta'x} h(x) dx$$

*uniformly in  $\theta \in \mathcal{K}$  and  $C$ .*

## 2 Proofs of Theorems 2 and 3

For any  $\lambda = \theta + \mathbf{i}\eta \in \Omega$ , define the functions:

$$\bar{v}(\lambda) = \max_{i,j} \{|v_j(\lambda)/v_i(\lambda)|\}, \quad \underline{v}(\lambda) = \min_{i,j} \{|v_j(\lambda)/v_i(\lambda)|\}, \quad \nu(\lambda) = \frac{\bar{v}(\lambda)}{\underline{v}(\lambda)}$$

and

$$\phi(\lambda) = \frac{\rho(\theta)}{|\rho(\lambda)|}.$$

All of these are strictly positive, continuous functions in  $\lambda \in \Omega$ , where  $\Omega$  comes from Theorem 1. Also, since  $|m_{ij}(\lambda)| \leq m_{ij}(\theta)$ , we know  $|\rho(\lambda)| \leq \rho(\theta)$  (Lancaster and Tismenetsky (1985, Theorem 15.2.1)) and so  $\phi(\lambda) \geq 1$ .

For each  $\alpha \in (1, 2]$  define  $\Omega_\alpha$  by  $\Omega_\alpha = \{\lambda \in \Omega : \alpha\lambda \in \Omega\}$ . Then  $\Omega_\alpha$  is an open subset of  $\Omega$ . For  $\alpha \in (1, 2]$  and  $\lambda \in \Omega_\alpha$ , let

$$\nu_1(\lambda) = \frac{\bar{v}(\lambda)^\alpha}{\underline{v}(\alpha\theta)} \quad \text{and} \quad \kappa(\lambda) = \frac{\rho(\alpha\theta)}{|\rho(\lambda)|^\alpha}.$$

Then  $\nu_1$  and  $\kappa$  are strictly positive continuous functions in  $\lambda \in \Omega_\alpha$ ; they depend on  $\alpha$ , but this has been suppressed in the notation. Note that  $\Omega_\alpha^2$ , defined at (1.5), can now be written as  $\Omega_\alpha^2 = \{\lambda \in \Omega_\alpha : \kappa(\lambda) < 1\}$  and is an open subset of  $\Omega_\alpha$ . Define

$$\beta(\theta) = \max_i \{E[\mathcal{W}_i^1(\theta)^\alpha]\}.$$

Then  $\Omega_\alpha^3$ , defined at (1.6), can be written as  $\Omega_\alpha^3 = \text{int}\{\lambda = \theta + \mathbf{i}\eta \in \Omega : \beta(\theta) < \infty\}$  and  $\beta$  is a real valued continuous function on  $\Omega_\alpha^3$ .

Let  $\alpha \geq 1$ . Define the  $\alpha$ th absolute central moment of  $X$ ,  $\sigma^\alpha(X)$ , by

$$\sigma^\alpha(X) = E |X - E[X]|^\alpha,$$

and the  $\alpha$ th absolute central moment conditional on the  $\sigma$ -algebra  $\mathcal{G}$  by

$$\sigma^\alpha(X | \mathcal{G}) = E [|X - E[X | \mathcal{G}]|^\alpha | \mathcal{G}].$$

Let  $\{z_{ik;s}^l : s\}$  be the positions of the particles making up  $Z_{ik}^l$ . Thus  $z_{ik;s}^l$  is the position of the  $s$ th particle of type  $k$  in generation  $l$  when the initial ancestor is of type  $i$ . Now, by looking at the particles in the  $n$ th generation as the  $(n-l)$ th generation children of the particles in generation  $l$ , we can introduce  $Z_{kj}^n(\cdot | l, s)$  as the point process giving the positions of type  $j$  particles in generation  $n$  descended from the  $s$ th particle with type  $k$  in generation  $l$  and relative to that particle's position. Given  $\mathcal{F}_l$ ,  $\{Z_{kj}^n(\cdot | l, s) : s\}$  are independent copies of the point process  $Z_{kj}^{n-l}$ . Thus, conditional on  $\mathcal{F}_l$ , the random variables

$$\mathcal{W}_k^{n-l}(\lambda | l, s) = \sum_j \frac{v_j(\lambda)}{v_k(\lambda)} \frac{1}{\rho(\lambda)^{n-l}} \int_{\mathbb{R}^d} e^{-\lambda'x} Z_{kj}^n(dx | l, s) \quad (2.1)$$

are, as  $s$  varies, independent and identical copies of  $\mathcal{W}_k^{n-l}(\lambda)$ . Furthermore,

$$\mathcal{W}_i^n(\lambda) = \sum_k \frac{v_k(\lambda)}{v_i(\lambda)} \frac{1}{\rho(\lambda)^l} \sum_s \exp(-\lambda'z_{ik;s}^l) \mathcal{W}_k^{n-l}(\lambda | l, s). \quad (2.2)$$

**Lemma 2.1**  $\{\mathcal{W}_i^n(\lambda)\}$  is a martingale with respect to  $\{\mathcal{F}_n\}$ .

**Proof.** From Theorem 1,  $E[\mathcal{W}_k^1(\lambda)] = 1$ . Then decomposition (2.2) with  $l = n - 1$  gives the result.  $\square$

The following result, which is an extension of inequality (4) in von Bahr and Esseen (1965) to complex valued random variables, is proved in Biggins (1992). Barral (2001) provides a very pretty approach to uniform convergence in the one-type case by applying a Lemma like this to Banach-space valued random variables.

**Lemma 2.2** *If  $\{X_n\}$  are independent complex valued random variables with  $E[X_n] = 0$ , or more generally, martingale differences, then for  $\alpha \in (1, 2]$ , there exists a constant  $C > 0$ , depending on  $\alpha$  but independent of  $n$  and the sequence  $\{X_n\}$ , such that*

$$E \left| \sum_{j=1}^n X_j \right|^\alpha \leq C \sum_{j=1}^n E |X_j|^\alpha.$$

The next lemma, which is the multitype version of Lemma 2 in Biggins (1992), gives some bounds related to the martingale  $\{\mathcal{W}_i^n(\lambda)\}$ . In bounding formulae like these the argument  $\lambda$  will often be omitted; thus, for example,  $\phi$  is  $\phi(\lambda)$ .

**Lemma 2.3** *Let  $\alpha \in (1, 2]$ ,  $\lambda = \theta + \mathbf{i}\eta \in \Lambda_\alpha$  and  $\varphi = \nu_1 \phi^\alpha \nu^\alpha \beta$ . Then there are constants  $c_1, c_2$ , and  $c_3$ , depending on  $\alpha$  but not on  $\lambda, i$  or  $n$ , such that the following hold.*

$$(i) \quad (E |\mathcal{W}_i^{n+1}(\lambda) - \mathcal{W}_i^n(\lambda)|)^\alpha \leq E |\mathcal{W}_i^{n+1}(\lambda) - \mathcal{W}_i^n(\lambda)|^\alpha \leq c_1 \varphi \kappa^n;$$



- (ii)  $\sigma^\alpha(\mathcal{W}_i^n(\lambda)) \leq c_2\varphi(1 - \kappa)^{-1}$ ;
- (iii)  $\sum_{k=n}^{\infty} E |\mathcal{W}_i^{k+1}(\lambda) - \mathcal{W}_i^k(\lambda)| \leq c_3\varphi^{1/\alpha}\kappa^{n/\alpha}(1 - \kappa^{1/\alpha})^{-1}$ ;
- (iv) For  $\vartheta \in (\Lambda_\alpha)_-$ , there is a  $\delta > 0$ , such that

$$\sup\{E |\mathcal{W}_i^n(\lambda)|^\alpha : \lambda \in B(\vartheta, \delta), i, n\} < \infty.$$

**Proof.** The first inequality in (i) is just  $E|X|^\alpha \geq [E|X|]^\alpha$ . Since  $E|\mathcal{W}_i^1(\lambda)|^\alpha \geq 1$ , we get

$$E |\mathcal{W}_i^1(\lambda) - 1|^\alpha \leq 4E |\mathcal{W}_i^1(\lambda)|^\alpha \leq 4\phi^\alpha\nu^\alpha\beta.$$

Using (2.2), Lemma 2.2 and that  $\mathcal{W}_k^1(\lambda|n, s)$  are independent copies of  $\mathcal{W}_k^1(\lambda)$  given  $\mathcal{F}_n$ ,

$$\begin{aligned} \sigma^\alpha(\mathcal{W}_i^{n+1}(\lambda) | \mathcal{F}_n) &= E [|\mathcal{W}_i^{n+1}(\lambda) - \mathcal{W}_i^n(\lambda)|^\alpha | \mathcal{F}_n] \\ &\leq C \sum_k \left| \frac{v_k(\lambda)}{v_i(\lambda)\rho(\lambda)^n} \right|^\alpha \sum_s |\exp(-\lambda'z_{ik;s}^n)|^\alpha E |\mathcal{W}_k^1(\lambda) - 1|^\alpha \\ &\leq c_1\nu_1\kappa^n\mathcal{W}_i^n(\alpha\theta)(\phi^\alpha\nu^\alpha\beta) = c_1\varphi\kappa^n\mathcal{W}_i^n(\alpha\theta); \end{aligned}$$

taking expectations now gives (i).

Note that, since  $\lambda \in \Lambda_\alpha$ ,  $\kappa < 1$ . Now, part (ii) follows directly from summing over  $n$  in (i) and using Lemma 2.2. To prove (iii), sum the inequalities in (i).

The continuity of the functions  $\nu, \nu_1, \phi, \kappa$ , on  $\Lambda_\alpha$  and part (ii) mean that there is a constant  $C > 0$ , depending only on  $\alpha$  and  $\delta > 0$ , such that for any  $\lambda \in \bar{B}(\vartheta, \delta)$ ,  $\sigma^\alpha(\mathcal{W}_i^n(\lambda)) \leq C$ . Since  $E[|\mathcal{W}_i^n(\lambda)|^\alpha] \leq 2 + 2\sigma^\alpha(\mathcal{W}_i^n(\lambda))$ , this implies (iv).  $\square$

**Proof of Theorem 2.** From part (iii) of Lemma 2.3,  $E \sum_{n=0}^{\infty} |\mathcal{W}_i^{n+1}(\lambda) - \mathcal{W}_i^n(\lambda)|$  is finite, and so  $\{\mathcal{W}_i^n(\lambda)\}$  is a Cauchy sequence almost surely. Thus as  $n \rightarrow \infty$ ,  $\{\mathcal{W}_i^n(\lambda)\}$  converges to  $\mathcal{W}_i(\lambda)$  almost surely. Applying Fatou's lemma, Lemma 2.2 and Lemma 2.3(i) gives

$$\begin{aligned} E|\mathcal{W}_i(\lambda) - \mathcal{W}_i^n(\lambda)|^\alpha &\leq \liminf_{N \rightarrow \infty} E|\mathcal{W}_i^{n+N}(\lambda) - \mathcal{W}_i^n(\lambda)|^\alpha \\ &\leq C \sum_{j=0}^{\infty} E |\mathcal{W}_i^{n+j+1}(\lambda) - \mathcal{W}_i^{n+j}(\lambda)|^\alpha \\ &\leq Cc_1\varphi \frac{\kappa^n}{1 - \kappa}. \end{aligned}$$

Since  $\kappa < 1$ ,  $\{\mathcal{W}_i^n(\lambda)\}$  converges in  $\alpha$ th mean.  $\square$

The distinguished boundary (see Hörmander (1973)) of  $B(x, r)$ , which is a subset of the topological boundary, is defined by

$$D(x, r) = \{y \in \mathbb{C}^d : |x_s - y_s| = r; s = 1, \dots, d\}.$$

We assume  $D(x, r)$  is parameterised so that

$$D(x, r) = \{z(t) = (z_1(t), \dots, z_d(t)) : z_s(t) = x_s + re^{its}; s = 1, \dots, d; t \in I\} \quad (2.3)$$

where  $I = [0, 2\pi]^d$  is a  $d$ -dimensional closed cube in  $\mathbb{R}^d$ . The next Lemma, which is Lemma 3 in Biggins (1992), is the key to obtaining bounds that hold uniformly.

**Lemma 2.4** *If  $f$  is analytic on the open ball  $B = B(x, 2r')$  and  $r < r'$  then*

$$\sup \{|f(\lambda)| : \lambda \in B(x, r)\} \leq \pi^{-d} \int_I |f(z(t))| dt,$$

where  $z(t) \in D(x, 2r)$ .

**Proof of Theorem 3.** The proof follows closely the one-type result in Biggins (1992). Let  $x \in \Lambda_-$  be fixed. There is an  $\alpha \in (1, 2]$  and a  $B = B(x, 3r) \subset \Lambda_\alpha$ , which means that  $B \subset \Omega_\alpha^2$  and  $B \subset \Omega_\alpha^3$ . For any two positive integers  $n$  and  $N$  ( $\geq n$ ), we apply Lemma 2.4 to  $\mathcal{W}_i^{N+1}(\lambda) - \mathcal{W}_i^n(\lambda)$  to get

$$\begin{aligned} \pi^d \sup_{\lambda \in B(x, r)} |\mathcal{W}_i^{N+1}(\lambda) - \mathcal{W}_i^n(\lambda)| &\leq \int_I \sum_{m=n}^N |\mathcal{W}_i^{m+1}(z(t)) - \mathcal{W}_i^m(z(t))| dt, \\ &\leq \sum_{m=n}^{\infty} \int_I |\mathcal{W}_i^{m+1}(z(t)) - \mathcal{W}_i^m(z(t))| dt, \end{aligned}$$

where  $z(t) \in D(x, 2r) \subset B$ . Applying Fubini's theorem and Lemma 2.3(iii), we get

$$E \sum_{m=n}^{\infty} \int_I |\mathcal{W}_i^{m+1}(z(t)) - \mathcal{W}_i^m(z(t))| dt \leq \int_I \left[ C\varphi^{1/\alpha} \frac{\kappa^{n/\alpha}}{1 - \kappa^{1/\alpha}} \right] dt,$$

where the argument  $z(t)$  has been suppressed on the right. Recall that  $\kappa < 1$  throughout  $\Omega_\alpha^2$  and  $\beta < \infty$  throughout  $\Omega_\alpha^3$ . Therefore, using the continuity of the various functions, there are constants  $\delta < 1$  and  $K < \infty$  such that the integrand on the right can be bounded by  $K\delta^n$  throughout  $B$  and hence  $D(x, 2r)$ . Thus

$$E \sum_{m=n}^{\infty} \int_I |\mathcal{W}_i^{m+1}(z(t)) - \mathcal{W}_i^m(z(t))| dt \leq K\delta^n \int_I dt < \infty$$

and so

$$\sum_{m=0}^{\infty} \int_I |\mathcal{W}_i^{m+1}(z(t)) - \mathcal{W}_i^m(z(t))| dt < \infty, \quad \text{almost surely.}$$

Therefore  $\{\mathcal{W}_i^n(\lambda)\}$  is a Cauchy sequence under the supremum norm  $\|\cdot\|$  on  $B(x, r)$ , which implies uniform convergence on  $B(x, r)$ , almost surely.

For the remainder of this proof, the supremum norm is defined over  $B(x, r)$ . From the almost sure uniform convergence of the martingale, for fixed  $n$ , as  $N \rightarrow \infty$ ,

$$\|\mathcal{W}_i^N(\lambda) - \mathcal{W}_i^n(\lambda)\| \longrightarrow \|\mathcal{W}_i(\lambda) - \mathcal{W}_i^n(\lambda)\|,$$

almost surely. Let  $n$  be fixed, then by taking expectation of both sides and applying Fatou's lemma, and then Lemma 2.4, we get

$$\begin{aligned} E\|\mathcal{W}_i(\lambda) - \mathcal{W}_i^n(\lambda)\| &\leq \liminf_{N \rightarrow \infty} E\|\mathcal{W}_i^N(\lambda) - \mathcal{W}_i^n(\lambda)\| \\ &\leq \pi^{-d} E \int_I \sum_{m=n}^{\infty} |\mathcal{W}_i^{m+1}(z(t)) - \mathcal{W}_i^m(z(t))| dt \\ &\leq \pi^{-d} K\delta^n \int_I dt < \infty \rightarrow 0 \quad \text{as } n \rightarrow \infty. \end{aligned}$$

Finally, define

$$\Gamma = \bigcup_{x \in \Lambda_-} B(x, r).$$

Then, by a compactness argument, the uniform convergence of  $\{\mathcal{W}_i^n(\lambda)\}$  in the open balls in  $\Gamma$  implies the uniform convergence in any compact subset of  $\Gamma$ .  $\square$

### 3 Proofs of Theorems 4 and 5

Let  $1 \leq l \leq n$ , and  $i, j$  be fixed. We start with a decomposition which is similar to (2.2),

$$\mathcal{W}_{ij}^n(\lambda) = \sum_k \sum_s \frac{v_k(\lambda)}{v_i(\lambda)} \frac{1}{\rho(\lambda)^l} \exp(-\lambda' z_{ik;s}^l) \mathcal{W}_{kj}^{n-l}(\lambda|l, s), \quad (3.1)$$

where, given  $\mathcal{F}_l$ ,  $\mathcal{W}_{kj}^{n-l}(\lambda|l, s)$  are independent copies of  $\mathcal{W}_{kj}^{n-l}(\lambda)$  as  $s$  varies. Hence

$$E[\mathcal{W}_{ij}^n(\lambda) | \mathcal{F}_l] = \frac{v_j(\lambda)}{v_i(\lambda)\rho(\lambda)^n} \sum_k \sum_s \exp(-\lambda' z_{ik;s}^l) m_{kj}^{n-l}(\lambda). \quad (3.2)$$

Therefore, given any sample path of the process,  $E[\mathcal{W}_{ij}^n(\lambda) | \mathcal{F}_l]$  is an analytic function in  $\lambda \in \Omega$ .

For all  $\lambda \in \Omega$  define

$$g_{ij}^n(\lambda | l) = \mathcal{W}_{ij}^n(\lambda) - E[\mathcal{W}_{ij}^n(\lambda) | \mathcal{F}_l],$$

and

$$h_{ij}^n(\lambda|l) = E[\mathcal{W}_{ij}^n(\lambda) | \mathcal{F}_l] - u_j(\lambda)v_j(\lambda)\mathcal{W}_i^l(\lambda).$$

The functions  $g_{ij}^n(\lambda)$  and  $h_{ij}^n(\lambda)$  are analytic functions in  $\lambda \in \Omega$ . Furthermore,

$$|\mathcal{W}_{ij}^n(\lambda) - u_j(\lambda)v_j(\lambda)\mathcal{W}_i(\lambda)| \leq |g_{ij}^n(\lambda|l)| + |h_{ij}^n(\lambda|l)| + |u_j(\lambda)v_j(\lambda)| |\mathcal{W}_i^l(\lambda) - \mathcal{W}_i(\lambda)|.$$

Hence  $|\mathcal{W}_{ij}^n(\lambda) - u_j(\lambda)v_j(\lambda)\mathcal{W}_i(\lambda)|^\alpha$  is less than or equal to

$$3(|g_{ij}^n(\lambda|l)|^\alpha + |h_{ij}^n(\lambda|l)|^\alpha + |u_j(\lambda)v_j(\lambda)|^\alpha |\mathcal{W}_i^l(\lambda) - \mathcal{W}_i(\lambda)|^\alpha). \quad (3.3)$$

The idea now is to let  $l = \lfloor n/2 \rfloor$ , where  $\lfloor x \rfloor$  is the greatest integer not exceeding  $x$ , and  $n$  tend to infinity. This motivates the lemmas we now give before returning to the main proof. It is in these results that we need the uniformity proved in Theorem 1(iii).

**Lemma 3.1** *Suppose  $\vartheta \in (\Lambda_\alpha)_-$ . Then there are positive constants  $c_1, c_2$  and  $c_3$ , depending on  $\alpha$ , a constant  $\gamma < 1$  and a neighbourhood  $B$  of  $\vartheta$  such that, for all  $\lambda \in B$ ,  $n = 1, 2, \dots$ , and all  $i, j$ ,*

- (i)  $|h_{ij}^n(\lambda|l)| \leq c_1 \gamma^{n-l} \nu \phi^l \mathcal{W}_i^l(\theta)$ ,
- (ii)  $E \sum_{l=0}^n |h_{ij}^n(\lambda|l)|^\alpha \leq c_2 \phi^{n\alpha}$ ,
- (iii)  $\sigma^\alpha(\mathcal{W}_{ij}^n(\lambda)) \leq c_3 \phi^{n\alpha}$ .

**Proof.** Using (3.2) and the definition of  $\mathcal{W}_i^l(\lambda)$ , at (1.4), we have

$$h_{ij}^n(\lambda|l) = \frac{v_j(\lambda)}{v_i(\lambda)\rho(\lambda)^l} \sum_k \sum_s \exp(-\lambda'z_{ik;s}^l) [\rho(\lambda)^{-(n-l)} m_{kj}^{n-l}(\lambda) - u_j(\lambda)v_k(\lambda)]. \quad (3.4)$$

Therefore, by Theorem 1(iii), there are constants  $c_1 > 0$ ,  $\gamma \in (0, 1)$ , and a neighbourhood  $B$  of  $\vartheta$  such that, for all  $\lambda \in B$  and all  $n$ ,

$$\begin{aligned} |h_{ij}^n(\lambda|l)| &\leq c_1 \nu \phi^l \gamma^{n-l} \sum_k \sum_s \left[ \frac{v_k(\theta)}{v_i(\theta)} \rho(\theta)^{-l} \exp(-\theta'z_{ik;s}^l) \right] \\ &= c_1 \nu \phi^l \gamma^{n-l} \mathcal{W}_i^l(\theta), \end{aligned} \quad (3.5)$$

as required. This and Lemma 2.3(iv) combine, after making  $B$  smaller if necessary, to show that, for some constant  $c'_1$ ,  $E|h_{ij}^n(\lambda|l)|^\alpha \leq c'_1 \nu^\alpha (\gamma/\phi)^{(n-l)\alpha} \phi^{n\alpha}$ . Since  $\gamma < 1$  and  $\phi \geq 1$ ,  $(\gamma/\phi) < 1$ , for all  $\lambda \in B$ . Thus

$$E \sum_{l=0}^n |h_{ij}^n(\lambda|l)|^\alpha \leq c'_1 \nu^\alpha \frac{1}{1 - (\gamma/\phi)^\alpha} \phi^{n\alpha}.$$

The continuity of the functions  $\nu$  and  $\phi$  now implies (ii). Finally, by Lemma 2.2,

$$\begin{aligned} \sigma^\alpha(\mathcal{W}_{ij}^n(\lambda)) &\leq C \sum_{l=1}^n E |E[\mathcal{W}_{ij}^n(\lambda)|\mathcal{F}_l] - E[\mathcal{W}_{ij}^n(\lambda)|\mathcal{F}_{l-1}]|^\alpha \\ &\leq C' \left( |u_j(\lambda)v_j(\lambda)|^\alpha \sum_{l=1}^n E |\mathcal{W}_i^l(\lambda) - \mathcal{W}_i^{l-1}(\lambda)|^\alpha + 2E \sum_{l=0}^n |h_{ij}^n(\lambda|l)|^\alpha \right). \end{aligned}$$

For  $\delta$  small enough, by suitable bounding of the continuous functions involved in Lemma 2.3(i), for all  $\lambda \in B(\vartheta, \delta)$

$$\sum_{l=1}^{\infty} E |\mathcal{W}_i^l(\lambda) - \mathcal{W}_i^{l-1}(\lambda)|^\alpha \leq K_1 < \infty,$$

and so, using this and part (ii),

$$\sigma^\alpha(\mathcal{W}_{ij}^n(\lambda)) \leq C' (|u_j(\lambda)v_j(\lambda)|^\alpha K_1 + 2c_2 \phi^{n\alpha}).$$

Now, the boundedness of  $u_j(\lambda)v_j(\lambda)$  on  $B$  and the fact that  $\phi \geq 1$  combine to prove (iii).  $\square$

**Lemma 3.2** *Suppose  $\alpha \in (1, 2]$  and  $\vartheta \in (\Lambda_\alpha)_-$ . Then there is a  $B = B(\vartheta, \delta) \subset \Lambda_\alpha$ , such that, for  $l = \lfloor n/2 \rfloor$ ,*

- (i) *for all  $\lambda \in B$ ,  $h_{ij}^n(\lambda|l) \rightarrow 0$  as  $n \rightarrow \infty$ , almost surely and in  $\alpha$ th mean;*
- (ii)  *$\sup \{|h_{ij}^n(\lambda|l)| : \lambda \in B\} \rightarrow 0$  as  $n \rightarrow \infty$ , almost surely and in mean.*

**Proof.** Let  $B = B(\vartheta, \delta)$  be a neighbourhood of  $\vartheta$  over which (3.5) holds. Take  $\delta$  smaller if necessary, so that  $\phi\gamma < 1$  throughout  $B$ . Applying Theorems 2 and 3 to  $\mathcal{W}_i^l(\theta)$  and simple bounding now give all the claimed results.  $\square$

**Lemma 3.3** Suppose  $\alpha \in (1, 2]$  and  $\vartheta \in (\Lambda_\alpha)_-$ . Then there is a neighbourhood of  $\vartheta$ , say  $B = B(\vartheta, r) \subset \Lambda_\alpha$ , such that, for  $l = [n/2]$ ,

- (i) for all  $\lambda \in B$ ,  $g_{ij}^n(\lambda|l) \rightarrow 0$  as  $n \rightarrow \infty$ , almost surely and in  $\alpha$ th mean;
- (ii)  $\sup \{|g_{ij}^n(\lambda|l)| : \lambda \in B\} \rightarrow 0$  as  $n \rightarrow \infty$ , almost surely and in mean.

**Proof.** Note first that, using (3.1) and Lemma 2.2,

$$\sigma^\alpha(\mathcal{W}_{ij}^n(\lambda) | \mathcal{F}_l) \leq C\bar{v}(\lambda)^\alpha |\rho(\lambda)|^{-l\alpha} \sum_k \sum_s \exp(-\alpha\theta^l z_{ik;s}^l) \sigma^\alpha(\mathcal{W}_{kj}^{n-l}(\lambda)). \quad (3.6)$$

Let  $B = B(\vartheta, \delta)$  be inside the neighbourhood of  $\vartheta$  introduced in Lemma 3.1. Then for some positive constant  $c$  and for all  $\lambda \in B$ ,  $n = 1, 2, \dots$ , and all  $i, j$ , applying Lemma 3.1(iii) to the bound (3.6) gives

$$\sigma^\alpha(\mathcal{W}_{ij}^n(\lambda) | \mathcal{F}_l) \leq c\nu_1 \mathcal{W}_i^l(\alpha\theta) \phi^{\alpha(n-l)} \kappa^l. \quad (3.7)$$

Now, since  $E[\mathcal{W}_i^l(\alpha\theta)] = 1$ ,

$$E[|g_{ij}^n(\lambda|l)|^\alpha] = E[\sigma^\alpha(\mathcal{W}_{ij}^n(\lambda) | \mathcal{F}_l)] \leq c\nu_1 \phi^{\alpha(n-l)} \kappa^l.$$

Choose  $\delta$  smaller if necessary so that  $\phi^\alpha \kappa < 1$  for all  $\lambda \in B$ . Then the right hand side of the above inequality converges to zero geometrically quickly, as  $n \rightarrow \infty$ , when  $l = [n/2]$ . This implies that  $g_{ij}^n(\lambda|l) \rightarrow 0$  in  $\alpha$ th mean and almost surely.

Jensen's inequality and (3.7) give

$$E|g_{ij}^n(\lambda|l)| \leq E[\sigma^\alpha(\mathcal{W}_{ij}^n(\lambda) | \mathcal{F}_l)]^{1/\alpha} \leq c^{1/\alpha} \nu_1^{1/\alpha} \phi^{n-l} \kappa^{l/\alpha}.$$

For  $l = [n/2]$ ,  $\phi^{n-l} \kappa^{l/\alpha}$  is asymptotically equivalent to  $(\phi \kappa^{1/\alpha})^l$ , and  $\phi \kappa^{1/\alpha} \leq \gamma_1 < 1$  in  $B$ . Hence there is a  $C < \infty$  such that for all  $\lambda \in B$ ,  $n = 1, 2, \dots$ ,  $l = [n/2]$ , and all  $i, j$ ,  $E|g_{ij}^n(\lambda|l)| \leq C\gamma_1^l$ . Let  $2r < \delta$  and apply Lemma 2.4 to the analytic functions  $g_{ij}^n(\lambda|l)$  to get

$$\sup_{\lambda \in B(\vartheta, r)} |g_{ij}^n(\lambda|l)| \leq \pi^{-d} \int_I |g_{ij}^n(z(t)|l)| dt,$$

where  $z(t) \in D(\vartheta, 2r)$ . The expectation of the right hand side here goes to zero geometrically quickly. Hence  $g_{ij}^n(\lambda|l)$  converges to zero uniformly in  $\lambda \in B(\vartheta, r)$  almost surely and in mean.  $\square$

**Proof of Theorems 4 and 5.** In (3.3), let  $l = [n/2]$  and then let  $n \rightarrow \infty$ . In a suitable neighbourhood of  $\vartheta \in (\Lambda_\alpha)_-$ , the almost sure and  $\alpha$ th mean convergence of  $\{\mathcal{W}_i^n(\lambda) - \mathcal{W}_i(\lambda)\}$ ,  $g_{ij}^n(\lambda|l)$  and  $h_{ij}^n(\lambda|l)$  are contained in Theorem 2, Lemma 3.3(i) and Lemma 3.2(i) respectively. Now take the union of these neighbourhoods as  $\Gamma_1$ . This proves Theorem 4.

Similarly, the almost sure and mean uniform convergence of  $\{\mathcal{W}_i^n(\lambda) - \mathcal{W}_i(\lambda)\}$ ,  $g_{ij}^n(\lambda|l)$  and  $h_{ij}^n(\lambda|l)$  are contained in Theorem 3, Lemma 3.3(ii) and Lemma 3.2(ii) respectively. A union of suitable neighbourhoods now provides  $\Gamma_2$ , proving Theorem 5.  $\square$

## 4 Proof of Theorem 6

**Lemma 4.1** *If the branching random walk is strongly non-lattice then for any compact set  $\mathcal{K} \subset \{\theta \in \mathbb{R}^d : \sup_{i,j} m_{ij}(\theta) < \infty\}$  and all  $a \in (0, 1)$  there is an  $\varepsilon < 1$  such that*

$$\varepsilon^{-n} \sup_{i,j} \sup_{\theta \in \mathcal{K}} \sup_{a \leq |\eta| \leq a^{-1}} \left| \frac{m_{ij}^n(\theta + \mathbf{i}\eta)}{\rho^n(\theta)} \right| \rightarrow 0 \quad \text{as } n \rightarrow \infty. \quad (4.8)$$

**Proof.** Let  $(k, l)$  be such that (1.7) holds. As a function of  $\eta$ ,  $m_{kl}(\theta + \mathbf{i}\eta)/m_{kl}(\theta)$  is the characteristic function of the probability measure  $m_{kl}(\theta)^{-1}e^{-\theta \cdot x} \mu_{kl}(dx)$  and (1.7) implies that this measure is strongly non-lattice. Then, using this and continuity,

$$\tilde{\varepsilon} = \sup_{\theta \in \mathcal{K}} \sup_{a \leq |\eta| \leq a^{-1}} \left| \frac{m_{kl}(\theta + \mathbf{i}\eta)}{m_{kl}(\theta)} \right| < 1.$$

Let  $M^\dagger(\theta)$  be the matrix with  $(k, l)$ th entry  $\tilde{\varepsilon} m_{kl}(\theta)$  and with other entries the same as those in  $M(\theta)$  and let  $(M^\dagger(\theta)^n)_{ij}$  be the  $(i, j)$ th entry of its  $n$ th power. Let  $\rho^\dagger(\theta)$  be the eigenvalue of  $M^\dagger(\theta)$  of maximum modulus, with right eigenvector  $v^\dagger(\theta)$ . Then  $\rho^\dagger(\theta) < \rho(\theta)$  (see Seneta (1973, Theorem 1.1)). Furthermore, since Theorem 1 applies to  $M^\dagger(\theta)$ , there is an  $\varepsilon < 1$  such that  $\rho^\dagger(\theta) < \varepsilon \rho(\theta)$  for all  $\theta \in \mathcal{K}$ , and

$$\sup_{\theta \in \mathcal{K}} \max_{i,j} \{v_i^\dagger(\theta)/v_j^\dagger(\theta)\} = c < \infty.$$

Now,

$$\sup_{a \leq |\eta| \leq a^{-1}} |(M(\theta + \mathbf{i}\eta)^n)_{ij}| \leq (M^\dagger(\theta)^n)_{ij} \leq c(\varepsilon \rho(\theta))^n,$$

proving (4.8). □

### Proof of Theorem 6.

It is sufficient to prove (1.8) with the compact set  $\mathcal{K}$  being some closed ball in  $\mathbb{R}^d$  centred on  $\vartheta \in \Lambda_-$ , since a simple covering argument extends the result to general  $\mathcal{K}$ . The proof starts by introducing various bounds which lead to the appropriate ball to consider. Recall that  $S(x, r)$  is a ball in  $\mathbb{R}^d$  centred on  $x$  with radius  $r$  and  $\overline{S}(x, r)$  is its closure.

Choose  $\vartheta \in \Lambda_-$ , then  $\vartheta \in \Lambda_\alpha$  for some  $\alpha$ , which is now fixed. Now take  $\delta_1$  such that  $S(\vartheta, 2\delta_1) \subset \Lambda_\alpha$  and  $S_1 = \overline{S}(\vartheta, \delta_1)$ . Let  $\varepsilon_1$  be an  $\varepsilon$  such that (4.8) holds when  $\mathcal{K} = \{\theta \in S_1\}$  and  $a = \varepsilon/2$ . Recall that  $\rho(\alpha\theta)^{1/\alpha}/\rho(\theta) < 1$  for  $\theta \in \Lambda_\alpha$ . Let

$$\varepsilon_2 = \max \left\{ \varepsilon_1, \sup_{\theta \in \overline{S}_1} \frac{\rho(\alpha\theta)^{1/\alpha}}{\rho(\theta)} \right\}.$$

Now let  $S_2 = \overline{S}(\vartheta, 2\delta_2)$  with  $2\delta_2 < \delta_1$  and let

$$\bar{\rho} = \sup_{\theta \in S_2} \rho(\theta) \quad \text{and} \quad \underline{\rho} = \inf_{\theta \in S_2} \rho(\theta).$$

Take  $\delta_2$  small enough that

$$\varepsilon_3 = \varepsilon_2 \frac{\sup_{\theta \in S_2} \rho(\theta)}{\inf_{\theta \in S_2} \rho(\theta)} = \varepsilon_2 \frac{\bar{\rho}}{\underline{\rho}} < 1. \quad (4.9)$$

Note that, for  $\theta \in S_2$ ,

$$\rho(\alpha\theta)^{1/\alpha} \leq \varepsilon_2 \rho(\theta) \leq \varepsilon_2 \bar{\rho}. \quad (4.10)$$

These preliminaries were to obtain a suitably small neighbourhood of  $\vartheta$ , where various inequalities hold. Having done this, let  $S = \bar{S}(\vartheta, \delta_2) \subset S_2 = \bar{S}(\vartheta, 2\delta_2)$ . Define the two regions

$$R = \{\theta + \mathbf{i}\eta : \theta \in S, \varepsilon \leq \eta \leq \varepsilon^{-1}\} \quad \text{and} \quad R_2 = \{\theta + \mathbf{i}\eta : \theta \in S_2, \varepsilon/2 \leq \eta \leq 2\varepsilon^{-1}\}.$$

We will prove (1.8) for  $\theta \in S$ , that is for  $\lambda \in R$ . Clearly

$$\sup_{\lambda \in R} |\mathcal{M}_{ij}^n(\lambda)| \leq \sup_{\lambda \in R} |\mathcal{M}_{ij}^n(\lambda) - m_{ij}^n(\lambda)| + \sup_{\lambda \in R} |m_{ij}^n(\lambda)|. \quad (4.11)$$

Furthermore,

$$\sup_{\lambda \in R} |m_{ij}^n(\lambda)| \leq \sup_{\lambda \in R_2} |m_{ij}^n(\lambda)| = \sup_{\lambda \in R_2} \left| \frac{m_{ij}^n(\lambda)}{\rho(\theta)^n} \rho(\theta)^n \right| \leq (\varepsilon_1 \bar{\rho})^n \leq (\varepsilon_2 \bar{\rho})^n. \quad (4.12)$$

Using Lemma 2.2,

$$E |\mathcal{M}_{ij}^n(\lambda) - m_{ij}^n(\lambda)|^\alpha \leq C \sum_{l=0}^{n-1} E |E[\mathcal{M}_{ij}^n(\lambda)|\mathcal{F}_{l+1}] - E[\mathcal{M}_{ij}^n(\lambda)|\mathcal{F}_l]|^\alpha \quad (4.13)$$

and splitting on the  $l$ th generation shows that

$$\begin{aligned} E[\mathcal{M}_{ij}^n(\lambda)|\mathcal{F}_l] &= \sum_k \left( \sum_s \exp(-\lambda' z_{ik;s}^l) \right) m_{kj}^{n-l}(\lambda) \\ &= \sum_k \left( \int \exp(-\lambda' z) Z_{ik}^l(dz) \right) m_{kj}^{n-l}(\lambda). \end{aligned} \quad (4.14)$$

Following (2.1), let

$$\mathcal{M}_{kj}^{n-l}(\lambda|l, s) = \int_{\mathbb{R}^d} e^{-\lambda' x} Z_{kj}^n(dx|l, s),$$

then

$$\int \exp(-\lambda' z) Z_{ik}^{l+1}(dz) = \sum_h \sum_s \exp(-\lambda' z_{ih;s}^l) \mathcal{M}_{hk}^1(\lambda|l, s). \quad (4.15)$$

Let  $\Delta_l = E[\mathcal{M}_{ij}^n(\lambda)|\mathcal{F}_{l+1}] - E[\mathcal{M}_{ij}^n(\lambda)|\mathcal{F}_l]$ . Then, using (4.14) and (4.15)

$$\Delta_l = \sum_k \left( \sum_h \sum_s \exp(-\lambda' z_{ih;s}^l) (\mathcal{M}_{hk}^1(\lambda|l, s) - m_{hk}(\lambda)) \right) m_{kj}^{n-l-1}(\lambda)$$

and, given  $\mathcal{F}_l$ ,  $\{\mathcal{M}_{hk}^1(\lambda|l, s) - m_{hk}(\lambda) : s\}$  are independent variables with zero mean. We need a bound on the moments of these variables. Let

$$c_1 = 4 \left( \max_{i,j} \sup_{\theta \in S_2} E |\mathcal{M}_{ij}^1(\theta)|^\alpha \right)^{1/\alpha},$$

which is finite because  $S_2 \subset \Lambda_\alpha$ . Then simple bounding gives

$$\sup \{ E |\mathcal{M}_{ij}^1(\lambda) - m_{ij}(\lambda)|^\alpha : \lambda = \theta + \mathbf{i}\eta, \theta \in S_2, i, j \} \leq (c_1)^\alpha.$$

Let

$$c_2 = \max_{i,j} \sup_{\theta \in S_2} \frac{v_j(\alpha\theta)}{v_i(\alpha\theta)},$$

which is finite, by Theorem 1(ii), because  $S_2 \subset \Lambda_\alpha$ . Then, using that  $m_{ij}^n(\alpha\theta)v_j(\alpha\theta) \leq \rho(\alpha\theta)^n v_i(\alpha\theta)$  and (4.10), for  $\theta \in S_2$

$$m_{ij}^n(\alpha\theta) \leq c_2 (\varepsilon_2 \bar{\rho})^{n\alpha}.$$

Now, applying Lemma 2.2 again, (4.12) and two bounds just given, for any  $\lambda \in R_2$

$$\begin{aligned} E|\Delta_{l-1}|^\alpha &\leq C \sum_k \left( \sum_h m_{ih}^{l-1}(\alpha\theta) E |\mathcal{M}_{hk}^1(\lambda) - m_{hk}(\lambda)|^\alpha \right) |m_{kj}^{n-l}(\lambda)|^\alpha \\ &\leq Cp^2 c_2 (\varepsilon_2 \bar{\rho})^{(l-1)\alpha} (c_1)^\alpha (\varepsilon_2 \bar{\rho})^{(n-l)\alpha} \\ &\leq c_3 (\varepsilon_2 \bar{\rho})^{n\alpha}, \end{aligned}$$

where  $c_3$  is independent of  $\theta \in S_2$  and  $l$ . Hence (4.13) gives

$$E |\mathcal{M}_{ij}^n(\lambda) - m_{ij}^n(\lambda)|^\alpha \leq nc_3 (\varepsilon_2 \bar{\rho})^{n\alpha}$$

for any  $\lambda \in R_2$ .

The region  $R$  can be covered with a finite number of polydiscs such that when their radii are doubled they still lie inside  $R_2$ . Hence, using (4.13), the bound just obtained and Lemma 2.4, for a suitable constant  $c_4$ ,

$$E \sup_{\lambda \in R} |\mathcal{M}_{ij}^n(\lambda) - m_{ij}^n(\lambda)| \leq c_4 (\varepsilon_2 \bar{\rho})^n n^{1/\alpha}.$$

Combining this with (4.11) and (4.12)

$$E \sup_{\lambda \in R} |\mathcal{M}_{ij}^n(\lambda)| \leq (c_4 n^{1/\alpha} + c_1) (\varepsilon_2 \bar{\rho})^n$$

and so, using (4.9),

$$E \sup_{\lambda \in R} \left| \frac{\mathcal{M}_{ij}^n(\lambda)}{\rho(\theta)^n} \right| \leq \frac{1}{\rho^n} E \sup_{\lambda \in R} |\mathcal{M}_{ij}^n(\lambda)| \leq (c_4 n^{1/\alpha} + c_1) \varepsilon_3^n.$$

Hence (1.8) holds for any  $\varepsilon \in (\varepsilon_3, 1)$ . □

## 5 Approximation of measures

This section contains the preparatory work on saddlepoint approximations for the proof of Theorem 7. The idea is to develop explicit estimates that apply to a particular measure through only a few of its attributes. Most treatments of these matters, with the application to variables that are, or look like, sums of independent identically distributed random variables in mind, bring  $n$  into the picture sooner than we do. The treatment draws on ideas from Stone (1967), von Bahr (1967), Bhattacharya (1972, 1977), Chaganty and Sethuraman (1993), and Jensen (1995).



Recall that  $S(x, \epsilon) = \{y : |x - y| < \epsilon\}$ . Let  $\Gamma$  be a probability measure with  $\gamma \equiv \Gamma(S(0, 1)) > 1/2$  and characteristic function  $\int e^{itu} \Gamma(du)$  vanishing outside  $\{t : |t| \geq \zeta\}$ . Let  $\Gamma_\epsilon$  be the measure given by  $\Gamma_\epsilon(A) = \Gamma(\{x : \epsilon x \in A\})$ , so that its characteristic function is zero outside  $\{t : |t| \geq \epsilon^{-1}\zeta\}$ . Let  $\nu$  and  $\mu$  be probability measures,  $b_\epsilon$  be the supremum of the modulus of the density of  $(\nu - \mu) * \Gamma_\epsilon$  and  $q$  be the supremum of the density of  $\mu$ . For any  $f$ , let

$$\omega_f(x, \epsilon) = \sup\{|f(y) - f(z)| : y, z \in S(x, \epsilon)\}$$

and

$$\bar{f}^\epsilon(x) = \sup\{|f(y)| : y \in S(x, \epsilon)\}.$$

The next result is contained in, and easily derived from, Lemma 2.2 in Bhattacharya (1972) and the remark following it.

**Lemma 5.1** *For any real-valued bounded Borel-measurable  $f$*

$$\sup_{u \in \mathbb{R}^d} \left| \int f(x)(\nu - \mu)(dx + u) \right| \leq \frac{1}{2\gamma - 1} \left[ b_\epsilon \int \bar{f}^\epsilon(x) dx + q \int \omega_f(x, 2\epsilon) dx \right].$$

The idea now is to make this bound simpler by imposing suitable smoothness conditions on  $f$ . The set  $\mathfrak{G}'(G)$  is made up of those  $f \in \mathfrak{D}$  for which there is a bounded decreasing function  $g$  with  $\int g(r)(r+1)^{d-1} dr = G$  and

$$\sup\{|f(x)|, |f'(x)| : |x| \geq (r-1)\} \leq g(r).$$

Note that  $\mathfrak{G}(G)$ , introduced to state Theorem 7, is defined in the same way, except that  $\int g(r)(r+1)^{d+1} dr = G$ ; therefore,  $\mathfrak{G}(G) \subset \mathfrak{G}'(G)$ . Also, let  $\mathfrak{C}$  be the convex sets in  $\mathbb{R}^d$ .

**Lemma 5.2** *For a constant  $\Delta$ , depending only on the dimension, and  $\epsilon < 1$*

$$\sup \left\{ \left| \int_C f(x)(\nu - \mu)(dx + u) \right| : u \in \mathbb{R}^d, C \in \mathfrak{C}, f \in \mathfrak{G}'(G) \right\} \leq G\Delta [b_\epsilon + \epsilon q].$$

**Proof.**

For a set  $A \subset \mathbb{R}^d$ , let  $A^\epsilon = \{y : |y - A| < \epsilon\}$  and  $\partial A$  be the boundary of  $A$ . Let  $\mathfrak{S}$  be the surface area of the unit ball. Then, for  $C \in \mathfrak{C}$ , from (15) in von Bahr (1967),

$$\int I(x \in (\partial C)^\epsilon) g(|x|) dx \leq 2\mathfrak{S}G\epsilon.$$

Take  $\epsilon < 1$ . Let  $h(x) = f(x)I(x \in C)$  with  $f \in \mathfrak{G}'(G)$ . If  $x \in (\partial C)^\epsilon$  then  $\omega_h(x, \epsilon) \leq 2 \sup\{|f(y)| : y \in S(x, \epsilon)\} \leq 2g(|x|)$ . If  $x \in S(x, \epsilon) \subset C$  then, using Taylor's theorem,  $\omega_h(x, \epsilon) = \omega_f(x, \epsilon) \leq 2\epsilon g(|x|)$ . Hence

$$\omega_h(x, \epsilon) \leq 2g(|x|) [I(x \in (\partial C)^\epsilon) + \epsilon I(x \in C)]$$

and so

$$\int \omega_h(x, \epsilon) dx \leq 2 \int [I(x \in (\partial C)^\epsilon) + \epsilon I(x \in C)] g(|x|) dx \leq 6\epsilon \mathfrak{S}G.$$

Also

$$\int \bar{h}^\epsilon(x) dx \leq \int \bar{f}^\epsilon(x) dx \leq \int g(|x|) dx = \mathfrak{S}G.$$

Substituting these into the bound in Lemma (5.1) gives the result.  $\square$

Recall that for  $A \subset \mathbb{C}^d$ ,  $A_-$  is its intersection with  $\mathbb{R}^d$ . Let  $\mathcal{K}$  be compact in  $\mathbb{R}^d$ ,  $\tau < 1$  and

$$\mathcal{K}_\tau = \{\theta + \lambda \in \mathbb{C}^d : \theta \in \mathcal{K}, |\lambda| \leq \tau\}.$$

Now, let  $B$  be an open set containing  $\mathcal{K}_\tau$  with  $B_-$  convex. For  $Z$  a measure on  $\mathbb{R}^d$  let

$$\widehat{Z}(\lambda) = \int e^{-\lambda x} Z(dx).$$

Suppose that  $\log \widehat{Z}$  is analytic on  $B$  and strictly convex on  $B_-$ . Let  $-m(\theta)$  be the vector of its first derivatives and  $\Sigma(\theta)$  be the matrix of its second derivatives on  $B_-$ . Then  $\Sigma(\theta)$  is positive-definite because  $\log \widehat{Z}$  is strictly convex. Let  $\mathbf{u}$  be a (finite) bound on the modulus of all derivatives of  $\log \widehat{Z}$  up to and including order 3 over  $\mathcal{K}_\tau$ . Let  $\mathbf{c}$  be a lower bound on the smallest eigenvalue of  $\Sigma(\theta)$  as  $\theta$  varies through  $\mathcal{K}$  and let  $\mathbf{r} = \mathbf{u}/\mathbf{c}$ . Since  $\log \widehat{Z}$  is strictly convex and analytic on  $B_-$ , we can take  $\mathbf{c} > 0$ . Also, let

$$\varepsilon(a) = a^{-d} \sup \left\{ \left| \frac{\widehat{Z}(\theta + \mathbf{i}t)}{\widehat{Z}(\theta)} \right| : \theta \in \mathcal{K}, a \leq |t| < 1/a \right\}.$$

Fix  $\vartheta \in \mathcal{K}$ . Let  $\nu$  be the probability measure given by

$$\nu(A) = \frac{\int I(x \in A) e^{-\vartheta' x} Z(dx)}{\int e^{-\vartheta' y} Z(dy)},$$

with mean  $m$  ( $= m(\vartheta)$ ) and covariance matrix  $\Sigma$  ( $= \Sigma(\vartheta)$ ). Let  $\widehat{\nu}(\lambda)$  be the Laplace transform of  $\nu$ , so that  $\widehat{\nu}(\lambda) = \widehat{Z}(\vartheta + \lambda)/\widehat{Z}(\vartheta)$ . Let  $\mu$  be the normal distribution with mean  $m$  and covariance matrix  $\Sigma$ . The idea is to use  $\mu$  to approximate  $\nu$ , and hence  $Z$ , through Lemma 5.2, in a way that is suitably uniform. Note that  $\nu$  and  $\mu$  both depend on  $\vartheta$ ; we use  $\nu_\vartheta$  and  $\mu_\vartheta$  in the next definition to emphasise this. Let

$$A(Z) = \sup \left\{ \left| \int_C f(x) (\nu_\vartheta - \mu_\vartheta)(dx + u) \right| : u \in \mathbb{R}^d, f \in \mathfrak{G}'(G), C \in \mathfrak{C}, \vartheta \in \mathcal{K} \right\}.$$

Obviously by bounding  $A(Z)$  we allow  $f$  to be shifted arbitrarily,  $f$  to vary within  $\mathfrak{G}'(G)$  and integration over an arbitrary convex set.

**Theorem 9** For  $\delta \leq \min\{\tau, e^{-d}/(4\mathbf{r})\}$ ,  $\epsilon \leq \min\{1, \zeta/\delta, \zeta\delta\}$  and a constant  $\Delta'$ , depending only on the dimension,

$$A(Z) \leq G\Delta' \left[ \frac{\mathbf{r}}{\mathbf{c}^{(d+1)/2}} + \varepsilon(\epsilon/\zeta) + \frac{(e^{-\delta^2\epsilon/4} + \epsilon)}{\mathbf{c}^{d/2}} \right].$$

**Proof.** Let  $q$  be a bound on the density of  $\mu$  and let  $b_\epsilon$  be a bound on the modulus of the density of  $(\nu - \mu) * \Gamma_\epsilon$ . Note first that  $q \leq (2\pi\mathbf{c})^{-d/2}$ , which is independent of  $\vartheta$ .

For  $|\lambda| \leq \tau$ , let

$$\begin{aligned} \psi(\lambda) &= \log \widehat{\nu}(\lambda) - m'\lambda + \frac{1}{2}\lambda'\Sigma\lambda \\ &= \log \widehat{Z}(\vartheta + \lambda) - \log \widehat{Z}(\vartheta) - m'\lambda + \frac{1}{2}\lambda'\Sigma\lambda. \end{aligned}$$

Then, by arrangement,  $\psi$  is analytic on  $|\lambda| \leq \tau$  with  $\psi(0) = 0$  and all its first and second derivatives vanishing at 0.

Using Taylor's theorem and the analyticity of  $\log \widehat{Z}$  on  $B$ , for  $|\lambda| \leq \tau$ ,

$$|\psi(\lambda)| \leq \frac{d^3}{3!} \mathbf{u} |\lambda|^3 \leq e^d \mathbf{u} |\lambda|^3$$

and then, for  $|\lambda| \leq \delta$ ,

$$|\psi(\lambda)| \leq e^d \mathbf{u} |\lambda|^3 \leq \left(4e^d \frac{\mathbf{u}}{\mathbf{c}} \delta\right) \frac{\mathbf{c}}{4} |\lambda|^2 \leq \frac{\mathbf{c}}{4} |\lambda|^2.$$

Using these two inequalities gives

$$|\exp\{\psi(\lambda)\} - 1| \leq e^d \mathbf{u} |\lambda|^3 \exp\left\{\frac{\mathbf{c}}{4} |\lambda|^2\right\} \quad \text{for } |\lambda| \leq \delta.$$

The key point is that right hand side here does not depend on  $\vartheta$ . Using this bound,

$$\begin{aligned} \int_{|t| < \delta} \left| \widehat{\nu}(\mathbf{i}t) - e^{\mathbf{i}m t} e^{-t\Sigma t/2} \right| dt &\leq \int_{|t| < \delta} e^{-t\Sigma t/2} |e^{\psi(\mathbf{i}t)} - 1| dt \\ &\leq e^d \mathbf{u} \int_{|t| < \delta} |t|^3 e^{\mathbf{c}|t|^2/4} e^{-t\Sigma t/2} dt \\ &\leq e^d \mathbf{u} \int_{|t| < \delta} |t|^3 e^{-|t|^2 \mathbf{c}/4} dt \\ &\leq e^d \mathbf{u} \frac{1}{\mathbf{c}^{(d+3)/2}} \int |z|^3 e^{-|z|^2/4} dz \\ &\leq \frac{\Delta_1 \mathbf{r}}{\mathbf{c}^{(d+1)/2}}, \end{aligned}$$

where  $\Delta_1$  depends only on the dimension,  $d$ . Then

$$\begin{aligned} (2\pi)^d b_\epsilon &\leq \int_{|t| < \zeta/\epsilon} \left| \widehat{\nu}(\mathbf{i}t) - e^{\mathbf{i}m t} e^{-t\Sigma t/2} \right| dt \\ &\leq \int_{|t| < \delta} \left| \widehat{\nu}(\mathbf{i}t) - e^{\mathbf{i}m t} e^{-t\Sigma t/2} \right| dt + \int_{\delta \leq |t| < \zeta/\epsilon} |\widehat{\nu}(\mathbf{i}t)| dt + \int_{\delta \leq |t|} e^{-t\Sigma t/2} dt \\ &\leq \frac{\Delta_1 \mathbf{r}}{\mathbf{c}^{(d+1)/2}} + \epsilon (\epsilon/\zeta) \frac{\mathfrak{G}}{d} + e^{-\delta^2 \mathbf{c}/4} \frac{(4\pi)^{d/2}}{\mathbf{c}^{d/2}}. \end{aligned}$$

Substituting these estimates for  $b_\epsilon$  and  $q$  into Lemma 5.2 gives the result.  $\square$

For the main result, we also need to approximate the normal distribution  $\mu$ . To formulate the theorem, let

$$B(v) = \sup \left\{ \left| \int_C f(x) \mu(dx + m - v) - \frac{\int_C f(x) dx}{\sqrt{(2\pi)^d \det[\Sigma]}} \right| : f \in \mathfrak{G}(G), C \in \mathfrak{C}, \vartheta \in \mathcal{K} \right\}.$$

Recall that  $\mu$ ,  $m$  and  $\Sigma$  each depend on  $\vartheta$ .

**Theorem 10** *For a constant  $\Delta''$ , depending only on the dimension,*

$$B(v) \leq \Delta'' \frac{1 + |v|^2}{\mathbf{c}^{(d+2)/2}} G.$$

**Proof.** Temporarily, let  $h(x) = f(x)I(x \in C)$  and, for a fixed  $\vartheta \in \mathcal{K}$ ,

$$B_* = \left| \sqrt{(2\pi)^d \det[\Sigma]} \int h(x) \mu(dx + m - v) - \int h(x) dx \right|.$$

Then

$$\begin{aligned} B_* &= \left| \int h(x+v) (\exp(-x'\Sigma^{-1}x/2) - 1) dx \right| \\ &\leq \frac{1}{2} \int |f(x+v)| (x'\Sigma^{-1}x) dx \\ &\leq \frac{1}{2\mathbf{c}} \int |f(x)| |x-v|^2 dx \\ &\leq \frac{1}{2\mathbf{c}} \int g(|x|) |x|^2 dx + \frac{|v|^2}{2\mathbf{c}} \int g(|x|) dx \\ &\leq \frac{1+|v|^2}{2\mathbf{c}} \mathfrak{S}G, \end{aligned}$$

and so

$$\frac{B_*}{\sqrt{(2\pi)^d \det[\Sigma]}} \leq \frac{B_*}{\sqrt{(2\pi)^d \mathbf{c}^d}} \leq \frac{\mathfrak{S}}{2\sqrt{(2\pi)^d}} \frac{1+|v|^2}{\mathbf{c}^{(d+2)/2}} G,$$

as required.  $\square$

It is worth, very briefly, relating these results to those in Chaganty and Sethuraman (1993). Temporarily following the notation used there, let  $T_n$  be a univariate random variable with moment generating function  $\exp(n\psi_n(z))$ . Suppose that  $\psi_n$  is analytic, and bounded in  $n$  and  $z$ , on  $\Omega$ , and has a second derivative bounded below by  $\alpha$  there. Let  $\{\tau_n\}$  be a positive bounded sequence inside  $\Omega_-$  and suppose that, for any  $a > 0$ ,

$$\varepsilon_n(a) = a \sup \left\{ \left| \frac{\exp(n\psi_n(\tau_n + \mathbf{i}t))}{\exp(n\psi_n(\tau_n))} \right| : a \leq |t| < 1/a \right\} = o(n^{-1/2}).$$

These are the conditions of Theorem 3.3 in Chaganty and Sethuraman (1993). Under these conditions we can take  $\mathbf{c} = n\alpha$ ,  $\mathbf{u} = nU$  and  $\mathbf{r}$  is bounded.

Take  $f(x)$  to be zero on  $(-\infty, -1]$ ,  $e^{-\tau x}$  on  $[0, \infty)$  and differentiable, with a bounded derivative, in between and at  $-1$  and  $0$ . For  $\tau \in (0, B]$  it is routine calculus to show that there is a constant  $C$  such that  $f$  is in both  $\mathfrak{G}'(C/\tau)$  and  $\mathfrak{G}(C/\tau^3)$ . Apply Theorems 9 and 10 with this  $f$ , the convex set  $C$  being  $(0, \infty)$  and  $\mathcal{K}$  being the point set  $\{\tau_n\}$  when estimating the  $n$ th distribution, to get

$$\frac{P[(T_n - n\psi'_n(\tau_n)) \in (0, \infty)]}{\exp(n\psi_n(\tau_n) - n\tau_n\psi'(\tau_n))} - \frac{1}{\tau_n \sqrt{2\pi n\psi''_n(\tau_n)}} = o\left(\frac{1}{\tau_n \sqrt{n}}\right),$$

which contains the conclusion of Theorem 3.3 in Chaganty and Sethuraman (1993). Clearly, a multivariate version of this result would also follow directly from the discussion here, as would a result uniform in  $\tau$  in compact subsets of  $(0, \infty) \cap \Omega_-$ . Also, when sums of independent identically distributed variables are considered, so that  $\psi_n(z)$  does not depend on  $n$ , the result contained in Theorem 3 in Stone (1967) is easily derived (for the non-lattice case).

## 6 Proof of Theorem 7 and its Corollary

### Proof of Theorem 7.

Note first that  $P(\mathcal{W}_i(\theta) = 0)$  is a fixed point of the multivariate generating function of the underlying Galton-Watson process. Hence when the martingale converges in mean, so that  $E\mathcal{W}_i(\theta) = 1$ , these probabilities must be less than one and so must equal the extinction probabilities from that starting type. Thus  $\mathcal{W}_i(\theta) > 0$  agrees with the survival set, almost surely for  $\theta \in \Lambda_-$ . The continuity of  $\mathcal{W}_i(\theta)$  now means the null set can be chosen independent of  $\theta$ . For the rest of the calculation we deal with sample paths in  $\mathcal{S}$ . Then there is an  $N$  such that  $\mathcal{W}_{ij}^n(\theta) > 0$  for all  $n \geq N$ . We take  $n \geq N$ .

Let  $\mathcal{K}$  be a compact subset of  $\Lambda_-$  and fix  $j$ . Let the function  $w^n$  be defined by

$$w^n(\lambda) = \log \left( \frac{\mathcal{W}_{ij}^n(\lambda)v_i(\lambda)}{v_j(\lambda)} \right).$$

Using Theorems 1(ii), 3 and 5, there is an open  $B$  (with  $\Lambda_- \subset B \subset \Lambda$ ) such that, for each  $n$ ,  $w^n$  is analytic in  $\lambda \in B$ ,  $\{w^n\}$  converges as  $n \rightarrow \infty$  to an analytic function on  $B$  and for some  $\tau$

$$\mathcal{K}_\tau = \{\theta + \lambda \in \mathbb{C}^d : \theta \in \mathcal{K}, |\lambda| \leq \tau\} \subset B.$$

Analyticity of  $w^n$  and its limit on  $B$  mean that all its derivatives are uniformly bounded on  $\mathcal{K}_\tau$ .

To make the connection with the previous section, for fixed  $j$  and  $n$  let the measure  $Z$  be  $Z_{ij}^n$  and, for fixed  $\theta$ , let  $\nu(dx) = e^{-\theta x} Z(dx) / \int e^{-\theta y} Z(dy)$ . Note first that, with  $f(x) = e^{\theta x} h(x)$  and  $v_1 = v_1(\theta)$

$$\begin{aligned} e^{n\xi(\theta)} \int_C h(x) Z_{ij}^n(dx + nv_1(\theta)) &= \frac{e^{-n\theta v_1(\theta)}}{\rho(\theta)^n} \int_C f(x) e^{-\theta x} Z_{ij}^n(dx + nv_1(\theta)) \\ &= \frac{\int e^{-\theta x} Z_{ij}^n(dx)}{\rho(\theta)^n} \int_C f(x) \nu(dx + nv_1) \\ &= \frac{v_i(\theta) \mathcal{W}_{ij}^n(\theta)}{v_j(\theta)} \int_C f(x) \nu(dx + nv_1). \end{aligned}$$

Now, by Theorem 5,  $v_i(\theta) \mathcal{W}_{ij}^n(\theta) / v_j(\theta)$  converges to  $v_i(\theta) u_j(\theta) \mathcal{W}_i(\theta)$  uniformly in  $\theta \in \mathcal{K}$ . By assumption,  $f \in \mathfrak{G}(G)$  and so the results of the previous section can be applied to consider the convergence of  $\int_C f(x) \nu(dx + nv_1)$  once we show  $Z$  satisfies the appropriate conditions.

For  $\lambda \in B$ , substitution shows that

$$\log \left( \widehat{Z}(\lambda) \right) - n \log(\rho(\lambda)) = w^n(\lambda).$$

It follows from the uniform boundedness of the derivatives of  $w^n$  that  $\{m - nv_1 : n\}$  and  $\{\Sigma - nv_2 : n\}$  are bounded uniformly in  $\theta \in \mathcal{K}$ . Furthermore there are constants  $l > 0$  and  $L < \infty$  such that, for all large enough  $n$ ,  $\mathbf{c}/n \geq l$  and  $\mathbf{u}/n \leq L$  and then  $\mathbf{r} \leq L/l$ .

For any  $\epsilon > 0$ ,

$$\varepsilon(\epsilon) = \epsilon^{-d} \sup \left\{ \left| \frac{\widehat{Z}(\theta + \mathbf{i}\eta)}{\widehat{Z}(\theta)} \right| : \theta \in \mathcal{K}, \epsilon \leq |\eta| < 1/\epsilon \right\}$$

$$\begin{aligned}
&= \epsilon^{-d} \sup \left\{ \frac{v_j(\theta)}{v_i(\theta)\mathcal{W}_{ij}^n(\theta)} \left| \frac{\mathcal{M}_{ij}^n(\theta + \mathbf{i}\eta)}{\rho(\theta)^n} \right| : \theta \in \mathcal{K}, \epsilon \leq |\eta| < 1/\epsilon \right\} \\
&\rightarrow 0
\end{aligned}$$

geometrically quickly as  $n \rightarrow \infty$ , using Theorems 5 and 6. Hence, using Theorem 9,  $n^{d/2}A(Z_{ij}^n) \rightarrow 0$ . Furthermore, since  $\{m - nv_1 : n\}$  is uniformly bounded, Theorem 10 gives  $n^{(d+2)/2}B(m - nv_1) \rightarrow 0$ . Hence

$$n^{d/2} \left| \int_C f(x)\nu(dx + nv_1) - \frac{1}{\sqrt{(2\pi)^d \det[v_2]}} \int_C f(x)dx \right| \rightarrow 0$$

uniformly in  $\theta \in \mathcal{K}$  and  $C \in \mathfrak{C}$ . □

### Proof of Corollary 2.

In Theorem 7, take  $f \in \mathcal{D}$  to be one on  $|x| \leq b$ , zero on  $|x| > b + 1$  and with bounded derivatives. □

## 7 Extended Perron–Frobenius

A matrix  $M = \{m_{ij}\}_{p \times p}$ , has  $n$ th power  $M^n$  with entries denoted by  $m_{ij}^n$ . The eigenvalues of  $M$  are the zeros of the characteristic polynomial  $q(z) = \det[zI - M]$ , which is of degree  $p$ . Denote the roots of  $q(z)$  by  $\rho_1, \dots, \rho_p$  with the roots listed in order, so that

$$|\rho_1| \geq |\rho_2| \geq \dots \geq |\rho_p|, \tag{7.16}$$

The eigenvalue  $\rho_1$  of  $M$  is the maximum-modulus eigenvalue if  $|\rho_1| > |\rho_2|$ .

A non-negative square matrix  $A = \{a_{ij}\}$  is called *positive regular* if all its entries are finite and, for some  $n$ ,  $A^n$  has only strictly positive entries (see Lancaster and Tismenetsky (1985) or Seneta (1973)). Let the entries  $m_{ij}(\lambda)$  of the matrix  $M(\lambda) = \{m_{ij}(\lambda)\}_{p \times p}$  be functions of  $\lambda \in L$ . Clearly the eigenvalues and their multiplicities all depend on  $\lambda$ .

### Proof of Theorem 1.

Conclusions (i) and (ii) arise from routine applications of the implicit function theorem; details can be seen in Biggins and Rahimzadeh Sani (2004). Similar results have been obtained before, particularly for matrices of Laplace transforms. See, for example, Miller (1961, Theorem 1(a)), Lancaster and Tismenetsky (1985, Theorem 11.5.1), Ney and Numellin (1987, Theorem 4.1) and Kontoyiannis and Meyn (2003, Proposition 4.8(iii)). Now, by (i) there is an open set containing  $\tilde{L}$  on which  $|\rho_1| > |\rho_2|$ , where  $\rho = \rho_1$  — the eigenvalues here depend on  $\lambda$  but this is left implicit in the notation. For the rest of the proof  $\lambda$  will be confined to this open set.

The resolvent of  $M(\lambda)$  is defined by

$$R(z) = \{r_{ij}(z)\}_{p \times p} = (I - zM(\lambda))^{-1},$$

which, for all  $i$  and  $j$ , has the expansion  $r_{ij}(z) = \sum_{n=0}^{\infty} z^n m_{ij}^n(\lambda)$  when  $z < \|M(\lambda)\|^{-1}$ , where  $\|\cdot\|$  is a matrix norm, (see Lancaster and Tismenetsky (1985, Theorem 11.1.1)). For fixed  $\lambda \in \tilde{L}$ , let  $B = B(\tilde{\lambda}, \delta)$  and  $M_1$  be such that for all  $\lambda \in B$ ,  $\|M(\lambda)\| < M_1$ . Now we take  $|z| < 1/M_1$ ,  $\lambda \in B$  and suppress  $\lambda$  in the notation.

Let  $h(z) = \det[I - zM] = (1 - z\rho) \prod_{k=2}^p (1 - z\rho_k)$ . Inverting  $I - zM$  and rewriting using partial fractions

$$r_{ij}(z) = \frac{d_{ij}(z)}{(1 - z\rho) \prod_{k=2}^p (1 - z\rho_k)} = \frac{a_{ij}}{1 - z\rho} + \frac{\sum_{k=0}^{p-2} b_{k,ij} z^k}{\prod_{k=2}^p (1 - z\rho_k)}. \quad (7.17)$$

For fixed  $\lambda$ ,  $|b_{k,ij}/\rho^k|$  are bounded over  $i, j$  and  $k$  ( $\leq p-2$ ), by  $C$  say. Then, expansion of (7.17) gives

$$r_{ij}(z) = \sum_{n=0}^{\infty} \left( a_{ij} \rho^n + \sum_{h=0}^{p-2} b_{h,ij} \sum_{k_2+\dots+k_p=n-h} \rho_2^{k_2} \dots \rho_p^{k_p} \right) z^n \quad (7.18)$$

and so

$$m_{ij}^n - a_{ij} \rho^n = \sum_{h=0}^{p-2} b_{h,ij} \sum_{k_2+\dots+k_p=n-h} \rho_2^{k_2} \dots \rho_p^{k_p}.$$

Hence, for  $\alpha \in [|\rho_2|/|\rho|, 1)$ ,

$$|\rho^{-n} m_{ij}^n - a_{ij}| \leq C(p-1)(n+1)^{p-1} \alpha^{n-p}. \quad (7.19)$$

Let  $A$  be the matrix with entries  $\{a_{ij}\}$ . Then  $\rho^{-n} M^n \rightarrow A$  and so  $MA = AM = \rho A$ . Parts (i) and (ii) of the theorem now give  $A = cvu$  for some scalar  $c$ ; but  $v = \rho^{-n} M^n v \rightarrow Av = cv$ , using  $uv = 1$ , and so  $c = 1$ .

To translate (7.18) with  $a_{ij} = v_i u_j$  into an asymptotic estimate of  $M(\lambda)^n$ , the boundness of  $b_{k,ij}$  as  $\lambda$  varies is now needed. Let  $c_1(B)$  be the continuously differentiable functions on  $B$ . The function  $h(z)$  is a polynomial in  $z$  with coefficients in  $c_1(B)$  and  $1/\rho$  is a simple root of this polynomial. Hence  $h_1(z) = h(z)/(1 - \rho z)$  is a polynomial in  $z$  with coefficients in  $c_1(B)$ ; the same is true of  $d_{ij}(z)$ . From (7.17) and  $A = vu$

$$d_{ij}(z) = v_i u_j h_1(z) + \sum_{k=0}^{p-2} b_{k,ij} z^k (1 - z\rho).$$

Equating powers of  $z$  here shows that the  $b_{k,ij}$  are in  $c_1(B)$ . Hence the supremum of  $|b_{k,ij}(\lambda)/\rho(\lambda)^k|$  over  $i, j, k$  ( $\leq p-2$ ) and  $\lambda$  in the closed ball of radius  $\tilde{\delta}/2$  centred at  $\tilde{\lambda}$  will be finite; denote this finite supremum by  $C$ .

Let  $3\epsilon$  be less than  $(|\rho(\tilde{\lambda})| - |\rho_2(\tilde{\lambda})|)$  and small enough that balls of radius  $3\epsilon$  centred on the distinct eigenvalues of  $M(\tilde{\lambda})$  are disjoint. Then any point within  $\epsilon$  of one of  $\rho_2(\tilde{\lambda}), \dots, \rho_p(\tilde{\lambda})$  is smaller in magnitude than every point in the  $\epsilon$  ball centred on  $\rho(\tilde{\lambda})$ . Let  $\delta < \tilde{\delta}/2$  be small enough to ensure that the maximum distance between the eigenvalues at  $\tilde{\lambda}$  and those at  $\lambda \in B(\tilde{\lambda}, \delta)$ , after a suitable pairing, is less than  $\epsilon$ . Then, for  $\lambda \in B(\tilde{\lambda}, \delta)$ , and  $j = 2, 3, \dots, p$

$$\left| \frac{\rho_j(\lambda)}{\rho(\lambda)} \right| \leq \frac{|\rho_2(\tilde{\lambda})| + \epsilon}{\rho(\tilde{\lambda}) - \epsilon} \leq \frac{\rho(\tilde{\lambda}) - 2\epsilon}{\rho(\tilde{\lambda}) - \epsilon} = \alpha < 1.$$

With these definitions, the estimate (7.19) holds uniformly on  $B(\tilde{\lambda}, \delta)$  and this implies (1.2).  $\square$

Although we do not need it here, it is worth noting the following result, which applies to suitable matrices of Fourier transforms. Its proof requires only obvious modifications of the discussion already given.

**Theorem 11** *Theorem 1 remains true when ‘analytic’ is replaced by ‘ $q(\geq 1)$  times differentiable’ and  $\mathbb{C}^d$  is replaced by  $\mathbb{R}^d$ .*

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