

# Asymptotic Theory for Linear-Chain Conditional Random Fields - Supplementary Material -

## PROOF OF THEOREM 1

The existence of the asymptotic ratios  $r_{ij}$  is well-known (Lemma 3.4, Seneta, 2006). Let us establish the geometric rate. For any  $\ell \times \ell$ -matrix  $\mathbf{A} = (a_{ij})$ , define

$$\phi(\mathbf{A}) = \min_{i,j,k,l} \frac{a_{ik}a_{jl}}{a_{jk}a_{il}}.$$

Note that  $\phi(\mathbf{A}) \leq 1$ . Using the concept of Birkhoff's contraction coefficient, one can show that

$$\frac{1 - \sqrt{\phi(\mathbf{M}_n)}}{1 + \sqrt{\phi(\mathbf{M}_n)}} \leq \prod_{t=1}^n \frac{1 - \sqrt{\phi(\mathbf{M}(x_t))}}{1 + \sqrt{\phi(\mathbf{M}(x_t))}}$$

(Chapter 3, Seneta, 2006). With  $\psi^2$  defined in Lemma 1 and using the fact that  $\sqrt{\phi(\mathbf{M}_n)} \leq 1$ , we obtain

$$\frac{1 - \sqrt{\phi(\mathbf{M}_n)}}{2} \leq \left( \frac{1 - \psi}{1 + \psi} \right)^n.$$

After a few elementary algebraic manipulations and applying Bernoulli's inequality, we obtain

$$\phi(\mathbf{M}_n) \geq 1 - 4 \left( \frac{1 - \psi}{1 + \psi} \right)^n.$$

Now, note that the quantities

$$\max_{k \in \mathcal{Y}} \left( \frac{m_n(i, k)}{m_n(j, k)} \right) \quad \text{and} \quad \min_{k \in \mathcal{Y}} \left( \frac{m_n(i, k)}{m_n(j, k)} \right)$$

are non-increasing and non-decreasing with  $n$ , respectively (Lemma 3.1, Seneta, 2006). Moreover, by the definition of  $\phi(\cdot)$ , the ratio of the minimum to the maximum is greater than  $\phi(\mathbf{M}_n)$ .  $\square$

## PROOF OF THEOREM 2

We show that  $c$  and  $\kappa$  satisfy

$$\left| P_{\lambda}^{(-n,n)}(Y_t = y_t, \dots, Y_{t+k} = y_{t+k} \mid \mathbf{X} = \mathbf{x}) - P_{\lambda}(Y_t = y_t, \dots, Y_{t+k} = y_{t+k} \mid \mathbf{X} = \mathbf{x}) \right| \leq c\kappa^n$$

for all  $n \in \mathbb{N}$  such that  $-n \leq t$  and  $n \geq t+k$ . Introduce the vectors  $\underline{\mathbf{r}}_i(n)$  and  $\bar{\mathbf{r}}_i(n)$  with the  $k$ th components given by

$$\begin{aligned} \underline{r}_{ki}(n) &= \min_{l \in \mathcal{Y}} \left( \frac{g_n(k, l)}{g_n(i, l)} \right), \\ \bar{r}_{ki}(n) &= \max_{l \in \mathcal{Y}} \left( \frac{g_n(k, l)}{g_n(i, l)} \right). \end{aligned}$$

In the same way, we define vectors  $\underline{\mathbf{s}}_j(n)$  and  $\bar{\mathbf{s}}_j(n)$  with respect to  $\mathbf{H}_n$ . It is easy to see that

$$\begin{aligned} \underline{\mathbf{r}}_i(n)^T \mathbf{F} \underline{\mathbf{s}}_j(n) &\leq \frac{\alpha_{-n}^t(\boldsymbol{\lambda}, \mathbf{x})^T \mathbf{F} \beta_{t+k}^n(\boldsymbol{\lambda}, \mathbf{x})}{\alpha_{-n}^t(\boldsymbol{\lambda}, \mathbf{x}, i) \beta_{t+k}^n(\boldsymbol{\lambda}, \mathbf{x}, j)} \\ &\leq \bar{\mathbf{r}}_i(n)^T \mathbf{F} \bar{\mathbf{s}}_j(n). \end{aligned}$$

Furthermore, according to Theorem 1,

$$\underline{\mathbf{r}}_i(n)^T \mathbf{F} \underline{\mathbf{s}}_j(n) \leq \mathbf{r}_i^T \mathbf{F} \mathbf{s}_j \leq \bar{\mathbf{r}}_i(n)^T \mathbf{F} \bar{\mathbf{s}}_j(n).$$

Hence,

$$\begin{aligned} \left| \frac{\alpha_{-n}^t(\boldsymbol{\lambda}, \mathbf{x})^T \mathbf{F} \beta_{t+k}^n(\boldsymbol{\lambda}, \mathbf{x})}{\alpha_{-n}^t(\boldsymbol{\lambda}, \mathbf{x}, i) \beta_{t+k}^n(\boldsymbol{\lambda}, \mathbf{x}, j)} - \mathbf{r}_i^T \mathbf{F} \mathbf{s}_j \right| \\ \leq \left| (\bar{\mathbf{r}}_i(n) - \underline{\mathbf{r}}_i(n))^T \mathbf{F} (\bar{\mathbf{s}}_j(n) - \underline{\mathbf{s}}_j(n)) \right|. \end{aligned}$$

According to Theorem 1, we obtain

$$\begin{aligned} \left| \frac{\alpha_{-n}^t(\boldsymbol{\lambda}, \mathbf{x})^T \mathbf{F} \beta_{t+k}^n(\boldsymbol{\lambda}, \mathbf{x})}{\alpha_{-n}^t(\boldsymbol{\lambda}, \mathbf{x}, i) \beta_{t+k}^n(\boldsymbol{\lambda}, \mathbf{x}, j)} - \mathbf{r}_i^T \mathbf{F} \mathbf{s}_j \right| \\ \leq 16 \|\mathbf{F}\| \left( \frac{m_{\sup}}{m_{\inf}} \right)^2 \left( \frac{(1 - \varphi)(1 - \psi)}{(1 + \varphi)(1 + \psi)} \right)^n \end{aligned}$$

where  $\|\mathbf{F}\|$  stands for the sum of all components of  $\mathbf{F}$ . Putting all together, we have

$$\begin{aligned} \left| P_{\lambda}^{(-n,n)}(Y_t = y_t, \dots, Y_{t+k} = y_{t+k} \mid \mathbf{X} = \mathbf{x}) - P_{\lambda}(Y_t = y_t, \dots, Y_{t+k} = y_{t+k} \mid \mathbf{X} = \mathbf{x}) \right| \\ \leq 16 \|\mathbf{F}\| \left( \frac{m_{\sup}}{m_{\inf}} \right)^2 \left( \frac{(1 - \varphi)(1 - \psi)}{(1 + \varphi)(1 + \psi)} \right)^n \\ \times \prod_{i=1}^k m_{\lambda}(x_{t+i}, y_{t+i-1}, y_{t+i}), \end{aligned}$$

and now the value for the constant is  $c$  obtained by noting that

$$\|\mathbf{F}\| \prod_{i=1}^k m_{\lambda}(x_{t+i}, y_{t+i-1}, y_{t+i}) \leq \ell^{k+1} m_{\sup}^{2k}.$$

The proof is complete.  $\square$

## PROOF OF LEMMA 2

Let  $\vec{A} = \times_{t \in \mathbb{Z}} A_t$ . Note that  $\vec{\tau}^{-1} \vec{A} = \times_{t \in \mathbb{Z}} A_{t-1}$ , and hence  $\pi(\vec{\tau}^{-1} \vec{A}) = \pi(\vec{A})$  for all  $\vec{A} \in \mathcal{A}$  implies

$$\begin{aligned} \vec{\pi}(\vec{\tau}^{-1} \vec{A}) &= \pi\left(\bigcap_{t \in \mathbb{Z}} \tau^{-t} A_{t-1}\right) \\ &= \pi\left(\tau^{-1} \bigcap_{t \in \mathbb{Z}} \tau^{-(t-1)} A_{t-1}\right) \\ &= \pi\left(\bigcap_{t \in \mathbb{Z}} \tau^{-(t-1)} A_{t-1}\right) \\ &= \vec{\pi}(\vec{A}). \end{aligned}$$

Now suppose  $\vec{\tau}^{-1} \vec{A} = \vec{A}$ . A necessary condition for this is  $A_t = A$  for all  $t \in \mathbb{Z}$ . Setting  $\vec{A} = \bigcap_{t \in \mathbb{Z}} \tau^{-t}(A)$ , we obtain  $\vec{\pi}(\vec{A}) = \pi(\vec{A})$ . Now note that  $\tau^{-1} \vec{A} = \vec{A}$ . Thus, if  $\pi$  is  $\tau$ -ergodic, we have  $\pi(\vec{A}) = 0$  or  $\pi(\vec{A}) = 1$ , and hence  $\vec{\pi}(\vec{A}) = 0$  or  $\vec{\pi}(\vec{A}) = 1$ .  $\square$

## PROOF OF LEMMA 3

The proof that the invariant measure  $\mu_\lambda$  is unique requires an alternative representation of Markov processes. Write  $Q(\lambda, \mathbf{x}_1 \dots \mathbf{x}_n, i, j)$  to denote the  $(i, j)$ -th component of the product  $\mathbf{Q}(\lambda, \mathbf{x}_1) \dots \mathbf{Q}(\lambda, \mathbf{x}_n)$ . For  $k > 1$  consider the  $k$ th iterate of  $Q_\lambda$ :

$$Q_\lambda^k(z, C) = \int_{\mathcal{Z}} Q_\lambda(z', C) Q_\lambda^{k-1}(z, dz').$$

Note that

$$Q_\lambda^k((\vec{x}, y'_0, y'_1), \vec{A} \times \{y_0\} \times \{y_1\}) = \begin{cases} Q(\lambda, \mathbf{x}_0 \dots \mathbf{x}_{k-2}, y'_1, y_0) Q(\lambda, \mathbf{x}_{k-1}, y_0, y_1) & \text{if } \vec{\tau}^k \vec{x} \in \vec{A}, \\ 0 & \text{otherwise.} \end{cases}$$

Now let  $L_1 = L_1(\mu_\lambda)$  denote the space of measurable functions  $u : \mathcal{Z} \rightarrow \mathbb{R}$  satisfying  $\int_{\mathcal{Z}} |u(z)| \mu_\lambda(dz) < \infty$ . For  $k \in \mathbb{N}$  let  $Q_\lambda^k$  be the operator on  $L_1$  defined by

$$Q_\lambda^k u(z) = \int_{\mathcal{Z}} u(z') Q_\lambda^k(z, dz').$$

Note that, if  $k > 1$ ,

$$\begin{aligned} Q_\lambda^k u(\vec{x}, y'_0, y'_1) &= \sum_{y_0, y_1 \in \mathcal{Y}} u(\vec{\tau}^k \vec{x}, y_0, y_1) \\ &\quad \times Q(\lambda, \mathbf{x}_0 \dots \mathbf{x}_{k-2}, y'_1, y_0) Q(\lambda, \mathbf{x}_{k-1}, y_0, y_1). \end{aligned}$$

For the proof that the invariant measure  $\mu_\lambda$  is unique, let  $u_0 \in L_1$  with  $u_0 > 0$  and consider the *conservative* set  $C^* \subset \mathcal{Z}$  given by

$$C^* = \left\{ z \in \mathcal{Z} : \lim_{n \rightarrow \infty} \sum_{k=1}^n Q_\lambda^k u_0(z) = \infty \right\}.$$

Note that the set  $C^*$  is independent of the choice of  $u_0$ . Furthermore, let  $\mathcal{C}_i$  denote the class of *invariant sets*,

$$\mathcal{C}_i = \{C \in \mathcal{C} : Q_\lambda \mathbf{1}_C = \mathbf{1}_C \text{ } \mu_\lambda\text{-almost everywhere}\}.$$

We say that  $\mathcal{C}_i$  is *trivial* if  $\mu_\lambda(C) = 0$  or  $\mu_\lambda(C) = 1$  for every  $C \in \mathcal{C}_i$ . A sufficient condition for the existence of at most one invariant probability measure on  $(\mathcal{Z}, \mathcal{C})$  is that  $C^* = \mathcal{Z}$  (up to a  $\mu_\lambda$ -null set) and  $\mathcal{C}_i$  is trivial (Theorem VI.A, Foguel, 1969). We first show that  $C^* = \mathcal{Z}$ . According to Corollary 1 (ii), we have

$$\begin{aligned} \inf \{Q(\lambda, \mathbf{x}_1 \dots \mathbf{x}_n, i, j) : n \in \mathbb{N}, i, j \in \mathcal{Y}\} \\ \geq \frac{1}{\ell} \left( \frac{m_{\inf}}{m_{\sup}} \right)^2 \end{aligned}$$

for every  $\vec{x} = (\mathbf{x}_t)_{t \in \mathbb{Z}}$ . Hence, for  $k > 1$ ,

$$Q_\lambda^k u_0(\vec{x}, y'_0, y'_1) \geq \frac{1}{\ell^2} \left( \frac{m_{\inf}}{m_{\sup}} \right)^4 \sum_{y_0, y_1 \in \mathcal{Y}} u_0(\vec{\tau}^k \vec{x}, y_0, y_1).$$

Furthermore, since  $\vec{P}_X$  is  $\vec{\tau}$ -ergodic on  $(\vec{\mathcal{X}}, \vec{\mathcal{A}})$ ,

$$\begin{aligned} \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{k=1}^n u_0(\vec{\tau}^k \vec{x}, y_0, y_1) \\ = \int_{\vec{\mathcal{X}}} u_0(\vec{x}', y_0, y_1) \vec{P}_X(d\vec{x}') \end{aligned}$$

for  $\vec{P}_X$ -almost every  $\vec{x} \in \vec{\mathcal{X}}$ . Now, under the assumption  $u_0 > 0$ , the integral on the right hand side is strictly greater than 0, hence the unnormalized series on the left hand side would tend to  $\infty$ . This argument shows that the series in the definition of  $C^*$  diverges for  $\mu_\lambda$ -almost every  $z \in \mathcal{Z}$ , and hence  $C^* = \mathcal{Z}$  up to a  $\mu_\lambda$ -null set.

To show that  $\mathcal{C}_i$  is trivial, let  $C \in \mathcal{C}_i$  be such that  $\mu_\lambda(C) > 0$  and  $Q_\lambda \mathbf{1}_C(z) = \mathbf{1}_C(z)$  for  $\mu_\lambda$ -almost every  $z \in \mathcal{Z}$ . Note that  $Q_\lambda \mathbf{1}_C(z) = Q_\lambda(z, C)$ . If (A1) holds, then all entries of the transition matrix  $\mathbf{Q}$  are strictly greater than 0, and hence a necessary condition for  $Q_\lambda(z, C) = 1$  is that  $C = \vec{A} \times \mathcal{Y} \times \mathcal{Y}$  for some set  $\vec{A} \in \vec{\mathcal{A}}$ , which implies that  $Q_\lambda(z, C) = \mathbf{1}_{\vec{A}}(\vec{\tau} \vec{x})$  and  $\mathbf{1}_C(z) = \mathbf{1}_{\vec{A}}(\vec{x})$  for  $\mu_\lambda$ -almost every  $z = (\vec{x}, y_0, y_1) \in \mathcal{Z}$ . Now note that  $\mathbf{1}_{\vec{A}}(\vec{\tau} \vec{x}) = \mathbf{1}_{\vec{A}}(\vec{x})$  is equivalent to  $\vec{A} = \vec{\tau}^{-1} \vec{A}$ , and if (A2) holds, then  $\vec{P}_X(\vec{A}) = 0$  or  $\vec{P}_X(\vec{A}) = 1$  for each set  $\vec{A}$  satisfying this condition.  $\square$

## PROOF OF LEMMA 5

We wish to establish that

$$\begin{aligned} \frac{1}{n} \sum_{t=1}^n E_\lambda^{(0:n)} [\mathbf{f}(X_t, Y_{t-1}, Y_t) | \mathbf{X}] \\ \sim \frac{1}{n} \sum_{t=1}^n E_\lambda [\mathbf{f}(X_t, Y_{t-1}, Y_t) | \mathbf{X}]. \end{aligned}$$

Let  $i, j \in \mathcal{Y}$ . Similar to the proof of Theorem 2, we obtain that  $P_\lambda^{(0:n)}(Y_{t-1} = i, Y_t = j | \mathbf{X} = \mathbf{x})$  converges to some limit  $P_\lambda^{(0:\infty)}(Y_{t-1} = i, Y_t = j | \mathbf{X} = \mathbf{x})$  as  $n$  tends to infinity, and there exist constants  $c > 0$  and  $0 < \kappa < 1$  not depending on  $\mathbf{x}$  such that

$$\begin{aligned} & |P_\lambda^{(0:n)}(Y_{t-1} = i, Y_t = j | \mathbf{X} = \mathbf{x}) - \\ & P_\lambda^{(0:\infty)}(Y_{t-1} = i, Y_t = j | \mathbf{X} = \mathbf{x})| \leq c\kappa^{n-t}. \end{aligned}$$

Consequently,

$$\begin{aligned} \lim_{n \rightarrow \infty} \frac{1}{n} \sum_{t=1}^n |P_\lambda^{(0:n)}(Y_{t-1} = i, Y_t = j | \mathbf{X} = \mathbf{x}) - \\ P_\lambda^{(0:\infty)}(Y_{t-1} = i, Y_t = j | \mathbf{X} = \mathbf{x})| = 0 \end{aligned}$$

which shows that

$$\begin{aligned} \frac{1}{n} \sum_{t=1}^n E_\lambda^{(0:n)}[\mathbf{f}(X_t, Y_{t-1}, Y_t) | \mathbf{X}] \\ \sim \frac{1}{n} \sum_{t=1}^n E_\lambda^{(0:\infty)}[\mathbf{f}(X_t, Y_{t-1}, Y_t) | \mathbf{X}], \end{aligned}$$

where  $E_\lambda^{(0:\infty)}$  stands for the conditional expectation with respect to  $P_\lambda^{(0:\infty)}$ . Now, noting that

$$E_\lambda^{(0:\infty)}[\mathbf{f}(X_t, Y_{t-1}, Y_t) | \mathbf{X}] \sim E_\lambda[\mathbf{f}(X_t, Y_{t-1}, Y_t) | \mathbf{X}],$$

we obtain the statement.  $\square$

## PROOF OF LEMMA 7

Let  $\mathbf{x} = (x_t)_{t \in \mathbb{Z}}$  be fixed. Using Corollary 1 (ii) and arguments similar to the proof of Theorem 2, it is not difficult to show that the difference between the probabilities  $P_\lambda(Y_{t-1} = i, Y_t = j, Y_{t+k-1} = l, Y_{t+k} = m | \mathbf{X} = \mathbf{x})$  and  $P_\lambda(Y_{t-1} = i, Y_t = j | \mathbf{X} = \mathbf{x}) \times P_\lambda(Y_{t+k-1} = l, Y_{t+k} = m | \mathbf{X} = \mathbf{x})$  decays at a geometric rate. Since  $\mathbf{f}$  is bounded, it follows that the covariance of  $\mathbf{f}(X_t, Y_{t-1}, Y_t)$  and  $\mathbf{f}(X_{t+k}, Y_{t+k-1}, Y_{t+k})$  conditional on  $\mathbf{X} = \mathbf{x}$  decays component-wise at a geometric rate, and integrating with respect to  $P_\mathbf{X}$  shows that  $\gamma_\lambda(k)$  decays to 0 at a geometric rate. Consequently,  $\sum_{k=1}^n \gamma_\lambda(k) < \infty$ . Similar to the proof of Lemma 5, we obtain that

$$\lim_{n \rightarrow \infty} \hat{\gamma}_\lambda^{(n)}(k) = \gamma_\lambda(k)$$

and

$$\nabla^2 \mathcal{L}_n(\boldsymbol{\lambda}) \sim - \left( \gamma_\lambda(0) + 2 \sum_{k=1}^n \gamma_\lambda(k) \right)$$

$P_{\lambda_0}$ -almost surely. The proof is complete.  $\square$