

On Occurrences of Words under Markovian Hypothesis

Ourania Chryssaphinou¹, Margarita Karaliopoulou¹, and Nikolaos Limnios²

¹ University of Athens
Department of Mathematics,
157 84, Athens,
Greece
(e-mail: ocrysaf@cc.uoa.gr, mkaraliop@math.uoa.gr)

² Universite de Technologie de Compiegne
Laboratoire de Mathematiques Appliquees Centre de Recherches de Royallieu
BP 20529,
60205 COMPIEGNE CEDEX, France
(e-mail: Nikolaos.Limnios@utc.fr)

Abstract. We consider a finite set of words $W = \{w_1, w_2, \dots, w_\nu\}$ which are produced under the Markovian hypothesis. We study the distances between word occurrences and we give explicit formulae for the corresponding distributions in the case of having words of equal lengths. The obtained results can be applied to certain problems concerning DNA sequences, as well as, general sequential analysis.

Keywords: Word, Markov chain, distance between occurrences, semi-Markov, waiting time.

1 Preliminaries

Consider an alphabet $\Omega = \{\alpha_1, \dots, \alpha_\ell\}$ with $\ell \geq 2$. We call word a finite sequence of elements of Ω . Let $W = \{w_1, \dots, w_\nu\}$ a finite sets of words where $w_i = (\alpha_{i_1}, \dots, \alpha_{i_{k_i}})$, $\alpha_{i_{n_i}} \in \Omega$, $n_i = 1, \dots, k_i$ where k_i denotes the length of word w_i and let $k_i > 1$. We assume that the set of words is reduced. Let us consider a sequence of outcomes $\{J_n^*\}_{n \geq 1}$ generated by a Markov chain with state space Ω , and let $\mathbf{P} = (\wp(\alpha_i, \alpha_j))_{\alpha_i, \alpha_j \in \Omega}$, the transition probability matrix. We write $\mathbf{P}_l^n = (\wp^n(\alpha_i, \alpha_j))_{\alpha_i, \alpha_j \in \Omega}$, where

$$\wp^n(\alpha_i, \alpha_j) = P(J_{n+1}^* = \alpha_j | J_1^* = \alpha_i).$$

A word w_i occurs at time γ iff $J_{\gamma-k_i+1}^* = \alpha_{i_1}, \dots, J_\gamma^* = \alpha_{i_{k_i}}$.

Definition 1 Let W_λ a subset of W . We define

$$U^* = \min\{\gamma \geq 1 : \text{a word occurs at } \gamma\},$$

$$M_{W_\lambda}^* = \min\{\gamma \geq 1 : \text{a word from the subset } W_\lambda \text{ occurs at } \gamma\},$$

and let y_0 be the first word which appears.

Clearly, the variable U^* indicates the waiting time (number of letters) for the first occurrence of any word from the set W , while the variable $M_{W_\lambda}^*$ indicates the waiting time (number of letters) for the first occurrence of any word from the subset W_λ . In section 2 we assume words of the same length and we obtain explicit formulae concerning the distributions of the above random variables. In section 3, under the same assumption we model the process of word occurrences via a semi-Markov model for which we derive the kernel as well as relative results. The under consideration random variables are of great interest for the study of biological sequences where the corresponding alphabet is $\Omega = \{A, C, G, T\}$.

Recurrent relations for variable $M_{\{w_i\}}^*$ are given by Blom and Thorburn (1982) in the i.i.d case, by Chryssaphinou and Papastavridis (1990) and Robin and Daudin (1999) in the Markov case.

2 Words of the same length and without a word at the beginning of the sequence

Let us examine the case where $k_i = k, \forall w_i \in W$. We construct a new Markov Chain $\{X_n^*\}_{n \geq 1}$ where

$$X_n^* = (J_n^*, \dots, J_{n+k-1}^*), \quad n \geq 1, \tag{1}$$

with state space $\Omega^k = \Omega \times \dots \times \Omega$ and

$$u_i = (\alpha_1^{u_i}, \dots, \alpha_k^{u_i}), \quad \forall i = 1, \dots, \ell^k \quad \text{and} \quad \alpha_n^{u_i} \in \Omega, \quad \forall n = 1, \dots, k,$$

The new transition matrix is

$$\tilde{P} = (\tilde{p}(u_i, u_j)), \quad u_i, u_j \in \Omega^k, \tag{2}$$

where

$$\begin{aligned} \tilde{p}(u_i, u_j) &= \mathbb{P}(X_{n+1}^* = u_j | X_n^* = u_i) \\ &= \mathbb{P}(J_{n+1}^* = \alpha_1^{u_j}, \dots, J_{n+k}^* = \alpha_k^{u_j} | J_n^* = \alpha_1^{u_i}, \dots, J_{n+k-1}^* = \alpha_k^{u_i}) \\ &= I_{\{\alpha_2^{u_i} = \alpha_1^{u_j}, \dots, \alpha_k^{u_i} = \alpha_{k-1}^{u_j}\}} \wp(\alpha_{k-1}^{u_j}, \alpha_k^{u_j}). \end{aligned} \tag{3}$$

The initial distribution is

$$\begin{aligned} \mathbb{P}(X_1^* = u_i) &= \mathbb{P}(J_1^* = \alpha_1^{u_i}, \dots, J_k^* = \alpha_k^{u_i}) \\ &= \sigma(\alpha_1^{u_i}) \wp(\alpha_1^{u_i}, \alpha_2^{u_i}), \dots, \wp(\alpha_{k-1}^{u_i}, \alpha_k^{u_i}), \end{aligned} \tag{4}$$

where σ is the initial distribution of Markov chain J^* . We note

$$\tilde{P}_1 = (\mathbb{P}(X_1^* = u_1), \dots, \mathbb{P}(X_1^* = u_{\ell^k})). \tag{5}$$

Since $W \subseteq \Omega^k, \exists r_1, \dots, r_\nu \in \{1, \dots, \ell^k\} : w_1 = u_{r_1}, \dots, w_\nu = u_{r_\nu}$. Let $B^c = \Omega^k \setminus B, \forall B \subseteq \Omega^k$. We define the matrices

$$\tilde{P}_{B^c B^c}, \quad \tilde{P}_{B^c B}, \tag{6}$$

which are the restriction of the transition matrices \tilde{P} in $B^c \times B^c$ and $B^c \times B$ respectively. Generally $\forall B_1, B_2 \subseteq \Omega^k$, let $\tilde{P}_{B_1 B_2}$ the restriction of \tilde{P} in $B_1 \times B_2$.

For $W_\lambda \subseteq W$, where $|W_\lambda| = \lambda$, we now define the n th-order transition matrix

$$\tilde{P}_{W_\lambda^c W_\lambda^c}^n = (\tilde{p}_{W_\lambda^c W_\lambda^c}^n(u_i, u_j)), \quad u_i, u_j \in W_\lambda^c, \tag{7}$$

where $\tilde{p}_{W_\lambda^c W_\lambda^c}^n(u_i, u_j) = \mathbb{P}(X_{n+1}^* = u_j, X_n^* \in W_\lambda^c, \dots, X_2^* \in W_\lambda^c | X_1^* = u_i)$, and $\mathbb{P}_{W_\lambda^c W_\lambda^c}^1 = \mathbb{P}_{W_\lambda^c W_\lambda^c}, \mathbb{P}_{W_\lambda^c W_\lambda^c}^0 = \mathbb{I}_{\ell^k - \lambda}$. Finally, let us define

$$\tilde{P}_{W_\lambda} = (\mathbb{P}(X_1^* = u_i)), \quad u_i \in W_\lambda, \quad \tilde{P}_{W_\lambda^c} = (\mathbb{P}(X_1^* = u_i)), \quad u_i \notin W_\lambda. \tag{8}$$

Now we are ready to present the following results.

Proposition 1 *With the above notation the distribution of the random variable $M_{W_\lambda}^*$ is given by*

$$\mathbb{P}(M_{W_\lambda}^* = n) = \begin{cases} 0, & n < k, \\ \tilde{P}_{W_\lambda} \mathbf{1}'_\lambda, & n = k, \\ [\tilde{P}_{W_\lambda^c}] \times [\tilde{P}_{W_\lambda^c W_\lambda^c}^{n-k-1}] \times [\tilde{P}_{W_\lambda^c W_\lambda}] \mathbf{1}'_\lambda, & n > k. \end{cases} \tag{9}$$

where $\mathbf{1}_\lambda = (1, \dots, 1)$, ($1 \times \lambda$ matrix)

Proof. It is $\mathbb{P}(M_{W_\lambda}^* = k) = \mathbb{P}(X_1^* \in W_\lambda) = \tilde{P}_{W_\lambda} \mathbf{1}'_\lambda$. For $n > k$ we have

$$\begin{aligned} \mathbb{P}(M_{W_\lambda}^* = n) &= \mathbb{P}(X_{n-k+1}^* \in W_\lambda, X_{n-k}^* \in W_\lambda^c, \dots, X_2^* \in W_\lambda^c, X_1^* \in W_\lambda^c) \\ &= \sum_{u_i \in W_\lambda^c} \mathbb{P}(X_{n-k+1}^* \in W_\lambda, X_{n-k}^* \in W_\lambda^c, \dots, X_2^* \in W_\lambda^c \mid X_1^* = u_i) \\ &\quad \mathbb{P}(X_1^* = u_i) \\ &= \sum_{u_i \in W_\lambda^c} \sum_{u_j \in W_\lambda} \mathbb{P}(X_{n-k+1}^* = u_j \mid X_{n-k}^* \in W_\lambda^c, \dots, X_2^* \in W_\lambda^c, X_1^* = u_i) \\ &\quad \mathbb{P}(X_{n-k}^* \in W_\lambda^c, X_{n-k-1}^* \in W_\lambda^c, \dots, X_2^* \in W_\lambda^c \mid X_1^* = u_i) \mathbb{P}(X_1^* = u_i) \\ &= \sum_{u_i \in W_\lambda^c} \sum_{u_j \in W_\lambda} \sum_{u_l \notin W_\lambda} \mathbb{P}(X_{n-k+1}^* = u_j \mid X_{n-k}^* = u_l) \\ &\quad \mathbb{P}(X_{n-k}^* = u_l, X_{n-k-1}^* \in W_\lambda^c, \dots, X_2^* \in W_\lambda^c \mid X_1^* = u_i) \mathbb{P}(X_1^* = u_i) \\ &= [\tilde{P}_{W_\lambda^c}] \times [\tilde{P}_{W_\lambda^c W_\lambda^c}^{n-k-1}] \times [\tilde{P}_{W_\lambda^c W_\lambda}] \mathbf{1}'_\lambda, \end{aligned}$$

which completes the proof.

Proposition 2 For every $w_i \in W$ the following is valid

$$\mathbb{P}(U^* = \gamma, y_0 = w_i) = \begin{cases} \tilde{P}_1 e'_{\ell^k; r_i}, & \gamma = k, \\ \tilde{P}_{W^c} [\tilde{P}_{W^c W^c}]^{\gamma-k-1} \tilde{P}_{W^c \Omega^k} e'_{\ell^k; r_i}, & \gamma > k, \end{cases} \quad (10)$$

where $e_{n;m} = (0, \dots, \underbrace{1}_{m\text{-position}}, \dots, 0)$ ($1 \times n$ matrix)

Proof. The case of $\gamma = k$ is obvious, since $\mathbb{P}(U^* = k, y_0 = w_i) = \mathbb{P}(X_1^* = u_{r_i})$. For $\gamma > k$ we proceed as follows

$$\begin{aligned} & \mathbb{P}(U^* = \gamma, y_0 = w_i) \\ &= \mathbb{P}(X_{\gamma-k+1}^* = u_{r_i}, X_{\gamma-k}^* \in W^c, \dots, X_2^* \in W^c, X_1^* \in W^c) \\ &= \sum_{u_s \notin W_\lambda} \mathbb{P}(X_{\gamma-k+1}^* = u_{r_i}, X_{\gamma-k}^* \in W^c, \dots, X_2^* \in W^c \mid X_1^* = u_s) \\ & \quad \mathbb{P}(X_1^* = u_s) \\ &= \sum_{u_s \notin W} \mathbb{P}(X_{\gamma-k+1}^* = u_{r_i} \mid X_{\gamma-k}^* \in W^c, \dots, X_2^* \in W^c, X_1^* = u_s) \\ & \quad \mathbb{P}(X_{\gamma-k}^* \in W^c, X_{\gamma-k-1}^* \in W^c \dots, X_2^* \in W^c \mid X_1^* = u_s) \\ & \quad \mathbb{P}(X_1^* = u_s) \\ &= \sum_{u_s \notin W} \sum_{u_l \notin W} \mathbb{P}(X_{\gamma-k+1}^* = u_{r_i} \mid X_{\gamma-k}^* = u_l) \\ & \quad \mathbb{P}(X_{\gamma-k}^* = u_l, X_{\gamma-k-1}^* \in W^c \dots, X_2^* \in W^c \mid X_1^* = u_s) \\ & \quad \mathbb{P}(X_1^* = u_s) \\ &= \tilde{P}_{W^c} [\tilde{P}_{W^c W^c}]^{\gamma-k-1} \tilde{P}_{W^c \Omega^k} e'_{\ell^k; r_i}, \end{aligned}$$

which ends the proof.

3 Words of the same length and with a word at the beginning of the sequence

We now consider $\{J_n\}$ where $J_n = J_{U^*+n}^*, \forall n \geq -U^* + 1$. We want to study the sequence J_0, J_1, \dots under the assumption that the word w_i has occurred with probability $\theta_i = \mathbb{P}(J_{-k_i+1} = \alpha_{i_1}, \dots, J_0 = \alpha_{i_{k_i}})$, $i = 1, \dots, \nu$. Without loss of generality we can take $\theta_i = \mathbb{P}(y_0 = w_i)$, $i = 1, \dots, \nu$.

The sequence $\{J_n, n \geq 0\}$ is a Markov chain with first order transition probabilities $\mathbb{P}(J_{n+1} = \alpha_j \mid J_n = \alpha_i) = \wp(\alpha_i, \alpha_j)$ and $\mathbb{P}(J_0 = \alpha_\zeta) = \sum_{i=1}^\nu I_{\{\alpha_{i_{k_i}} = \alpha_\zeta\}} \theta_i$.

In this case a word w_i occurs at time γ iff $J_{\gamma-k_i+1} = \alpha_{i_1}, \dots, J_\gamma = \alpha_{i_{k_i}}$.

3.1 The Semi-Markov Model

In the case where words do not overlap, Biggins and Cannings (1987), introduced the idea of modelling the process of word occurrences via a semi-Markov model. Recently, Robin and Daudin (2001) generalized the idea considering the fact that words may overlap. Our aim is to determine the kernel of this new process. In order to present our results we need some more definitions and notations.

Definition 2 *Let us define the stochastic processes $\{U_n, n \geq 0\}, \{y_n, n \geq 0\}$, which describe the times of word occurrences and the corresponding words respectively, where $U_0 = 0$ and*

$$U_n = \min\{\gamma > U_{n-1} : a \text{ word from } W \text{ occurs at } \gamma\}, \quad n \geq 1, \quad (11)$$

$$y_n = w_i, \quad w_i \in W, \quad i = 1, \dots, \ell. \quad (12)$$

The process $\{(y_n, U_n), n \in \mathbb{N}\}$ is an homogenous Discrete time Markov Renewal Process(DTMRP) since

$$\begin{aligned} &\mathbb{P}(y_{n+1} = w_j, U_{n+1} - U_n = \gamma | y_0, \dots, y_n = w_i, U_0, \dots, U_n) = \\ &\mathbb{P}(y_{n+1} = w_j, U_{n+1} - U_n = \gamma | y_n = w_i) = \\ &\mathbb{P}(y_1 = w_j, U_1 = \gamma | y_0 = w_i) = q_{ij}(\gamma), \quad \forall n \geq 1 \end{aligned}$$

Let us consider the following notation

- \mathcal{M}_E , the set of non negative matrices on $E \times E$.
- $\mathbf{I}_E \in \mathcal{M}_E$, the identity matrix, $0_A \in \mathcal{M}_E$, the null matrix.
- $\mathcal{M}_E(\mathbb{N})$, the set of matrix-valued functions: $\mathbb{N} \rightarrow \mathcal{M}_E$. If $A \in \mathcal{M}_E(\mathbb{N})$, we have $A = (A(\gamma) : \gamma \in \mathbb{N})$, where for fixed $\gamma \in \mathbb{N}$, $A(\gamma) = (A_{ij}(\gamma) : i, j \in E) \in \mathcal{M}_E$.

Then $q \in \mathcal{M}_E(\mathbb{N})$ ($E = \{1, \dots, \nu\}$) is the discrete time semi-Markov kernel relevant to the DTMRP $\{(y_n, U_n), n \in \mathbb{N}\}$. We have

$$q_{ij}^{(r)}(\gamma) = \mathbb{P}(y_r = w_j, U_r = \gamma | y_0 = w_i), \quad (13)$$

where $q^{(r)}$ is the r- fold convolution of q .Then we can define

$$\psi_{ij}(\gamma) = \sum_{r=0}^{\gamma} q_{ij}^{(r)}(\gamma), \quad w_i, w_j \in W, \quad \gamma \in \mathbb{N}. \quad (14)$$

We can write

$$\psi_{ij}(\gamma) = q_{ij}(\gamma) + \sum_{s=1}^{\nu} \sum_{z=1}^{\gamma-1} \psi_{is}(z)q_{sj}(\gamma - z), \quad \text{for } \gamma \geq 1. \quad (15)$$

Definition 3 For all $r \in \mathbb{N}^*$, $\forall w_i, w_j \in W$ let $M_{ij}^{(r)}$ be the number of letters of the r -th occurrence of w_j after w_i 's occurrence.

Definition 4 For all $W_\lambda \subseteq W$ we define

$$M_{iW_\lambda} = \min_{n \geq 1} \{U_n : y_n \in W_\lambda\} \text{ under the event } \{y_0 = w_i\}. \quad (16)$$

We will note M_{ij} for $M_{i\{w_j\}}$. Obviously $M_{ij} = M_{ij}^{(1)}$.

If $g_{ij}(\gamma) = \mathbb{P}(M_{ij} = \gamma)$ and $g_{ij}^{(r)}(\gamma) = \mathbb{P}(M_{ij}^{(r)} = \gamma)$, then

$$\psi_{ij}(\gamma) = \begin{cases} \sum_{r=0}^{\gamma} g_{jj}^{(r)}(\gamma), & i = j \\ \sum_{r=0}^{\gamma} g_{ij} * g_{jj}^{(r)}(\gamma), & i \neq j. \end{cases} \quad (17)$$

Definition 5 We assume $k_i = k$ for all $w_i \in W$. We define

$$X_n = (J_{n-k+1}, \dots, J_n), \quad n \geq 0, \quad (18)$$

with

$$\begin{aligned} \mathbb{P}(X_0 = u_{r_i}) &= \mathbb{P}(y_0 = w_i) = \mathbb{P}(J_{-k+1} = \alpha_{i_1}, \dots, J_0 = \alpha_{i_k}), \\ \mathbb{P}(X_0 = u_i) &= 0, \quad \forall u_i \notin W. \end{aligned}$$

Clearly, $\{X_n, n \geq 0\}$ is a Markov Chain with state space $\Omega^k = \{u_1, \dots, u_{\ell^k}\}$, where $\forall i = 1, \dots, \ell^k$, $u_i = (\alpha_1^{u_i}, \dots, \alpha_k^{u_i})$, $\alpha_\zeta^{u_i} \in \Omega$, $\forall \zeta = 1, \dots, k$ and $\tilde{P} = (\tilde{p}(u_i, u_j)) = (\mathbb{P}(X_{n+1} = u_j | X_n = u_i)), \forall u_i, u_j \in \Omega^k$.

Using the above definitions and notations we obtain the following results:

Proposition 3 For every $w_i, w_j \in W$ we have:

$$q_{ij}(\gamma) = \begin{cases} e_{\ell^k; r_i} \tilde{P} e'_{\ell^k; r_j}, & \gamma = 1 \\ e_{\ell^k; r_i} [\tilde{P}_{\Omega^k W^c}] [\tilde{P}_{W^c W^c}]^{\gamma-2} [\tilde{P}_{W^c \Omega^k}] e'_{\ell^k; r_j}, & \gamma \geq 2 \end{cases} \quad (19)$$

Proof. It is

$$q_{ij}(1) = \mathbb{P}(X_1 = w_j | X_0 = w_i) = e_{\ell^k; r_i} \tilde{P} e'_{\ell^k; r_j}.$$

For $\gamma \geq 2$ we have

$$\begin{aligned} q_{ij}(\gamma) &= \mathbb{P}(X_\gamma = w_j, X_1 \notin W, \dots, X_{\gamma-1} \notin W | X_0 = w_i) \\ &= \mathbb{P}(X_\gamma = u_{r_j}, X_1 \notin W, \dots, X_{\gamma-1} \notin W | X_0 = u_{r_i}) \\ &= \sum_{u_n, u_s \notin W} \\ &\mathbb{P}(X_\gamma = u_{r_j}, X_{\gamma-1} = u_n, X_{\gamma-2} \notin W, \dots, X_2 \notin W, X_1 = u_s | X_0 = u_{r_i}) \\ &= \sum_{u_n, u_s \notin W} \\ &\mathbb{P}(X_\gamma = u_{r_j}, X_{\gamma-1} = u_n, X_{\gamma-2} \notin W, \dots, X_2 \notin W, | X_1 = u_s, X_0 = u_{r_i}) \\ &\mathbb{P}(X_1 = u_s | X_0 = u_{r_i}) \end{aligned}$$

that is

$$\begin{aligned}
 & q_{ij}(\gamma) \\
 &= \sum_{u_n, u_s \notin W} \mathbb{P}(X_\gamma = u_{r_j}, X_{\gamma-1} = u_n, X_{\gamma-2} \notin W, \dots, X_2 \notin W, |X_1 = u_s) \\
 &\quad \mathbb{P}(X_1 = u_s | X_0 = u_{r_i}) \\
 &= \sum_{u_n, u_s \notin W} \mathbb{P}(X_\gamma = u_{r_j} | X_{\gamma-1} = u_n, X_{\gamma-2} \notin W, \dots, X_2 \notin W, X_1 = u_s) \\
 &\quad \mathbb{P}(X_{\gamma-1} = u_n, X_{\gamma-2} \notin W, \dots, X_2 \notin W, |X_1 = u_s) \mathbb{P}(X_1 = u_s | X_0 = u_{r_i}) \\
 &= \sum_{u_n, u_s \notin W} \mathbb{P}(X_\gamma = u_{r_j} | X_{\gamma-1} = u_n) \\
 &\quad \mathbb{P}(X_{\gamma-1} = u_n, X_{\gamma-2} \notin W, \dots, X_2 \notin W, |X_1 = u_s) \mathbb{P}(X_1 = u_s | X_0 = u_{r_i}) \\
 &= e_{\ell^k; r_i} [\tilde{P}_{\Omega^k W^c}] [\tilde{P}_{W^c W^c}]^{\gamma-2} [\tilde{P}_{W^c \Omega^k}] e'_{\ell^k; r_j} .
 \end{aligned}$$

Proposition 4 *It is*

$$\mathbb{P}(M_{iW_\lambda} = \gamma) = \begin{cases} e_{\ell^k; r_i} \tilde{P}_{\Omega^k W_\lambda} \mathbf{1}'_\lambda, & \gamma = 1 \\ e_{\ell^k; r_i} \tilde{P}_{\Omega^k W_\lambda^c} [\tilde{P}_{W_\lambda^c W_\lambda^c}]^{\gamma-2} \tilde{P}_{W_\lambda^c W_\lambda} \mathbf{1}'_\lambda, & \gamma \geq 2. \end{cases} \quad (20)$$

Proof. The random variable M_{iW_λ} can be expressed as follows

$$M_{iW_\lambda} = \min\{n \geq 1 : X_n \in W_\lambda\} \text{ over } \{X_0 = w_i\}. \quad (21)$$

If $\gamma = 1$, then

$$\mathbb{P}(M_{iW_\lambda} = 1) = \mathbb{P}(X_1 \in W_\lambda \mid X_0 = w_i) = e_{\ell^k; r_i} \tilde{P}_{\Omega^k W_\lambda} \mathbf{1}'_\lambda.$$

If $\gamma \geq 2$, then

$$\begin{aligned}
 & \mathbb{P}(M_{iW_\lambda} = \gamma) = \mathbb{P}(X_\gamma \in W_\lambda, X_{\gamma-1} \notin W_\lambda, \dots, X_1 \notin W_\lambda \mid X_0 = w_i) \\
 &= \sum_{u_r \in W_\lambda} \mathbb{P}(X_\gamma = u_r, X_{\gamma-1} \notin W_\lambda, \dots, X_1 \notin W_\lambda \mid X_0 = w_i) \\
 &= \sum_{u_r \in W_\lambda} \sum_{u_s, u_n \notin W_\lambda} \mathbb{P}(X_\gamma = u_r, X_{\gamma-1} = u_s, \dots, X_1 = u_n \mid X_0 = w_i) \\
 &= \dots = \sum_{u_r \in W_\lambda} \sum_{u_s \notin W_\lambda} \sum_{u_n \notin W_\lambda} \mathbb{P}(X_\gamma = u_r \mid X_{\gamma-1} = u_s) \\
 &\quad \mathbb{P}(X_{\gamma-1} = u_s, X_{\gamma-2} \notin W_\lambda, \dots, X_2 \notin W_\lambda \mid X_1 = u_n) \\
 &\quad \mathbb{P}(X_1 = u_n \mid X_0 = w_i) \\
 &= e_{\ell^k; r_i} \tilde{P}_{\Omega^k W_\lambda^c} [\tilde{P}_{W_\lambda^c W_\lambda^c}]^{\gamma-2} \tilde{P}_{W_\lambda^c W_\lambda} \mathbf{1}'_\lambda.
 \end{aligned}$$

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