## Notes

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## 1 Derivation of Log-Likelihood

## 1.1 Forward Algorithm

Let  $u \in \mathbb{R}^n$  be the initial distribution (which can be the stationary distribution  $\delta$ ),  $\Gamma$  the transition probability matrix and P(x) ....

Then the forward algorithm is defined as follows:

$$\alpha_0 := u$$

$$\alpha_i := \alpha_{i-1} \Gamma^t P(x) \quad \text{for } i \ge 1$$

The latter can be expressed as follows:

$$\alpha_i(k) = \sum_{i=1}^{m} \alpha_{i-1}(j) \Gamma^{d_i}(j,k) P_{(k,k)}(x_i)$$

We leverage the following identity:

$$log_b\left(\sum_{i=0}^{N} a_i\right) = log_b(a_0) + log_b\left(1 + \sum_{i=1}^{N} b^{log_b(a_i) - log_b(a_0)}\right)$$
(1)

In particular, we note that the expression on the right-hand side can be evaluated by *exclusively* relying on the logarithm of  $a_i$ .

Let  $\beta_{j,k,i} := \Gamma^{d_i}(j,k) P_{(k,k)}(x_i)$ . Note that neither  $\Gamma^{d_i}(j,k)$  nor  $P_{(k,k)}(x_i)$  depend on the position within the MC;  $\alpha_{i-1}(j)$  does, however.

Then, we have:

$$\alpha_i(k) = \sum_{j=1}^m \underbrace{\alpha_{i-1}(j)\,\beta_{j,k,i}}_{a_{j,k}} \tag{2}$$

By combining equations (2) and (1) we obtain

$$log_{b}\left(\sum_{j=0}^{m} a_{j}\right) = log_{b}\left(\alpha_{i-1}(0) \beta_{0,k,i}\right) + log_{b}\left(1 + \sum_{i=1}^{m} b^{log_{b}(\alpha_{i-1}(j) \beta_{j,k,i}) - log_{b}(\alpha_{i-1}(0) \beta_{0,k,i})}\right)$$

$$= log_{b}(\alpha_{i-1}(0)) + log_{b}(\beta_{0,k,i}) + log_{b}\left(1 + \sum_{i=1}^{m} b^{log_{b}(\alpha_{i-1}(j)) + log_{b}(\beta_{j,k,i}) - log_{b}(\alpha_{i-1}(0)) - log_{b}(\beta_{0,k,i})}\right)$$

$$(3)$$

Further setting

$$\tilde{\alpha}_i(k) := log(\alpha_i(k)) = log_b\left(\sum_{j=0}^m a_{j,k}\right)$$

we obtain

$$\tilde{\alpha}_{i}(k) = \tilde{\alpha}_{i-1}(0) + \log\left(\beta_{0,k,i}\right) + \log\left(1 + \sum_{j=1}^{m} b^{\tilde{\alpha}_{i-1}(j) + \log_{b}(\beta_{j,k,i}) - \tilde{\alpha}_{i-1}(0) - \log_{b}(\beta_{0,k,i})}\right)$$

w.l.o.g. the maximum element is the first (0).