

Supplementary Information for:

“HHMMiR: Efficient de novo Prediction of MicroRNAs using Hierarchical Hidden Markov Models”,

by

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APBC 2009 and *BMC Bioinformatics* 2009

Symbolism.

Finite alphabet:	Σ
Observed string:	$O = o_1 o_2 \dots o_N$ such that $o_i \in \Sigma$
Highest level of hierarchy (root):	1
Lowest level of hierarchy (leaves):	D
Depth of hierarchy:	$d \in \{1, \dots, D\}$
i^{th} state at hierarchical level d:	q_i^d
Number of substates of q_i^d :	$ q_i^d $
Parameters of HHMM:	

$$\lambda = \left\{ \lambda^{q^d} \right\}_{d \in \{1, \dots, D\}} = \left\{ \begin{array}{l} \left\{ A(q^d) \right\}_{d \in \{1, \dots, D\}}, \\ \left\{ \prod(q^d) \right\}_{d \in \{1, \dots, D\}}, \\ \left\{ E(q^D) \right\} \end{array} \right\}$$

1. Substate Transition Matrix:

$$\left\{ A(q^d) \right\}_{d \in \{1, \dots, D\}} \text{ such that } A(q^d) = \left(a_{jk}^{q^d} \right) = P\left(q_k^{d+1} | q_j^{d+1} \right)$$

$a_{jk}^{q^d}$ is the probability that the j^{th} substate of q^d will transition to its k^{th} substate.

2. Initial Substate distribution:

$$\{\Pi(q^d)\}_{d \in \{1, \dots, D\}} \text{ such that } \Pi(q^d) = \{\pi(q_j^{d+1} | q^d)\} = \{P(q_j^{d+1} | q^d)\}$$

$\pi(q_j^{d+1} | q^d)$ is the probability that q^d will make a vertical transition to its j^{th} substate at level $d+1$.

3. Output probability distribution:

$$\{E(q^D)\} \text{ such that } E(q^D, q^{D-1}) = \{e(\sigma_l | q^D, q^{D-1})\} = \{P(\sigma_l | q^D, q^{D-1})\}$$

$e(\sigma_l | q^D, q^{D-1})$ is the probability that production state q^D will emit symbol $\sigma_l \in \Sigma$.

Modified Baum Welch algorithm

Calculate the following probabilities:

1. Forward Probabilities

$$\alpha(t, t+k, q_i^{d+1}, q^d) = P(o_t \cdots o_{t+k}, q_i^{d+1} \text{ finished at } o_{t+k} | q^d \text{ started at } o_t)$$

Initialization:

Production states:

$$\alpha(t, t, q_i^D, q^{D-1}) = \pi(q_i^D | q^{D-1}) e(o_t | q_i^D, q^{D-1})$$

Internal States:

$$\alpha(t, t, q_i^d, q^{d-1}) = \pi(q_i^d | q^{d-1}) \left[\sum_{j=1}^{|q_i^d|} \alpha(t, t, q_j^{d+1}, q_i^d) \cdot a_{j \text{ end}}^{q_i^d} \right]$$

Iteration:

Production states:

$$\alpha(t, t+k, q_i^D, q^{D-1}) = \left[\sum_{j=1}^{|q^{D-1}|} \alpha(t, t+k-1, q_j^D, q^{D-1}) \right] e(o_{t+k} | q_i^D, q^{D-1})$$

Internal States:

$$\begin{aligned} \alpha(t, t+k, q_i^d, q^{d-1}) &= \sum_{l=0}^{k-1} \left[\sum_{j=1}^{|q^{d-1}|} \alpha(t, t+l, q_j^d, q^{d-1}) \cdot a_{ji}^{q_i^d} \right] \cdot \\ &\quad \left[\sum_{s=1}^{|q_i^d|} \alpha(t+l+1, t+k, q_s^{d+1}, q_i^d) \cdot a_{s \text{ end}}^{q_i^d} \right] \\ &\quad + \pi(q_i^d | q^{d-1}) \left[\sum_{j=1}^{|q_i^d|} \alpha(t, t+k, q_j^{d+1}, q_i^d) \cdot a_{j \text{ end}}^{q_i^d} \right] \end{aligned}$$

2. Backward Probabilities

$$\beta(t, t+k, q_i^d, q^{d-1}) = P(o_t \cdots o_{t+k} | q_i^d \text{ started at } o_t, q^{d-1} \text{ finished at } o_{t+k})$$

Initialization:

Production states:

$$\beta(t, t, q_i^D, q^{D-1}) = e(o_t | q_i^D, q^{D-1}) \cdot a_{i \text{ end}}^{q^{D-1}}$$

Internal States:

$$\beta(t, t, q_i^d, q^{d-1}) = \left[\sum_{j=1}^{|q_i^d|} \pi(q_j^{d+1} | q_i^d) \cdot \beta(t, t, q_j^{d+1}, q_i^d) \right] a_{i \text{ end}}^{q^{d-1}}$$

Iteration:

Production states:

$$\beta(t, t+k, q_i^D, q^{D-1}) = e(o_t | q_i^D, q^{D-1}) \left[\sum_{j \neq \text{end}}^{|q^{D-1}|} a_{ij}^{q^{D-1}} \cdot \beta(t+1, t+k, q_j^D, q^{D-1}) \right]$$

Internal States:

$$\begin{aligned} \beta(t, t+k, q_i^d, q^{d-1}) &= \sum_{l=0}^{k-1} \left[\sum_{j=1}^{|q_i^d|} \pi(q_j^{d+1} | q_i^d) \beta(t, t+l, q_j^{d+1}, q_i^d) \right] \cdot \\ &\quad \left[\sum_{s=1}^{|q^{d-1}|} a_{ij}^{q^{d-1}} \cdot \beta(t+l+1, t+k, q_s^d, q^{d-1}) \right] \\ &+ \left[\sum_{j=1}^{|q_i^d|} \pi(q_j^{d+1} | q_i^d) \cdot \beta(t, t+k, q_j^{d+1}, q_i^d) \cdot a_{i \text{end}}^{q^{d-1}} \right] \end{aligned}$$

3. Auxiliary variables:

A. $\eta_{in}(t, q_i^d, q^{d-1}) = P(o_1 \cdots o_{t-1}, q_i^d \text{ started at } o_t | \lambda)$

Initialization:

$$\eta_{in}(1, q_i^2, q^1) = \pi(q_i^2 | q^1)$$

$$\eta_{in}(1, q_i^d, q_j^{d-1}) = \eta_{in}(1, q_j^{d-1}, q^{d-2}) \cdot \pi(q_i^d | q_j^{d-1})$$

Iteration:

For $1 < t$

$$\eta_{in}(t, q_i^2, q^1) = \sum_{j=1}^{|q^1|} \alpha(1, t-1, q_j^2, q^1) a_{ji}^{q^1}$$

$$\begin{aligned} \eta_{in}(t, q_i^d, q_j^{d-1}) &= \sum_{s=1}^{t-1} \eta_{in}(s, q_j^{d-1}, q^{d-2}) \left[\sum_{l=1}^{|q_j^{d-1}|} \alpha(s, t-1, q_l^d, q_j^{d-1}) a_{li}^{q_j^{d-1}} \right] \\ &+ \eta_{in}(t, q_j^{d-1}, q^{d-2}) \cdot \pi(q_i^d | q_j^{d-1}) \end{aligned}$$

B. $\eta_{out}(t, q_i^d, q^{d-1}) = P(q_i^d \text{ finished at } o_t, o_{1t+1} \cdots o_N | \lambda)$

Initialization:

For $t < N$

$$\eta_{out}(t, q_i^2, q^1) = \sum_{j=1}^{|q^1|} a_{ij}^{q^1} \cdot \beta(t+1, N, q_j^2, q^1)$$

Iteration:

For $t < N$

$$\eta_{in}(t, q_i^d, q_j^{d-1}) = \sum_{k=t+1}^N \left[\sum_{l=1}^{|q_j^{d-1}|} a_{il}^{q_j^{d-1}} \beta(t+1, N, q_l^d, q_j^{d-1}) \right] \eta_{out}(k, q_j^{d-1}, q^{d-2})$$

$$+ a_{i \text{ end}}^{q_j^{d-1}} \cdot \eta_{out}(t, q_j^{d-1}, q^{d-2})$$

$$\eta_{out}(N, q_i^d, q_j^{d-1}) = a_{j \text{ end}}^{q_j^{d-1}} \cdot \eta_{out}(N, q_j^{d-1}, q^{d-2})$$

4. Horizontal Transition Probabilities

$$\xi(t, q_i^{d+1}, q_j^{d+1}, q^d) = P(o_1 \cdots o_t, q_i^{d+1} \rightarrow q_j^{d+1}, o_{t+1} \cdots o_N | \lambda)$$

Estimation:

For $t < N$

$$\xi(t, q_i^2, q_j^2, q^1) = \frac{\alpha(1, t, q_i^2, q^1) \cdot a_{ij}^{q^1} \cdot \beta(t+1, N, q_j^2, q^1)}{P(O|\lambda)}$$

$$\xi(N, q_i^2, q_j^2, q^1) = \frac{\alpha(1, N, q_i^2, q^1) \cdot a_{ij}^{q^1}}{P(O|\lambda)}$$

For $t < N$

$$\xi(t, q_i^d, q_j^d, q_l^{d-1}) = \frac{1}{P(O|\lambda)} \left[\sum_{s=1}^t \eta_{in}(s, q_l^{d-1}, q^{d-2}) \cdot \alpha(s, t, q_i^d, q_l^{d-1}) \right] a_{ij}^{q_l^{d-1}}$$

$$+ \left[\sum_{k=t+1}^N \beta(t+1, k | q_j^d, q_l^{d-1}) \cdot \eta_{out}(k, q_l^{d-1}, q^{d-2}) \right]$$

$$\xi(t, q_i^d, q_{end}^d, q_j^{d-1}) = \frac{1}{P(O|\lambda)} \left[\sum_{s=1}^t \eta_{in}(s, q_j^{d-1}, q^{d-2}) \cdot \alpha(s, t, q_i^d, q_j^{d-1}) \right] \\ a_{i_{end}}^{q_j^{d-1}} \cdot \eta_{out}(t, q_j^{d-1}, q^{d-2})$$

5. Vertical Transition Probabilities

$$\chi(t, q_i^d, q^{d-1}) = P(q_i^d \text{ started at } t | \lambda, O) \\ q^{d-1} \\ = P(o_1 \cdots o_{t-1}, \downarrow, o_t \cdots o_N | \lambda, O) \\ q_i^d$$

Initiation:

$$\chi(1, q_i^2, q^1) = \frac{\pi(q_i^2 | q^1) \cdot \beta(1, N, q_i^2, q^1)}{P(O|\lambda)}$$

Iteration:

For $2 < d$

$$\chi(t, q_i^d, q_j^{d-1}) = \frac{\eta_{in}(t, q_j^{d-1}, q^{d-2}) \cdot \pi(q_i^d | q_j^{d-1})}{P(O|\lambda)} \\ \left[\sum_{k=t}^N \beta(t, k, q_i^d, q_j^{d-1}) \cdot \eta_{out}(k, q_j^{d-1}, q^{d-2}) \right]$$

Parameter Estimation

1. $\gamma_{in}(t, q_i^{d+1}, q^d)$ is the probability of performing a horizontal transition to q_i^{d+1} which is

substate of q^d before o_t is emitted

$$\gamma_{in}(t, q_i^{d+1}, q^d) = \sum_{k=1}^{|q^d|} \xi(t-1, q_k^{d+1}, q_i^{d+1}, q^d)$$

2. $\gamma_{out}(t, q_i^{d+1}, q^d)$ is the probability of performing a horizontal transition from q_i^{d+1} which

is substate of q^d to any of the other substates of q^d after o_t is emitted

$$\gamma_{out}(t, q_i^{d+1}, q^d) = \sum_{k=1}^{|q^d|} \xi(t, q_i^{d+1}, q_k^{d+1}, q^d)$$

Thus,

$$\hat{\pi}(q_i^2 | q^1) = \chi(t, q_i^2, q^1)$$

$$\hat{\pi}(q_i^{d+1} | q^d) = \frac{\sum_{t=1}^T \chi(t, q_i^{d+1}, q^d)}{\sum_{i=1}^{|q^d|} \sum_{t=1}^T \chi(t, q_i^{d+1}, q^d)} \quad (1 < d < D-1)$$

$$\hat{a}_{jk}^{q^d} = \frac{\sum_{t=1}^N \xi(t, q_i^{d+1}, q_j^{d+1}, q^d)}{\sum_{k=1}^{|q^d|} \sum_{t=1}^N \xi(t, q_i^{d+1}, q_k^{d+1}, q^d)} = \frac{\sum_{t=1}^N \xi(t, q_i^{d+1}, q_j^{d+1}, q^d)}{\sum_{t=1}^N \gamma_{out}(t, q_i^{d+1}, q^d)}$$

$$\begin{aligned} \hat{e}(\sigma_l | q^D, q^{D-1}) &= \left(\sum_{o_t = \sigma_l} \chi(t, q_i^D, q^{D-1}) \right) \\ &+ \sum_{t > 1, o_t = \sigma_l} \gamma_{in}(t, q_i^D, q^{D-1}) / \left(\sum_{t=1}^T \chi(t, q_i^D, q^{D-1}) \right) \\ &+ \sum_{t=2}^T \gamma_{in}(t, q_i^D, q^{D-1}) \end{aligned}$$