

Untitled

A set of n genes $N \equiv \{i \in \{1, \dots, n\}\}$ is partially order accordingly to its relations as parent and offsprings such that for any $i, j \in N$ for which a directed pathlength exists, $i < j$ iff i is a parent node and j a descendant.

For leaf nodes we use missclassification probabilities ψ , for which we have 2^P different states, hence we obtain an $N \times 2^P$ array. Each row is filled with a particular combination of $\{0, 1\}^P$, e.g. $s = \{0, 0, 0\}$.

Each leaf is defined by a vector $z_l \equiv \{z_{lp}\}_{p=1}^P$ with

$$z_{lp} = \begin{cases} 1 & \text{if the function } p \text{ is active} \\ 0 & \text{if the function } p \text{ is not active} \\ 9 & \text{if we don't have information} \end{cases} \quad (1)$$

This way, given that the true state is $s \equiv \{s_p\}_{p=1}^P$, the probability of having correctly classified Z_l is:

$$\Psi_{ss'} = \Pr\{Z = s' \mid S = s\} = \begin{cases} \prod_p \{\psi_p^{\mathbf{1}\{s'_{lp}=s_{lp}\}} (1 - \psi_p)^{\mathbf{1}\{s'_{lp} \neq s_{lp}\}}\} & \text{if } s'_{lp} \neq 9 \\ 1 & \text{otherwise} \end{cases} \quad (2)$$

For internal node i we use gain and loss functions, also defined by a $N \times 2^P$ array. This is conditional on i 's O_i offsprings $o_i \subset \{j \in N : j < i\}$ and the true state s . Let $x_i \equiv \{x_{ip}\}_{p=1}^P$ denote the vector of functional states of i , then, the probability that we observe $X = x_i$ conditional on the true state of its offsprings $j \in o_i$ is:

$$\Pr\{\mu, \psi\}_{n,s} = \prod_{o_n} \sum_{s'_n} \Psi_{s'_n s'} \prod_p \left(\underbrace{\left[\begin{array}{c} \mu_0^{\mathbf{1}\{s'_{np}\}} \\ \text{Gain} \end{array} \right]}_{\text{Gain}} \underbrace{(1 - \mu_0)^{\mathbf{1}\{\neg s'_{np}\}}}_{\text{No gain}} \right)^{\mathbf{1}\{\neg s_p\}} \times \left(\underbrace{\left[\begin{array}{c} \mu_1^{\mathbf{1}\{\neg s'_{np}\}} \\ \text{Loss} \end{array} \right]}_{\text{Loss}} \underbrace{(1 - \mu_1)^{\mathbf{1}\{s'_{np}\}}}_{\text{No Loss}} \right)^{\mathbf{1}\{s_p\}} \quad (3)$$

Computationally, observe that the larger parenthesis can be computed only once and then retrieved depending on the values of $\{s'_{np}, s_p\}$. Let $M \equiv \{m_{s_n, s}\}$ to be an array of size 2×2 holding the Gain/Loss probabilities, then, the previous equation reduces to:

$$\Pr\{\mu, \psi\}_{n,s} = \prod_{o_n} \sum_{s'_n} \Psi_{s'_n s'} \prod_p m_{s'_{np}, s} \quad (4)$$

Finally, define $\pi \equiv \{\pi_0, \pi_1\}$ to be the root node state probabilities, then, the likelihood function can be computed as

$$L(\pi, \mu, \Psi) = \sum_s \pi_s \Pr\{\mu, \psi\}_{n,s} \quad (5)$$