

Sample Average Approximation for Optimizing Probability of Improvement and Expectation of Improvement

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Let $\theta^{(k)}$ represent the full set of Naive Bayes parameters, $\theta_{y,i}^{(k)}(j)$, for $y = 0, 1$ being whether the peptide is a substrate for enzyme k , $i = -20, \dots, -1, 1, \dots, 20$ being position of amino-acids in the peptide and $j = 1, \dots, 8$ denoting the class each amino-acid belongs to. Here k denotes different types of enzyme, and in our case $k = 1, 2$.

We have

$$P(Y(x, k) = 1 | \theta^{(k)}) = \frac{P(Y(x, k) = 1) \prod_i \theta_{1,i}^{(k)}(x_i)}{P(Y(x, k) = 1) \prod_i \theta_{1,i}^{(k)}(x_i) + P(Y(x, k) = 0) \prod_i \theta_{0,i}^{(k)}(x_i)}$$

$$P(Y(x, k) = 0 | \theta^{(k)}) = \frac{P(Y(x, k) = 0) \prod_i \theta_{0,i}^{(k)}(x_i)}{P(Y(x, k) = 1) \prod_i \theta_{1,i}^{(k)}(x_i) + P(Y(x, k) = 0) \prod_i \theta_{0,i}^{(k)}(x_i)}$$

Therefore,

$$\frac{P(Y(x, k) = 1 | \theta^{(k)})}{P(Y(x, k) = 0 | \theta^{(k)})} \propto \prod_i \frac{\theta_{1,i}^{(k)}(x_i)}{\theta_{0,i}^{(k)}(x_i)}$$

We let

$$\eta_i^{(k)}(x_i) = \frac{\theta_{1,i}^{(k)}(x_i)}{\theta_{0,i}^{(k)}(x_i)}$$

Then we have

$$P(Y(x, k) = 1 | \theta^{(k)}) = \frac{\prod_i \eta_i^{(k)}(x_i)}{1 + \prod_i \eta_i^{(k)}(x_i)}$$

Now, suppose we're given a set of N peptides: $S = (X)_{n=1}^n$, then the probability that all peptides in S are not substrate for enzyme k is:

$$\begin{aligned}
P(Y(X_1, k) = 0, \dots, Y(X_N, k) = 0) &= E_{\theta^{(k)}}[P(Y(X_1, k) = 0, \dots, Y(X_N, k) = 0 | \theta^{(k)})] \\
&= E_{\theta^{(k)}}[\prod_n P(Y(X_n, k) = 0 | \theta^{(k)})] \\
&= E_{\theta^{(k)}}[\prod_n \frac{1}{1 + \prod_i \eta_i^{(k)}(x_i)}]
\end{aligned}$$

The second equality follows from that given $\theta^{(k)}$, the probability each peptide in S is a substrate for enzyme k is independent. To estimate the above expectation, we can use the sample average approximation approach. We first simulate a set of L different $\eta^{(k)}$ parameters: $\{\eta_{i,l}^{(k)}(j) : i = -20, \dots, -1, 1, \dots, 20, j = 1, \dots, 8\}_{l=1}^L$, then we estimate the above expectation as

$$\hat{E}_{\theta^{(k)}}[\prod_n \frac{1}{1 + \prod_i \eta_i^{(k)}(x_i)}] = \frac{1}{L} \sum_{l=1}^L \frac{1}{\prod_n [1 + \prod_i \eta_{i,l}^{(k)}(x_i)]}$$

To increase the probability of improvement, we should minimize the above the expectation. If we introduce a new set of variables z :

$$z_{i,j}^n = \begin{cases} 1 & \text{if } x_{n,i} = j \\ 0 & \text{otherwise} \end{cases}$$

we're facing an optimization problem:

$$\begin{aligned}
\min_z \quad & \frac{1}{L} \sum_{l=1}^L \frac{1}{\prod_n [1 + \prod_i (\sum_j \eta_{i,l}^{(k)}(j) z_{i,j}^n)]} \\
\text{s.t.} \quad & \sum_{j=1}^8 z_{i,j}^n = 1 \quad i = -20, \dots, -1, 1, \dots, 20, n = 1, \dots, N \\
& z_{i,j}^n \in \{0, 1\} \quad i = -20, \dots, -1, 1, \dots, 20, j = 1, \dots, 8, n = 1, \dots, N
\end{aligned}$$