

The Model:

P neurons $\{f_1, \dots, f_P\}$, each discretized with M parameters $(\theta_1, \dots, \theta_M)$.

$r_k | \theta \sim \text{Poisson}(f_{k,j})$, $k = 1, \dots, P$, $j = 1, \dots, M$, if $\theta \in I_j$.

Can think of f_k as a piecewise constant tuning curve with each piece length = w_j :

$$f_k(\theta) = \sum_{j=1}^M f_{k,j} 1_{\{\theta \in I_j\}}$$

where $|I_j| = w_j$, $I_j = [\sum_{l=1}^{j-1} w_l, \sum_{l=1}^j w_l)$, $\sum_{l=1}^M w_l = 1$.

The stimulus distribution is uniform, so:

$$p(\theta) = \sum_{j=1}^M w_j 1_{\{\theta \in I_j\}}$$

Thus

$$\begin{aligned} P(r_k | \theta) &= \sum_j 1_{\{\theta \in I_j\}} \frac{f_{k,j}^{r_k}}{r_k!} e^{-f_{k,j}} \\ P(\mathbf{r} | \theta) &= \sum_j 1_{\{\theta \in I_j\}} \prod_{k=1}^P \frac{f_{k,j}^{r_k}}{r_k!} e^{-f_{k,j}} \\ P(\mathbf{r}) &= \int p(\theta) d\theta P(\mathbf{r} | \theta) = \sum_{j=1}^M w_j P(\mathbf{r} | \theta_j) \end{aligned}$$

The Mutual Information:

$$\begin{aligned} I(\mathbf{r}; \theta) &= \int d\theta p(\theta) D_{KL}(p(\mathbf{r} | \theta) \| p(\mathbf{r})) \\ &= \sum_{j=1}^M w_j D_{KL}(p(\mathbf{r} | \theta_j) \| p(\mathbf{r})) \\ &= \sum_{j=1}^M w_j \left[- \sum_{\mathbf{r}} P_j(\mathbf{r}) \ln S_j(\mathbf{r}) \right] \\ &= \sum_{j=1}^M w_j E_{\mathbf{r} | \theta_j} [-\ln S_j(\mathbf{r})] \end{aligned} \tag{1}$$

where

$$\begin{aligned} P_j(\mathbf{r}) &:= P(\mathbf{r} | \theta_j) = \prod_{k=1}^P \frac{f_{k,j}^{r_k}}{r_k!} e^{-f_{k,j}} \\ S_j(\mathbf{r}) &:= \frac{p(\mathbf{r})}{P(\mathbf{r} | \theta_j)} = \sum_{l=1}^M w_l \prod_{k=1}^P \frac{f_{k,l}^{r_k} e^{-f_{k,l}}}{f_{k,j}^{r_k} e^{-f_{k,j}}} \end{aligned}$$

Derivatives:

$$(x^r e^{-x})' = \left(\frac{r}{x} - 1\right) x^r e^{-x}, \quad (x^{-r} e^x)' = -\left(\frac{r}{x} - 1\right) x^{-r} e^{-x}$$

$$\begin{aligned}
\frac{\partial P_j(\mathbf{r})}{\partial f_{k,l}} &= 1_{\{l=j\}} \left(\frac{r_k}{f_{k,l}} - 1 \right) P_j(\mathbf{r}) \\
\frac{\partial S_j(\mathbf{r})}{\partial f_{k,l}} &= \sum_{l'=1}^M w_{l'} \frac{\partial}{\partial f_{k,l}} \left(\prod_{k=1}^P \frac{f_{k,l'}^{r_{k,l'}} e^{-f_{k,l'}}}{f_{k,j}^{r_k} e^{-f_{k,j}}} \right) \\
&= \sum_{l'=1}^M w_{l'} \prod_{k' \neq k} \frac{f_{k',l'}^{r_{k',l'}} e^{-f_{k',l'}}}{f_{k',j}^{r_{k',j}} e^{-f_{k',j}}} \frac{\partial}{\partial f_{k,l}} \left(\frac{f_{k,l'}^{r_{k,l'}} e^{-f_{k,l'}}}{f_{k,j}^{r_k} e^{-f_{k,j}}} \right) \\
&= \sum_{l'=1}^M w_{l'} \prod_{k' \neq k} \frac{f_{k',l'}^{r_{k',l'}} e^{-f_{k',l'}}}{f_{k',j}^{r_{k',j}} e^{-f_{k',j}}} \left(1_{\{l=l'\}} \left(\frac{r_k}{f_{k,l'}} - 1 \right) \frac{f_{k,l'}^{r_{k,l'}} e^{-f_{k,l'}}}{f_{k,j}^{r_k} e^{-f_{k,j}}} + 1_{\{l=j\}} \left(-\frac{r_k}{f_{k,j}} + 1 \right) \frac{f_{k,l'}^{r_{k,l'}} e^{-f_{k,l'}}}{f_{k,j}^{r_k} e^{-f_{k,j}}} + 1_{\{l'=j\}} 0 \right) \\
&= w_l \left(\frac{r_k}{f_{k,l}} - 1 \right) \prod_{k'=1}^P \frac{f_{k',l}^{r_{k',l}} e^{-f_{k',l}}}{f_{k',j}^{r_{k',j}} e^{-f_{k',j}}} + 1_{\{l=j\}} \left(-\frac{r_k}{f_{k,j}} + 1 \right) \sum_{l'=1}^M w_{l'} \prod_{k'=1}^P \frac{f_{k',l'}^{r_{k',l'}} e^{-f_{k',l'}}}{f_{k',j}^{r_{k',j}} e^{-f_{k',j}}} \\
&= w_l \left(\frac{r_k}{f_{k,l}} - 1 \right) \prod_{k'=1}^P \frac{f_{k',l}^{r_{k',l}} e^{-f_{k',l}}}{f_{k',j}^{r_{k',j}} e^{-f_{k',j}}} + 1_{\{l=j\}} \left(-\frac{r_k}{f_{k,j}} + 1 \right) S_j(\mathbf{r})
\end{aligned}$$

$$\begin{aligned}
\frac{\partial}{\partial f_{k,l}} \left(\sum_{\mathbf{r}} P_j(\mathbf{r}) \ln S_j(\mathbf{r}) \right) &= \sum_{\mathbf{r}} \frac{\partial P_j(\mathbf{r})}{\partial f_{k,l}} \ln S_j(\mathbf{r}) + \sum_{\mathbf{r}} P_j(\mathbf{r}) \frac{1}{S_j(\mathbf{r})} \frac{\partial S_j(\mathbf{r})}{\partial f_{k,l}} \\
&= 1_{\{l=j\}} \sum_{\mathbf{r}} \left(\frac{r_k}{f_{k,l}} - 1 \right) P_j(\mathbf{r}) \ln S_j(\mathbf{r}) + \sum_{\mathbf{r}} P_j(\mathbf{r}) \frac{1}{S_j(\mathbf{r})} \frac{\partial S_j(\mathbf{r})}{\partial f_{k,l}} \\
&= 1_{\{l=j\}} E_{\mathbf{r}|\theta_j} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) \ln S_j(\mathbf{r}) \right] + E_{\mathbf{r}|\theta_j} \left[\frac{1}{S_j(\mathbf{r})} \frac{\partial S_j(\mathbf{r})}{\partial f_{k,l}} \right] \\
&= 1_{\{l=j\}} E_{\mathbf{r}|\theta_j} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) \ln S_j(\mathbf{r}) \right] \\
&\quad + E_{\mathbf{r}|\theta_j} \left[w_l \left(\frac{r_k}{f_{k,l}} - 1 \right) \frac{\prod_{k'=1}^P \frac{f_{k',l}^{r_{k',l}} e^{-f_{k',l}}}{f_{k',j}^{r_{k',j}} e^{-f_{k',j}}}}{\sum_{l=1}^M w_l \prod_{k=1}^P \frac{f_{k,l}^{r_{k,l}} e^{-f_{k,l}}}{f_{k,j}^{r_k} e^{-f_{k,j}}}} \right] \\
&\quad + 1_{\{l=j\}} E_{\mathbf{r}|\theta_j} \left[\left(-\frac{r_k}{f_{k,j}} + 1 \right) \frac{S_j(\mathbf{r})}{S_j(\mathbf{r})} \right] \\
&= 1_{\{l=j\}} E_{\mathbf{r}|\theta_j} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) \ln S_j(\mathbf{r}) \right] + 1_{\{l=j\}} E_{\mathbf{r}|\theta_j} \left(-\frac{r_k}{f_{k,j}} \right) + 1 \\
&\quad + E_{\mathbf{r}|\theta_j} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) \frac{w_l \prod_{k'=1}^P \frac{f_{k',l}^{r_{k',l}} e^{-f_{k',l}}}{f_{k',j}^{r_{k',j}} e^{-f_{k',j}}}}{\sum_{l'=1}^M w_{l'} \prod_{k'=1}^P \frac{f_{k',l'}^{r_{k',l'}} e^{-f_{k',l'}}}{f_{k',j}^{r_{k',j}} e^{-f_{k',j}}}} \right] \\
&= 1_{\{l=j\}} E_{\mathbf{r}|\theta_j} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) \ln S_j(\mathbf{r}) \right] + E_{\mathbf{r}|\theta_j} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) \frac{w_l \prod_{k'=1}^P \frac{f_{k',l}^{r_{k',l}} e^{-f_{k',l}}}{f_{k',j}^{r_{k',j}} e^{-f_{k',j}}}}{\sum_{l'=1}^M w_{l'} \prod_{k'=1}^P \frac{f_{k',l'}^{r_{k',l'}} e^{-f_{k',l'}}}{f_{k',j}^{r_{k',j}} e^{-f_{k',j}}}} \right]
\end{aligned}$$

$$\begin{aligned}
\frac{\partial I(\mathbf{r}; \theta)}{\partial f_{k,l}} &= - \sum_j w_j \sum_{\mathbf{r}} \frac{\partial}{\partial f_{k,l}} \left(\sum_{\mathbf{r}} P_j(\mathbf{r}) \ln S_j(\mathbf{r}) \right) \\
&= - \sum_{j=1}^M w_j 1_{\{l=j\}} E_{\mathbf{r}|\theta_j} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) \ln S_j(\mathbf{r}) \right] \\
&\quad - \sum_{j=1}^M w_j E_{\mathbf{r}|\theta_j} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) \frac{w_l \prod_{k'=1}^P f_{k',l}^{r_{k'}} e^{-f_{k',l}}}{\sum_{l'=1}^M w_{l'} \prod_{k'=1}^P f_{k',l'}^{r_{k'}} e^{-f_{k',l'}}} \right] \\
&= -w_l E_{\mathbf{r}|\theta_l} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) \ln S_l(\mathbf{r}) \right] - \sum_{j=1}^M w_j E_{\mathbf{r}|\theta_j} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) \frac{w_l \prod_{k'=1}^P f_{k',l}^{r_{k'}} e^{-f_{k',l}}}{\sum_{l'=1}^M w_{l'} \prod_{k'=1}^P f_{k',l'}^{r_{k'}} e^{-f_{k',l'}}} \right] \tag{2}
\end{aligned}$$

In conclusion:

$$I(\mathbf{r}; \theta) = \sum_{j=1}^M w_j E_{\mathbf{r}|\theta_j} [-\ln S_j(\mathbf{r})] \tag{3}$$

$$\frac{\partial I(\mathbf{r}; \theta)}{\partial f_{k,l}} = -w_l E_{\mathbf{r}|\theta_l} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) \ln S_l(\mathbf{r}) \right] - \sum_{j=1}^M w_j E_{\mathbf{r}|\theta_j} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) \frac{w_l \prod_{k'=1}^P f_{k',l}^{r_{k'}} e^{-f_{k',l}}}{\sum_{l'=1}^M w_{l'} \prod_{k'=1}^P f_{k',l'}^{r_{k'}} e^{-f_{k',l'}}} \right] \tag{4}$$

where

$$\begin{aligned}
S_j(\mathbf{r}) &= \sum_{l=1}^M w_l \prod_{k=1}^P \frac{f_{k,l}^{r_k} e^{-f_{k,l}}}{f_{k,j}^{r_k} e^{-f_{k,j}}} \\
&= \sum_{l=1}^M w_l \exp \sum_{k=1}^P \log \left(\frac{f_{k,l}^{r_k} e^{-f_{k,l}}}{f_{k,j}^{r_k} e^{-f_{k,j}}} \right) \\
&= \sum_{l=1}^M w_l \exp \sum_{p=1}^P \left[r_p \log \left(\frac{f_{p,l}}{f_{p,j}} \right) + (f_{p,j} - f_{p,l}) \right]
\end{aligned}$$

Constraints:

$$f_+ \leq f_{k,l} \leq f_-$$

$$\sum_{i=1}^M w_i f_{k,i} = \bar{f}_k$$

If there is convolution:

$$\begin{aligned}
g_{k,i} &= \sum_{l=1}^M f_{k,l} s_{i-l} \\
\sum_{i=1}^M w_i g_{k,i} &= \sum_{i=1}^M \sum_{l=1}^M f_{k,l} s_{i-l} = \sum_{i=1}^M w_i \sum_{l=1}^M f_{k,l} s_{i-l} = \sum_{l=1}^M w_l s_{i-l} \sum_i f_{k,l}
\end{aligned}$$

?, where

$$\int s(\theta) d\theta = \sum_{i=1}^M w_i s_i = 1$$

$$\frac{\partial I(\mathbf{r}; \theta)}{\partial f_{k,l}} = \sum_{j=1}^M \frac{\partial I(\mathbf{r}; \theta)}{\partial g_{k,j}} s_{j-l}$$

Notes for the code

$k, p = 1, \dots, P, i, j, l, m = 1, \dots, M$

$$\begin{aligned}
lrate_m(p, i) &:= \ln \left(\frac{f_{p,i}}{f_{p,m}} \right) \\
dexp_m(i) &:= \sum_{p=1}^P (f_{p,m} - f_{p,i}) \\
mexp_m(i) &:= \sum_{p=1}^P r_p lrate_m(p, i) + dexp_m(i) \\
Max_m &:= \max_i mexp_m(i) \\
prodq_m(i) = Q_m(i) &:= \exp [mexp_m(i) - Max] = \prod_{k=1}^P \frac{f_{k,i}^{r_k} e^{-f_{k,i}}}{f_{k,m}^{r_k} e^{-f_{k,m}}} e^{-Max} \\
S_m(\mathbf{r}) &= \sum_{i=1}^M w_i \exp [mexp(i)] \\
&= \sum_{i=1}^M w_i \exp [mexp(i) - Max] \cdot e^{Max} \\
mean_m &:= \ln (S_m(\mathbf{r})) \text{ where } \mathbf{r} \sim P(\mathbf{r}|\theta_m)
\end{aligned}$$

$$\begin{aligned}
grad1(p, i) &= E_{\mathbf{r}|\theta_i} \left[\left(\frac{r_k}{f_{k,i}} - 1 \right) \ln S_i(\mathbf{r}) \right], \mathbf{r} \sim P(\mathbf{r}|\theta_i) \\
tmpgrad2_m(p, i) &= E_{\mathbf{r}|\theta_m} \left[\left(\frac{r_p}{f_{p,i}} - 1 \right) \frac{w_i Q_m(i)}{\sum_{l'=1}^M w_{l'} Q_m(l')} \right], \mathbf{r} \sim P(\mathbf{r}|\theta_m) \\
&= E_{\mathbf{r}|\theta_m} \left[\left(\frac{r_p}{f_{p,i}} - 1 \right) \frac{w_i \exp [mexp_m(i) - Max]}{\sum_{l'=1}^M w_{l'} \exp [mexp_m(l') - Max]} \right] \\
grad(p, i) &= -w_i grad1(p, i) - \sum_{m=1}^M w_m tmpgrad2_m(p, i) \\
I &= -\sum_{m=1}^M w_m mean_m
\end{aligned}$$

$$I(\mathbf{r}; \theta) = \sum_{j=1}^M w_j E_{\mathbf{r}|\theta_j} [-\ln S_j(\mathbf{r})]$$

$$\frac{\partial I(\mathbf{r}; \theta)}{\partial f_{k,l}} = -w_l E_{\mathbf{r}|\theta_l} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) \ln S_l(\mathbf{r}) \right] - \sum_{j=1}^M w_j E_{\mathbf{r}|\theta_j} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) \frac{w_l \prod_{k'=1}^P f_{k',l}^{r_{k'}} e^{-f_{k',l}}}{\sum_{l'=1}^M w_{l'} \prod_{k'=1}^P f_{k',l'}^{r_{k'}} e^{-f_{k',l'}}} \right]$$

Constraints:?

The Mutual Information:

$$\begin{aligned}
I(\mathbf{r}; \theta) &= \sum_{j=1}^M w_j \left[- \sum_{\mathbf{r}} P_j(\mathbf{r}) \ln S_j(\mathbf{r}) \right] \\
&\approx \sum_{r_1=0}^T \cdots \sum_{r_P=0}^T \left[- \sum_{j=1}^M w_j P_j(\mathbf{r}) \ln S_j(\mathbf{r}) \right]
\end{aligned}$$

where

$$\begin{aligned}
P_j(\mathbf{r}) &:= P(\mathbf{r}|\theta_j) = \prod_{k=1}^P \frac{f_{k,j}^{r_k}}{r_k!} e^{-f_{k,j}} \\
S_j(\mathbf{r}) &:= \frac{p(\mathbf{r})}{P(\mathbf{r}|\theta_j)} = \sum_{l=1}^M w_l \prod_{k=1}^P \frac{f_{k,l}^{r_k} e^{-f_{k,l}}}{f_{k,j}^{r_k} e^{-f_{k,j}}}
\end{aligned}$$

$$\begin{aligned}
P_j(\mathbf{r}) \ln S_j(\mathbf{r}) &= \prod_{k=1}^P \frac{f_{k,j}^{r_k}}{r_k!} e^{-f_{k,j}} \ln \left[\sum_{l=1}^M w_l \prod_{k=1}^P \frac{f_{k,l}^{r_k} e^{-f_{k,l}}}{f_{k,j}^{r_k} e^{-f_{k,j}}} \right] \\
&= \exp \left(\sum_{k=1}^P (r_k \log(f_{k,j}) - f_{k,j} - \log(r_k!)) \right) \cdot \ln \left[\sum_{l=1}^M w_l \exp \sum_{k=1}^P (r_k \log(f_{k,l}) - f_{k,l} - (r_k \log(f_{k,j}) - f_{k,j})) \right] \\
&= \exp \left(\sum_{k=1}^P (\text{exp}_{k,j} - \log(r_k!)) \right) \ln \left[\sum_{l=1}^M w_l \exp \sum_{k=1}^P (\text{exp}_{k,l} - \text{exp}_{k,j}) \right]
\end{aligned}$$

$$\begin{aligned}
\frac{\partial I(\mathbf{r}; \theta)}{\partial f_{k,l}} &= -w_l E_{\mathbf{r}|\theta_l} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) \ln S_l(\mathbf{r}) \right] - \sum_{j=1}^M w_j E_{\mathbf{r}|\theta_j} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) \frac{w_l \prod_{k'=1}^P f_{k',l}^{r_{k'}} e^{-f_{k',l}}}{\sum_{l'=1}^M w_{l'} \prod_{k'=1}^P f_{k',l'}^{r_{k'}} e^{-f_{k',l'}}} \right] \\
&= - \sum_{\mathbf{r}} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) w_l P_l(\mathbf{r}) \ln S_l(\mathbf{r}) \right] - \sum_{\mathbf{r}} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) \sum_{j=1}^M w_j P_j(\mathbf{r}) \frac{w_l \prod_{k'=1}^P f_{k',l}^{r_{k'}} e^{-f_{k',l}}}{\sum_{l'=1}^M w_{l'} \prod_{k'=1}^P f_{k',l'}^{r_{k'}} e^{-f_{k',l'}}} \right] \\
&= - \sum_{\mathbf{r}} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) w_l P_l(\mathbf{r}) \ln S_l(\mathbf{r}) \right] - \sum_{\mathbf{r}} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) \sum_{j=1}^M w_j P_j(\mathbf{r}) \frac{w_l}{S_l(\mathbf{r})} \right] \\
&= - \sum_{\mathbf{r}} \left[\left(\frac{r_k}{f_{k,l}} - 1 \right) \left(w_l P_l(\mathbf{r}) \ln S_l(\mathbf{r}) + \sum_{j=1}^M w_j P_j(\mathbf{r}) \frac{w_l}{S_l(\mathbf{r})} \right) \right]
\end{aligned}$$

The Arimoto Algorithm:

$$\max_{p(\theta)} I(\mathbf{r}; \theta) \Leftrightarrow \max_{p(\theta)} \max_{q(\theta|\mathbf{r})} \int d\mathbf{r} \int d\theta p(\theta) p(\mathbf{r}|\theta) \log \frac{q(\theta|\mathbf{r})}{p(\theta)}$$

At capacity, $D_{KL}(p(\mathbf{r}|\theta)||p(\mathbf{r})) = \text{const}$, for all θ .

Iteration:

$$q^{(n+1)}(\theta_j|\mathbf{r}) \propto p^{(n)}(\theta_j) p(\mathbf{r}|\theta_j) \propto w_j^{(n)} \prod_{k=1}^P f_{k,j}^{r_k} e^{-f_{k,j}}$$

$$\begin{aligned} p^{(n+1)}(\theta_j) &\propto \exp \left[\sum_{\mathbf{r}} p(\mathbf{r}|\theta_j) \log q^{(n+1)}(\theta_j|\mathbf{r}) \right] \\ &\propto \exp \left[E_{\mathbf{r}|\theta_j} \log \frac{w_j^{(n)} \prod_{k=1}^P f_{k,j}^{r_k} e^{-f_{k,j}}}{\sum_l w_l^{(n)} \prod_{k=1}^P f_{k,l}^{r_k} e^{-f_{k,l}}} \right] \end{aligned}$$

Codes notes: $\mathbf{r} \sim P(\mathbf{r}|\theta_m)$,

$$\frac{w_j^{(n)} \prod_{k=1}^P f_{k,j}^{r_k} e^{-f_{k,j}}}{\sum_l w_l^{(n)} \prod_{k=1}^P f_{k,l}^{r_k} e^{-f_{k,l}}} = \frac{w_j^{(n)} \exp \sum_{k=1}^P (r_k \log(f_{k,j}) - f_{k,j})}{\sum_l w_l^{(n)} \exp \sum_{k=1}^P (r_k \log(f_{k,l}) - f_{k,l})}$$

$$\begin{aligned} \text{rexp}_m(i) &:= \sum_{p=1}^P (r_p \log f_{p,i} - f_{p,i}) \\ \text{Max}_m &:= \max_i \text{rexp}_m(i) \\ \text{qexp}_m &:= \frac{w_m \exp[\text{rexp}_m(m) - \text{Max}]}{\sum_{l=1}^M w_l \exp[\text{rexp}_m(l) - \text{Max}]} \\ \text{mean}_m &:= \ln(\text{qexp}_m) \\ \text{mean}_m &= \frac{1}{N_{\text{iter}}} \text{mean}_m^{(N)} \\ p(\theta_m) &\propto \exp(\text{mean}_m) \end{aligned}$$

$$\begin{aligned} p^{(n+1)}(\theta_j) &\propto \exp \left[E_{\mathbf{r}|\theta_j} \log \frac{w_j^{(n)} \prod_{k=1}^P f_{k,j}^{r_k} e^{-f_{k,j}}}{\sum_l w_l^{(n)} \prod_{k=1}^P f_{k,l}^{r_k} e^{-f_{k,l}}} \right] \\ &\propto \exp \left[\sum_{\mathbf{r}} P_j(\mathbf{r}) \log \left(\frac{w_j^{(n)}}{S_j^{(n)}(\mathbf{r})} \right) \right] \end{aligned}$$

$$\begin{aligned} P_j(\mathbf{r}) &:= P(\mathbf{r}|\theta_j) = \prod_{k=1}^P \frac{f_{k,j}^{r_k}}{r_k!} e^{-f_{k,j}} \\ S_j^{(n)}(\mathbf{r}) &:= \frac{p(\mathbf{r})}{P(\mathbf{r}|\theta_j)} = \sum_{l=1}^M w_l^{(n)} \prod_{k=1}^P \frac{f_{k,l}^{r_k} e^{-f_{k,l}}}{f_{k,j}^{r_k} e^{-f_{k,j}}} \end{aligned}$$