

$\hat{\tau}_k$: Optimum value of τ_k

$$\hat{\tau}_k = \arg \max_{\tau_k} \left[\underbrace{\ln p(b|s^2) + \eta \sum_k \tau_k - \eta}_{\mathcal{L}(\tau)} \right] \quad \left| \begin{array}{l} \text{constraint } \sum_k \tau_k = 1 \\ \text{use} \\ \text{Lagrange multiplier.} \end{array} \right.$$

$$\begin{aligned} \ln p(b|s^2) &= \ln \prod_{j=1}^P \int p(b_j | w_j, s^2) p(w_j) dw_j \\ &= \sum_{j=1}^P \ln \left(\sum_{k=1}^K \tau_k \phi_{jk} \right) \end{aligned}$$

$$\begin{aligned} \frac{\partial}{\partial \tau_k} \mathcal{L}(\tau) &= \frac{\partial}{\partial \tau_k} \left[\sum_{j=1}^P \ln \left(\sum_{k=1}^K \tau_k \phi_{jk} \right) + \eta \sum_k \tau_k - \eta \right] \\ &= \sum_{j=1}^P \frac{\phi_{jk}}{\sum_k \tau_k \phi_{jk}} + \eta = \frac{1}{\tau_k} \sum_{j=1}^P \frac{\tau_k \phi_{jk}}{\sum_k \tau_k \phi_{jk}} + \eta = 0 \quad (\text{at optimum}) \quad \textcircled{1} \end{aligned}$$

There is no closed form solution, but we can cheat a little and assume $\ln \frac{\tau_k \phi_{jk}}{\sum_k \tau_k \phi_{jk}}$ is constant, which yields

$$\hat{\tau}_k = -\frac{1}{\eta} \sum_{j=1}^P \frac{\tau_k \phi_{jk}}{\sum_k \tau_k \phi_{jk}}$$

Summing over k in Eq. ① gives $\eta = -P$

$$\Rightarrow \hat{\tau}_k = +\frac{1}{P} \sum_{j=1}^P \frac{\tau_k \phi_{jk}}{\sum_k \tau_k \phi_{jk}}$$

☑ Point mass at λ_k . ~~Prior~~

Prior.
$$p_k(w) = \prod_{j=1}^p p_k(w_j) = \prod_{j=1}^p \delta(w_j - \lambda_k) = \delta(w - \mathbb{I} \lambda_k)$$

Posterior.
$$p_k(w_j | b_j, s^2) = \frac{p(b_j | w_j, s^2) \delta(w_j - \lambda_k)}{p(b_j | \lambda_k, s^2)} = \frac{p(b_j | w_j, s^2) p_k(w_j)}{\phi_{jk}}$$

☑ Mixture of point mass.

$$\left[\begin{aligned} \phi_{jk} &:= p(b_j | \lambda_k, s^2) \\ &= \mathcal{N}(b_j | 0, s^2 \lambda_k) \end{aligned} \right.$$

Prior.
$$p(w_j) = \sum_{k=1}^K \tau_k p_k(w_j) = \sum_{k=1}^K \tau_k \delta(w_j - \lambda_k)$$

Posterior.
$$p(w_j | b_j, s^2) = \frac{p(b_j | w_j, s^2) p(w_j)}{\int p(b_j | w_j, s^2) p(w_j) dw_j} = \frac{p(b_j | w_j, s^2) p(w_j)}{\phi_j}$$

$$= \frac{\sum_{k=1}^K \tau_k p_k(w_j) p(b_j | w_j, s^2) \phi_{jk}}{\phi_j}$$

$$\left[\phi_j := \int p(b_j | s^2) \right.$$

$$= \frac{\sum_{k=1}^K \tau_k \phi_{jk} p_k(w_j | b_j, s^2)}{\phi_j}$$

$$= \sum_{k=1}^K \tau'_k p_k(w_j | b_j, s^2)$$

$$\text{where } \tau'_k = \frac{\tau_k \phi_{jk}}{\phi_j}$$

Now,
$$\phi_j = \int p(b_j | w_j, s^2) \sum_{k=1}^K \tau_k p_k(w_j) dw_j$$

$$= \sum_{k=1}^K \tau_k p(b_j | \lambda_k, s^2) = \sum_{k=1}^K \tau_k \phi_{jk} \Rightarrow$$

$$\boxed{\tau'_k = \frac{\tau_k \phi_{jk}}{\sum_{k=1}^K \tau_k \phi_{jk}}}$$

We can now compute the expectation of any function of w_j :

$$\mathbb{E}[f(w_j)] = \int f(w_j) p(w_j | b_j, s^2) dw_j = \sum_{k=1}^K \tau'_k \int f(w_j) p_k(w_j | b_j, s^2) dw_j$$

Here, we used the definition of $p_k(w_j | b_j, s^2)$ from point-mass prior.

$$= \sum_{k=1}^K \tau'_k f(\lambda_k)$$

Interestingly this is true for any mixture prior