

3.21 Eigenvalue expansion of a real, symmetric matrix \mathbf{A} is given by:

$$\mathbf{A} = \sum_{i=1}^D \lambda_i \mathbf{u}_i \mathbf{u}_i^T$$

where D are its dimensions, λ refers to the eigenvalue and \mathbf{u}_i the eigenvector.

Proving 3.117:

$$\begin{aligned} \text{L.H.S} &= \frac{d}{d\alpha} \ln |\mathbf{A}| = \frac{d}{d\alpha} \ln \left(\prod_{i=1}^D \lambda_i \right) \quad \text{from C.47} \\ &= \frac{d}{d\alpha} \sum_{i=1}^D \ln(\lambda_i) \\ &= \sum_{i=1}^D \frac{1}{\lambda_i} \frac{d\lambda_i}{d\alpha} \\ \\ \text{R.H.S} &= \text{Tr} \left(\mathbf{A}^{-1} \frac{d}{d\alpha} \mathbf{A} \right) \\ &= \text{Tr} \left(\left(\sum_{i=1}^D \frac{1}{\lambda_i} \mathbf{u}_i \mathbf{u}_i^T \right) \left(\frac{d}{d\alpha} \sum_{j=1}^D \lambda_j \mathbf{u}_j \mathbf{u}_j^T \right) \right) \\ &= \text{Tr} \left(\sum_{i=1}^D \sum_{j=1}^D \frac{1}{\lambda_i} \mathbf{u}_i \mathbf{u}_i^T \frac{d}{d\alpha} \lambda_j \mathbf{u}_j \mathbf{u}_j^T \right) \\ &= \text{Tr} \left(\sum_{i=1}^D \sum_{j=1}^D \frac{1}{\lambda_i} \mathbf{u}_i \mathbf{u}_i^T \left(\mathbf{u}_j \mathbf{u}_j^T \frac{d\lambda_j}{d\alpha} + \lambda_j \mathbf{u}_j \frac{d\mathbf{u}_j^T}{d\alpha} + \lambda_j \frac{d\mathbf{u}_j}{d\alpha} \mathbf{u}_j^T \right) \right) \\ &= \text{Tr} \left(\left(\sum_{i=1}^D \sum_{j=1}^D \frac{1}{\lambda_i} \mathbf{u}_i \mathbf{u}_i^T \mathbf{u}_j \mathbf{u}_j^T \frac{d\lambda_j}{d\alpha} \right) + \left(\sum_{i=1}^D \sum_{j=1}^D \frac{1}{\lambda_i} \mathbf{u}_i \mathbf{u}_i^T \lambda_j \mathbf{u}_j \frac{d\mathbf{u}_j^T}{d\alpha} \right) + \left(\sum_{i=1}^D \sum_{j=1}^D \frac{1}{\lambda_i} \mathbf{u}_i \mathbf{u}_i^T \lambda_j \frac{d\mathbf{u}_j}{d\alpha} \mathbf{u}_j^T \right) \right) \\ &= \text{Tr} \left(\sum_{i=1}^D \sum_{j=1}^D \frac{1}{\lambda_i} \mathbf{u}_i \mathbf{u}_i^T \mathbf{u}_j \mathbf{u}_j^T \frac{d\lambda_j}{d\alpha} \right) + \text{Tr} \left(\sum_{i=1}^D \sum_{j=1}^D \frac{1}{\lambda_i} \mathbf{u}_i \mathbf{u}_i^T \lambda_j \mathbf{u}_j \frac{d\mathbf{u}_j^T}{d\alpha} \right) + \text{Tr} \left(\sum_{i=1}^D \sum_{j=1}^D \frac{1}{\lambda_i} \mathbf{u}_i \mathbf{u}_i^T \lambda_j \frac{d\mathbf{u}_j}{d\alpha} \mathbf{u}_j^T \right) \end{aligned}$$

Evaluating these 3 terms separately,

$$\begin{aligned}
Tr \left(\sum_{i=1}^D \sum_{j=1}^D \frac{1}{\lambda_i} \mathbf{u}_i \mathbf{u}_i^T \mathbf{u}_j \mathbf{u}_j^T \frac{d\lambda_j}{d\alpha} \right) &= Tr \left(\sum_{i=1}^D \sum_{j=1}^D \frac{1}{\lambda_i} \mathbf{u}_i (\mathbf{u}_i^T \mathbf{u}_j) \mathbf{u}_j^T \frac{d\lambda_j}{d\alpha} \right) \\
&= Tr \left(\sum_{i=1}^D \frac{1}{\lambda_i} \mathbf{u}_i \mathbf{u}_i^T \frac{d\lambda_i}{d\alpha} \right) \\
&= \left(\sum_{i=1}^D \frac{1}{\lambda_i} Tr(\mathbf{u}_i \mathbf{u}_i^T) \frac{d\lambda_i}{d\alpha} \right) \\
&= \left(\sum_{i=1}^D \frac{1}{\lambda_i} (1) \frac{d\lambda_i}{d\alpha} \right) \\
&= \left(\sum_{i=1}^D \frac{1}{\lambda_i} \frac{d\lambda_i}{d\alpha} \right)
\end{aligned}$$

Next,

$$\begin{aligned}
Tr \left(\sum_{i=1}^D \sum_{j=1}^D \frac{1}{\lambda_i} \mathbf{u}_i \mathbf{u}_i^T \lambda_j \mathbf{u}_j \frac{d\mathbf{u}_j^T}{d\alpha} \right) &= Tr \left(\sum_{i=1}^D \sum_{j=1}^D \frac{\lambda_j}{\lambda_i} \mathbf{u}_i (\mathbf{u}_i^T \mathbf{u}_j) \frac{d\mathbf{u}_j^T}{d\alpha} \right) \\
&= Tr \left(\sum_{i=1}^D \mathbf{u}_i \frac{d\mathbf{u}_i^T}{d\alpha} \right)
\end{aligned}$$

Finally,

$$\begin{aligned}
Tr \left(\sum_{i=1}^D \sum_{j=1}^D \frac{1}{\lambda_i} \mathbf{u}_i \mathbf{u}_i^T \lambda_j \frac{d\mathbf{u}_j}{d\alpha} \mathbf{u}_j^T \right) &= Tr \left(\sum_{i=1}^D \sum_{j=1}^D \frac{\lambda_j}{\lambda_i} \mathbf{u}_i \mathbf{u}_i^T \frac{d\mathbf{u}_j}{d\alpha} \mathbf{u}_j^T \right) \\
&= Tr \left(\sum_{i=1}^D \sum_{j=1}^D \frac{\lambda_j}{\lambda_i} \frac{d\mathbf{u}_j}{d\alpha} \mathbf{u}_j^T \mathbf{u}_i \mathbf{u}_i^T \right) \quad \text{since } Tr(\mathbf{AB}) = Tr(\mathbf{BA}) \\
&= Tr \left(\sum_{i=1}^D \sum_{j=1}^D \frac{\lambda_j}{\lambda_i} \frac{d\mathbf{u}_j}{d\alpha} (\mathbf{u}_j^T \mathbf{u}_i) \mathbf{u}_i^T \right)
\end{aligned}$$

$$= Tr \left(\sum_{i=1}^D \frac{d\mathbf{u}_i}{d\alpha} \mathbf{u}_i^T \right)$$

Summing all 3, we get:

$$\begin{aligned} & \left(\sum_{i=1}^D \frac{1}{\lambda_i} \frac{d\lambda_i}{d\alpha} \right) + Tr \left(\sum_{i=1}^D \mathbf{u}_i \frac{d\mathbf{u}_i^T}{d\alpha} \right) + Tr \left(\sum_{i=1}^D \frac{d\mathbf{u}_i}{d\alpha} \mathbf{u}_i^T \right) \\ &= \left(\sum_{i=1}^D \frac{1}{\lambda_i} \frac{d\lambda_i}{d\alpha} \right) + Tr \left(\sum_{i=1}^D \mathbf{u}_i \left(\frac{d\mathbf{u}_i^T}{d\alpha} \right) + \frac{d\mathbf{u}_i}{d\alpha} \mathbf{u}_i^T \right) \\ &= \left(\sum_{i=1}^D \frac{1}{\lambda_i} \frac{d\lambda_i}{d\alpha} \right) + Tr \left(\sum_{i=1}^D \left(\frac{d\mathbf{u}_i \mathbf{u}_i^T}{d\alpha} \right) \right) \\ &= \left(\sum_{i=1}^D \frac{1}{\lambda_i} \frac{d\lambda_i}{d\alpha} \right) + \left(\sum_{i=1}^D \left(\frac{d(Tr(\mathbf{u}_i \mathbf{u}_i^T))}{d\alpha} \right) \right) \quad \text{since } d(Tr(\mathbf{X})) = Tr(d\mathbf{X}) \\ &= \left(\sum_{i=1}^D \frac{1}{\lambda_i} \frac{d\lambda_i}{d\alpha} \right) + \left(\sum_{i=1}^D \left(\frac{d(1)}{d\alpha} \right) \right) \\ &= \left(\sum_{i=1}^D \frac{1}{\lambda_i} \frac{d\lambda_i}{d\alpha} \right) + 0 \\ &= \left(\sum_{i=1}^D \frac{1}{\lambda_i} \frac{d\lambda_i}{d\alpha} \right) \\ &= \text{L.H.S} \end{aligned}$$

Now we use this result to derive 3.92, starting from 3.86.

Equation 3.86 gives us:

$$\begin{aligned} \ln p(\mathbf{t}|\alpha, \beta) &= \frac{M}{2} \ln \alpha + \frac{N}{2} \ln \beta - E(\mathbf{m}_N) - \frac{1}{2} \ln |\mathbf{A}| - \frac{N}{2} \ln(2\pi) \\ &= \frac{M}{2} \ln \alpha + \frac{N}{2} \ln \beta - \left(\frac{\beta}{2} \|\mathbf{t} - \Phi \mathbf{m}_N\|^2 + \frac{\alpha}{2} \mathbf{m}_N^T \mathbf{m}_N \right) - \frac{1}{2} \ln |\mathbf{A}| - \frac{N}{2} \ln(2\pi) \end{aligned}$$

Taking derivative w.r.t α ,

$$\begin{aligned}\frac{d \ln p(\mathbf{t}|\alpha, \beta)}{d\alpha} &= \frac{M}{2\alpha} + 0 - \frac{d}{d\alpha} \left(\frac{\beta}{2} \|\mathbf{t} - \Phi \mathbf{m}_N\|^2 + \frac{\alpha}{2} \mathbf{m}_N^T \mathbf{m}_N \right) - \frac{1}{2} \text{Tr} \left(\mathbf{A}^{-1} \frac{d}{d\alpha} \mathbf{A} \right) - 0 \\ &= \frac{M}{2\alpha} - \frac{d}{d\alpha} \left(\frac{\beta}{2} \|\mathbf{t} - \Phi \mathbf{m}_N\|^2 + \frac{\alpha}{2} \mathbf{m}_N^T \mathbf{m}_N \right) - \frac{1}{2} \text{Tr} \left(\mathbf{A}^{-1} \frac{d}{d\alpha} \mathbf{A} \right)\end{aligned}$$

Solving for the second term:

$$\begin{aligned}& \frac{d}{d\alpha} \left(\frac{\beta}{2} \|\mathbf{t} - \Phi \mathbf{m}_N\|^2 + \frac{\alpha}{2} \mathbf{m}_N^T \mathbf{m}_N \right) \\ &= \frac{d}{d\alpha} \left(\frac{\beta}{2} \|\mathbf{t} - \Phi \mathbf{m}_N\|^2 \right) + \frac{d}{d\alpha} \left(\frac{\alpha \mathbf{m}_N^T \mathbf{m}_N}{2} \right) \\ &= \frac{d}{d\mathbf{m}_N} \left(\frac{\beta}{2} \|\mathbf{t} - \Phi \mathbf{m}_N\|^2 \right) \frac{d\mathbf{m}_N}{d\alpha} + \frac{\alpha}{2} \left(\frac{d(\mathbf{m}_N^T \mathbf{m}_N)}{d\alpha} \right) + \left(\frac{\mathbf{m}_N^T \mathbf{m}_N}{2} \right) \frac{d\alpha}{d\alpha} \\ &= \frac{d}{d\mathbf{m}_N} \left(\frac{\beta}{2} \|\mathbf{t} - \Phi \mathbf{m}_N\|^2 \right) \frac{d\mathbf{m}_N}{d\alpha} + \frac{\alpha}{2} \left(\frac{d(\mathbf{m}_N^T \mathbf{m}_N)}{d\mathbf{m}_N} \right) \frac{d\mathbf{m}_N}{d\alpha} + \frac{\mathbf{m}_N^T \mathbf{m}_N}{2} \\ &= \left(\beta \Phi^T (\Phi \mathbf{m}_N - \mathbf{t}) \right)^T \frac{d\mathbf{m}_N}{d\alpha} + \frac{\alpha}{2} (2\mathbf{m}_N)^T \frac{d\mathbf{m}_N}{d\alpha} + \frac{\mathbf{m}_N^T \mathbf{m}_N}{2} \\ &= \left(\beta \Phi^T (\Phi \mathbf{m}_N - \mathbf{t}) + \alpha \mathbf{m}_N \right)^T \frac{d\mathbf{m}_N}{d\alpha} + \frac{\mathbf{m}_N^T \mathbf{m}_N}{2} \\ &= \left(\beta \Phi^T \Phi \mathbf{m}_N - \beta \Phi^T \mathbf{t} + \alpha \mathbf{m}_N \right)^T \frac{d\mathbf{m}_N}{d\alpha} + \frac{\mathbf{m}_N^T \mathbf{m}_N}{2} \\ &= \left((\beta \Phi^T \Phi + \alpha \mathbf{I}) \mathbf{m}_N - \beta \Phi^T \mathbf{t} \right)^T \frac{d\mathbf{m}_N}{d\alpha} + \frac{\mathbf{m}_N^T \mathbf{m}_N}{2} \\ &= \left(\mathbf{A} \mathbf{m}_N - \beta \Phi^T \mathbf{t} \right)^T \frac{d\mathbf{m}_N}{d\alpha} + \frac{\mathbf{m}_N^T \mathbf{m}_N}{2}\end{aligned}$$

Since $\mathbf{m}_N = \beta \mathbf{A}^{-1} \Phi^T \mathbf{t}$, $\beta \Phi^T \mathbf{t} = \mathbf{A} \mathbf{m}_N$, and the above term becomes

$$= (\mathbf{A} \mathbf{m}_N - \mathbf{A} \mathbf{m}_N)^T \frac{d\mathbf{m}_N}{d\alpha} + \frac{\mathbf{m}_N^T \mathbf{m}_N}{2}$$

$$= \frac{\mathbf{m}_N^T \mathbf{m}_N}{2}$$

Solving for the third term:

$$\begin{aligned} & \frac{1}{2} \text{Tr} \left(\mathbf{A}^{-1} \frac{d}{d\alpha} \mathbf{A} \right) \\ &= \frac{1}{2} \text{Tr} \left(\mathbf{A}^{-1} \frac{d}{d\alpha} (\alpha \mathbf{I} + \beta \Phi^T \Phi) \right) \\ &= \frac{1}{2} \text{Tr} (\mathbf{A}^{-1} \mathbf{I}) \\ &= \frac{1}{2} \text{Tr} (\mathbf{A}^{-1}) \\ &= \frac{1}{2} \left(\sum_{i=1}^M \frac{1}{\alpha + \lambda_i} \right) \end{aligned}$$

Adding the three terms, we get:

$$\frac{d \ln p(\mathbf{t}|\alpha, \beta)}{d\alpha} = \frac{M}{2\alpha} - \frac{\mathbf{m}_N^T \mathbf{m}_N}{2} - \frac{1}{2} \left(\sum_{i=1}^M \frac{1}{\alpha + \lambda_i} \right)$$

Setting the derivative to 0, we get stationary points of the evidence function w.r.t α :

$$\begin{aligned} 0 &= \frac{M}{2\alpha} - \frac{\mathbf{m}_N^T \mathbf{m}_N}{2} - \frac{1}{2} \left(\sum_{i=1}^M \frac{1}{\alpha + \lambda_i} \right) \\ \implies 0 &= \frac{M}{\alpha} - \mathbf{m}_N^T \mathbf{m}_N - \sum_{i=1}^M \frac{1}{(\alpha + \lambda_i)} \\ \implies 0 &= M - \alpha \mathbf{m}_N^T \mathbf{m}_N - \sum_{i=1}^M \frac{\alpha}{(\alpha + \lambda_i)} \\ \implies \alpha \mathbf{m}_N^T \mathbf{m}_N &= M - \sum_{i=1}^M \frac{\alpha}{(\alpha + \lambda_i)} \\ \implies \alpha \mathbf{m}_N^T \mathbf{m}_N &= \sum_{i=1}^M 1 - \sum_{i=1}^M \frac{\alpha}{(\alpha + \lambda_i)} \end{aligned}$$

$$\implies \alpha \mathbf{m}_N^T \mathbf{m}_N = \sum_{i=1}^M \frac{(\alpha + \lambda_i) - \alpha}{(\alpha + \lambda_i)}$$

$$\implies \alpha \mathbf{m}_N^T \mathbf{m}_N = \sum_{i=1}^M \frac{\lambda_i}{(\alpha + \lambda_i)}$$

$$\implies \alpha \mathbf{m}_N^T \mathbf{m}_N = \gamma$$

$$\implies \alpha = \frac{\gamma}{\mathbf{m}_N^T \mathbf{m}_N}.$$