
Active Sequential Learning with Tactile Feedback: Supplementary material

Hannes P. Saal
 University of Edinburgh
 Edinburgh, UK
 hannes.saal@ed.ac.uk

Jo-Anne Ting
 University of British Columbia
 Vancouver, BC, Canada
 jting@acm.org

Sethu Vijayakumar
 University of Edinburgh
 Edinburgh, UK
 sethu.vijayakumar@ed.ac.uk

Supplementary terms

The terms \mathbf{q} , \mathbf{Q} , and \mathbf{Z} introduced in Section 3.2.1 can be expanded as follows:

$$\begin{aligned}
 (\mathbf{q}_m)_i &= \alpha_m^2 |\boldsymbol{\Sigma}_m(\mathbf{H}_m^\theta)^{-1} + \mathbf{I}|^{\frac{1}{2}} \\
 &\quad \times \exp\left(-\frac{1}{2}(\boldsymbol{\mu} - \boldsymbol{\tau}^i)^T(\boldsymbol{\Sigma}_m + \mathbf{H}_m^\theta)^{-1}(\boldsymbol{\mu} - \boldsymbol{\tau}^i)\right) \\
 &\quad \times \exp\left(-\frac{1}{2}(\mathbf{x} - \boldsymbol{\chi}^i)^T(\mathbf{H}_m^\mathbf{x})^{-1}(\mathbf{x} - \boldsymbol{\chi}^i)\right) \\
 (\mathbf{Q}_{mn})_{ij} &= \alpha_m^2 \alpha_n^2 |((\mathbf{H}_m^\theta)^{-1} + (\mathbf{H}_n^\theta)^{-1})\boldsymbol{\Sigma} + \mathbf{I}|^{\frac{1}{2}} \\
 &\quad \times \exp\left(-\frac{1}{2}(\mathbf{z}^{mj} - \mathbf{z}^{nj})^T \mathbf{U}_{mn}^{-1} (\mathbf{z}^{mi} - \mathbf{z}^{nj})\right) \\
 &\quad \times \exp\left(-\frac{1}{2}(\boldsymbol{\tau}^i - \boldsymbol{\tau}^j)^T(\mathbf{H}_m^\theta + \mathbf{H}_n^\theta)^{-1}(\boldsymbol{\tau}^i - \boldsymbol{\tau}^j)\right) \\
 &\quad \times \exp\left(-\frac{1}{2}(\mathbf{x} - \boldsymbol{\chi}^i)^T(\mathbf{H}_m^\mathbf{x})^{-1}(\mathbf{x} - \boldsymbol{\chi}^i)\right) \\
 &\quad \times \exp\left(-\frac{1}{2}(\mathbf{x} - \boldsymbol{\chi}^j)^T(\mathbf{H}_n^\mathbf{x})^{-1}(\mathbf{x} - \boldsymbol{\chi}^j)\right) \\
 (\mathbf{Z}_m)_{\bullet i} &= (\mathbf{q}_m)_i (\boldsymbol{\Sigma}^{-1} + (\mathbf{H}_m^\theta)^{-1})^{-1} \\
 &\quad \times (\boldsymbol{\Sigma}^{-1} \boldsymbol{\mu} + (\mathbf{H}_m^\theta)^{-1} \boldsymbol{\tau}^i)
 \end{aligned}$$

where $\mathbf{z}^{mi} = \mathbf{H}_m^\theta(\boldsymbol{\tau}^i - \boldsymbol{\mu})$, $\mathbf{R}_{mn} = (\mathbf{H}_m^\theta + \mathbf{H}_n^\theta)^{-1} + \boldsymbol{\Sigma}$ and $\mathbf{U}_{mn} = (\mathbf{H}_m^\theta + \mathbf{H}_n^\theta)^{-1} \mathbf{R}^{-1} \boldsymbol{\Sigma} (\mathbf{H}_m^\theta + \mathbf{H}_n^\theta)^{-1}$. $(\mathbf{Z}_m)_{\bullet i}$ denotes the i -th column vector of matrix \mathbf{Z}_m .

The gradients are defined as:

$$\begin{aligned}
 \frac{\partial(\mathbf{q}_m)_i(\mathbf{x})}{\partial \mathbf{x}} &= (\mathbf{q}_m)_i (\mathbf{H}_m^\mathbf{x})^{-1} (\mathbf{x} - \boldsymbol{\chi}^i) \\
 \frac{\partial(\mathbf{Q}_{mn})_{ij}}{\partial \mathbf{x}} &= (\mathbf{Q}_{mn})_{ij} \left(-(\mathbf{H}_m^\mathbf{x})^{-1} (\mathbf{x} - \boldsymbol{\chi}^i) \right. \\
 &\quad \left. - (\mathbf{H}_n^\mathbf{x})^{-1} (\mathbf{x} - \boldsymbol{\chi}^j) \right) \\
 \frac{\partial(\mathbf{Z}_m)_{ij}}{\partial \mathbf{x}} &= \frac{\partial(\mathbf{q}_m)_i(\mathbf{x})}{\partial \mathbf{x}} (\boldsymbol{\Sigma}^{-1} + (\mathbf{H}_m^\theta)^{-1})^{-1} \\
 &\quad \times (\boldsymbol{\Sigma}^{-1} \boldsymbol{\mu} + (\mathbf{H}_m^\theta)^{-1} \boldsymbol{\tau})
 \end{aligned}$$