

Kernels and Derivatives of Kernels

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1 Introduction

In these notes, we describe the popular kernels such as exponentiated quadratic, Matérn 3/2, and Matérn 5/2 kernels [Rasmussen and Williams, 2006], and their derivatives. Before the details of them, we first introduce notations as follows.

Table 1: Notation.

Symbol	Meaning	Dimensionality
\mathbf{x}	a data point	\mathbb{R}^d
s	a signal scale	\mathbb{R}
ℓ	a length scale	\mathbb{R}
\mathbf{L}	a length-scale matrix	$\mathbb{R}^{d \times d}$
σ_n^2	a noise variance	\mathbb{R}
\mathbf{d}_{ij}	a difference between \mathbf{x}_i and \mathbf{x}_j (i.e., $\mathbf{x}_i - \mathbf{x}_j$)	\mathbb{R}^d
$k(\cdot, \cdot)$	a kernel	\mathbb{R}
$\widehat{k}(\cdot, \cdot)$	a kernel with observation noise	\mathbb{R}

Note that a length-scale matrix \mathbf{L} for automatic relevance determination [Rasmussen and Williams, 2006] is a diagonal matrix:

$$\mathbf{L} = \begin{bmatrix} \ell_1 & 0 & \cdots & 0 \\ 0 & \ell_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \ell_d \end{bmatrix} \in \mathbb{R}^{d \times d}. \quad (1)$$

In addition, a kernel with observation noise $\widehat{k}(\mathbf{x}_i, \mathbf{x}_j)$ is defined as

$$\widehat{k}(\mathbf{x}_i, \mathbf{x}_j) = k(\mathbf{x}_i, \mathbf{x}_j) + \sigma_n^2 \delta_{ij}, \quad (2)$$

where δ_{ij} is the Dirac delta function, which is 1 if $i = j$, and 0 otherwise.

2 Exponentiated Quadratic Kernel

Exponentiated quadratic kernel (also known as squared exponential kernel) is defined as

$$k(\mathbf{x}_i, \mathbf{x}_j) = s^2 \exp\left(-\frac{1}{2\ell^2}(\mathbf{x}_i - \mathbf{x}_j)^\top (\mathbf{x}_i - \mathbf{x}_j)\right), \quad (3)$$

$$\widehat{k}(\mathbf{x}_i, \mathbf{x}_j) = s^2 \exp\left(-\frac{1}{2\ell^2}(\mathbf{x}_i - \mathbf{x}_j)^\top (\mathbf{x}_i - \mathbf{x}_j)\right) + \sigma_n^2 \delta_{ij}, \quad (4)$$

for the case with a length scale ℓ , or

$$k(\mathbf{x}_i, \mathbf{x}_j) = s^2 \exp\left(-\frac{1}{2}(\mathbf{x}_i - \mathbf{x}_j)^\top \mathbf{L}^{-2}(\mathbf{x}_i - \mathbf{x}_j)\right), \quad (5)$$

$$\widehat{k}(\mathbf{x}_i, \mathbf{x}_j) = s^2 \exp\left(-\frac{1}{2}(\mathbf{x}_i - \mathbf{x}_j)^\top \mathbf{L}^{-2}(\mathbf{x}_i - \mathbf{x}_j)\right) + \sigma_n^2 \delta_{ij}, \quad (6)$$

for the case with a length-scale matrix \mathbf{L} .

3 Derivatives of Exponentiated Quadratic Kernel

The derivatives of (4) with respect to hyperparameters, s , ℓ , and σ_n are defined as

$$\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial s} = 2s \exp\left(-\frac{1}{2\ell^2}(\mathbf{x}_i - \mathbf{x}_j)^\top (\mathbf{x}_i - \mathbf{x}_j)\right), \quad (7)$$

$$\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial \ell} = s^2 \exp\left(-\frac{1}{2\ell^2}(\mathbf{x}_i - \mathbf{x}_j)^\top (\mathbf{x}_i - \mathbf{x}_j)\right) \cdot \frac{1}{\ell^3}(\mathbf{x}_i - \mathbf{x}_j)^\top (\mathbf{x}_i - \mathbf{x}_j), \quad (8)$$

$$\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial \sigma_n} = 2\sigma_n \delta_{ij}. \quad (9)$$

The derivatives of (6) with respect to hyperparameters, s , ℓ_1, \dots, ℓ_d , and σ_n are defined as

$$\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial s} = 2s \exp\left(-\frac{1}{2}(\mathbf{x}_i - \mathbf{x}_j)^\top \mathbf{L}^{-2}(\mathbf{x}_i - \mathbf{x}_j)\right), \quad (10)$$

$$\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial \ell_k} = s^2 \exp\left(-\frac{1}{2}(\mathbf{x}_i - \mathbf{x}_j)^\top \mathbf{L}^{-2}(\mathbf{x}_i - \mathbf{x}_j)\right) \cdot \frac{1}{\ell_k^3}(x_{ik} - x_{jk})^2, \quad (11)$$

$$\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial \sigma_n} = 2\sigma_n \delta_{ij}, \quad (12)$$

for $k = 1, \dots, d$, where x_{ik} is the k -th element of \mathbf{x}_i . The elaborate derivation of (11) is

$$\begin{aligned}
\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial \ell_k} &= \frac{\partial}{\partial \ell_k} \left(s^2 \exp \left(-\frac{1}{2} \mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij} \right) + \sigma_n^2 \delta_{ij} \right) \\
&= s^2 \exp \left(-\frac{1}{2} \mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij} \right) \cdot \frac{\partial}{\partial \ell_k} \left(-\frac{1}{2} \mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij} \right) \\
&= s^2 \exp \left(-\frac{1}{2} \mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij} \right) \cdot -\frac{1}{2} \cdot \frac{\partial}{\partial \ell_k} \left(\frac{(x_{i1} - x_{j1})^2}{\ell_1^2} + \dots + \frac{(x_{id} - x_{jd})^2}{\ell_d^2} \right) \\
&= s^2 \exp \left(-\frac{1}{2} \mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij} \right) \cdot \frac{1}{\ell_k^3} (x_{ik} - x_{jk})^2. \tag{13}
\end{aligned}$$

4 Matérn 3/2 Kernel

Matérn 3/2 kernel is defined as

$$k(\mathbf{x}_i, \mathbf{x}_j) = s^2 \left(1 + \frac{\sqrt{3}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right) \exp \left(-\frac{\sqrt{3}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right), \tag{14}$$

$$\widehat{k}(\mathbf{x}_i, \mathbf{x}_j) = s^2 \left(1 + \frac{\sqrt{3}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right) \exp \left(-\frac{\sqrt{3}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right) + \sigma_n^2 \delta_{ij}, \tag{15}$$

for the case with a length scale ℓ , or

$$k(\mathbf{x}_i, \mathbf{x}_j) = s^2 \left(1 + \sqrt{3} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \right) \exp \left(-\sqrt{3} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \right), \tag{16}$$

$$\widehat{k}(\mathbf{x}_i, \mathbf{x}_j) = s^2 \left(1 + \sqrt{3} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \right) \exp \left(-\sqrt{3} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \right) + \sigma_n^2 \delta_{ij}, \tag{17}$$

for the case with a length-scale matrix \mathbf{L} .

5 Derivatives of Matérn 3/2 Kernel

The derivatives of (15) with respect to hyperparameters, s , ℓ , and σ_n are defined as

$$\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial s} = 2s \left(1 + \frac{\sqrt{3}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right) \exp \left(-\frac{\sqrt{3}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right), \tag{18}$$

$$\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial \ell} = 3s^2 \exp \left(-\frac{\sqrt{3}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right) \cdot \frac{1}{\ell^3} \mathbf{d}_{ij}^\top \mathbf{d}_{ij}, \tag{19}$$

$$\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial \sigma_n} = 2\sigma_n \delta_{ij}. \tag{20}$$

The elaborate derivation of (19) is

$$\begin{aligned}
\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial \ell} &= \frac{\partial}{\partial \ell} \left(s^2 \left(1 + \frac{\sqrt{3}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right) \exp \left(-\frac{\sqrt{3}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right) + \sigma_n^2 \delta_{ij} \right) \\
&= s^2 \exp \left(-\frac{\sqrt{3}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right) \cdot \frac{\sqrt{3}}{\ell^2} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \\
&\quad - s^2 \exp \left(-\frac{\sqrt{3}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right) \cdot \frac{\sqrt{3}}{\ell^2} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \\
&\quad + \frac{s^2 \sqrt{3}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \exp \left(-\frac{\sqrt{3}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right) \cdot \frac{\sqrt{3}}{\ell^2} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \\
&= \frac{3s^2}{\ell^3} \mathbf{d}_{ij}^\top \mathbf{d}_{ij} \exp \left(-\frac{\sqrt{3}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right). \tag{21}
\end{aligned}$$

The derivatives of (17) with respect to hyperparameters, s , ℓ_1, \dots, ℓ_d , and σ_n are defined as

$$\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial s} = 2s \left(1 + \sqrt{3} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \right) \exp \left(-\sqrt{3} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \right), \tag{22}$$

$$\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial \ell_k} = 3s^2 \exp \left(-\sqrt{3} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \right) \cdot \frac{1}{\ell_k^3} (x_{ik} - x_{jk})^2, \tag{23}$$

$$\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial \sigma_n} = 2\sigma_n \delta_{ij}, \tag{24}$$

for $k = 1, \dots, d$, where x_{ik} is the k -th element of \mathbf{x}_i . The elaborate derivation of (23) is

$$\begin{aligned}
\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial \ell_k} &= \frac{\partial}{\partial \ell_k} \left(s^2 \left(1 + \sqrt{3} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \right) \exp \left(-\sqrt{3} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \right) + \sigma_n^2 \delta_{ij} \right) \\
&= s^2 \exp \left(-\sqrt{3} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \right) \cdot -\sqrt{3} \cdot \frac{\partial}{\partial \ell_k} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \\
&\quad + s^2 \exp \left(-\sqrt{3} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \right) \cdot \sqrt{3} \cdot \frac{\partial}{\partial \ell_k} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \\
&\quad + s^2 \sqrt{3} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \exp \left(-\sqrt{3} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \right) \cdot -\sqrt{3} \cdot \frac{\partial}{\partial \ell_k} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \\
&= -3s^2 \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \exp \left(-\sqrt{3} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \right) \cdot \frac{\partial}{\partial \ell_k} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \\
&= -3s^2 \left(\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij} \right)^{\frac{1}{2}} \exp \left(-\sqrt{3} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \right) \cdot \frac{1}{2} \left(\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij} \right)^{-\frac{1}{2}} \\
&\quad \cdot \frac{\partial}{\partial \ell_k} \left(\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij} \right) \\
&= -3s^2 \exp \left(-\sqrt{3} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \right) \cdot \frac{1}{2} \cdot \frac{\partial}{\partial \ell_k} \left(\frac{(x_{i1} - x_{j1})^2}{\ell_1^2} + \dots + \frac{(x_{id} - x_{jd})^2}{\ell_d^2} \right) \\
&= 3s^2 \exp \left(-\sqrt{3} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \right) \frac{(x_{ik} - x_{jk})^2}{\ell_k^3}. \tag{25}
\end{aligned}$$

6 Matérn 5/2 Kernel

Matérn 5/2 kernel is defined as

$$k(\mathbf{x}_i, \mathbf{x}_j) = s^2 \left(1 + \frac{\sqrt{5}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} + \frac{5}{3\ell^2} \mathbf{d}_{ij}^\top \mathbf{d}_{ij} \right) \exp \left(-\frac{\sqrt{5}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right), \quad (26)$$

$$\widehat{k}(\mathbf{x}_i, \mathbf{x}_j) = s^2 \left(1 + \frac{\sqrt{5}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} + \frac{5}{3\ell^2} \mathbf{d}_{ij}^\top \mathbf{d}_{ij} \right) \exp \left(-\frac{\sqrt{5}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right) + \sigma_n^2 \delta_{ij}, \quad (27)$$

for the case with a length scale ℓ , or

$$k(\mathbf{x}_i, \mathbf{x}_j) = s^2 \left(1 + \sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} + \frac{5}{3} \mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij} \right) \exp \left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \right), \quad (28)$$

$$\widehat{k}(\mathbf{x}_i, \mathbf{x}_j) = s^2 \left(1 + \sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} + \frac{5}{3} \mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij} \right) \exp \left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \right) + \sigma_n^2 \delta_{ij}, \quad (29)$$

for the case with a length-scale matrix \mathbf{L} .

7 Derivatives of Matérn 5/2 Kernel

The derivatives of (27) with respect to hyperparameters, s , ℓ , and σ_n are defined as

$$\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial s} = 2s \left(1 + \frac{\sqrt{5}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} + \frac{5}{3\ell^2} \mathbf{d}_{ij}^\top \mathbf{d}_{ij} \right) \exp \left(-\frac{\sqrt{5}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right), \quad (30)$$

$$\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial \ell} = \frac{5s^2}{3} \left(1 + \frac{\sqrt{5}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right) \exp \left(-\frac{\sqrt{5}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right) \cdot \frac{1}{\ell^3} \mathbf{d}_{ij}^\top \mathbf{d}_{ij}, \quad (31)$$

$$\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial \sigma_n} = 2\sigma_n \delta_{ij}. \quad (32)$$

The elaborate derivation of (31) is

$$\begin{aligned} \frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial \ell} &= \frac{\partial}{\partial \ell} \left(s^2 \left(1 + \frac{\sqrt{5}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} + \frac{5}{3\ell^2} \mathbf{d}_{ij}^\top \mathbf{d}_{ij} \right) \exp \left(-\frac{\sqrt{5}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right) + \sigma_n^2 \delta_{ij} \right) \\ &= s^2 \exp \left(-\frac{\sqrt{5}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right) \cdot \frac{\sqrt{5}}{\ell^2} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \\ &\quad - s^2 \exp \left(-\frac{\sqrt{5}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right) \cdot \frac{\sqrt{5}}{\ell^2} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \\ &\quad + \frac{s^2 \sqrt{5}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \exp \left(-\frac{\sqrt{5}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right) \cdot \frac{\sqrt{5}}{\ell^2} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \\ &\quad - \frac{10s^2}{3\ell^3} \mathbf{d}_{ij}^\top \mathbf{d}_{ij} \exp \left(-\frac{\sqrt{5}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \right) \end{aligned}$$

$$\begin{aligned}
& + \frac{5s^2}{3\ell^2} \mathbf{d}_{ij}^\top \mathbf{d}_{ij} \exp\left(-\frac{\sqrt{5}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}}\right) \cdot \frac{\sqrt{5}}{\ell^2} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \\
& = \frac{5s^2}{3\ell^3} \mathbf{d}_{ij}^\top \mathbf{d}_{ij} \exp\left(-\frac{\sqrt{5}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}}\right) \\
& \quad + \frac{5s^2\sqrt{5}}{3\ell^4} \mathbf{d}_{ij}^\top \mathbf{d}_{ij} \exp\left(-\frac{\sqrt{5}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}}\right) \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}} \\
& = \frac{5s^2}{3\ell^3} \mathbf{d}_{ij}^\top \mathbf{d}_{ij} \exp\left(-\frac{\sqrt{5}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}}\right) \left(1 + \frac{\sqrt{5}}{\ell} \sqrt{\mathbf{d}_{ij}^\top \mathbf{d}_{ij}}\right). \tag{33}
\end{aligned}$$

The derivatives of (29) with respect to hyperparameters, s , ℓ_1, \dots, ℓ_d , and σ_n are defined as

$$\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial s} = 2s \left(1 + \sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} + \frac{5}{3} \mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}\right) \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right), \tag{34}$$

$$\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial \ell_k} = \frac{5s^2}{3} \left(1 + \sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) \cdot \frac{1}{\ell_k^3} (x_{ik} - x_{jk})^2, \tag{35}$$

$$\frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial \sigma_n} = 2\sigma_n \delta_{ij}, \tag{36}$$

for $k = 1, \dots, d$, where x_{ik} is the k -th element of \mathbf{x}_i . The elaborate derivation of (35) is

$$\begin{aligned}
& \frac{\partial \widehat{k}(\mathbf{x}_i, \mathbf{x}_j)}{\partial \ell_k} \\
& = \frac{\partial}{\partial \ell_k} \left(s^2 \left(1 + \sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} + \frac{5}{3} \mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}\right) \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) + \sigma_n^2 \delta_{ij} \right) \\
& = s^2 \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) \cdot -\sqrt{5} \cdot \frac{\partial}{\partial \ell_k} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \\
& \quad + s^2 \sqrt{5} \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) \cdot \frac{\partial}{\partial \ell_k} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \\
& \quad + s^2 \sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) \cdot -\sqrt{5} \cdot \frac{\partial}{\partial \ell_k} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \\
& \quad + \frac{5s^2}{3} \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) \cdot \frac{\partial}{\partial \ell_k} \mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij} \\
& \quad + \frac{5s^2}{3} \mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij} \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) \cdot -\sqrt{5} \cdot \frac{\partial}{\partial \ell_k} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \\
& = -5s^2 \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) \cdot \frac{\partial}{\partial \ell_k} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \\
& \quad + \frac{5s^2}{3} \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) \cdot \frac{\partial}{\partial \ell_k} \mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij} \\
& \quad - \frac{5s^2\sqrt{5}}{3} \mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij} \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) \cdot \frac{\partial}{\partial \ell_k} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}
\end{aligned}$$

$$\begin{aligned}
&= -5s^2 \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) \cdot \frac{\partial}{\partial \ell_k} \left(\frac{(x_{i1} - x_{j1})^2}{\ell_1^2} + \dots + \frac{(x_{id} - x_{jd})^2}{\ell_d^2} \right)^{\frac{1}{2}} \\
&\quad + \frac{5s^2}{3} \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) \cdot \frac{\partial}{\partial \ell_k} \left(\frac{(x_{i1} - x_{j1})^2}{\ell_1^2} + \dots + \frac{(x_{id} - x_{jd})^2}{\ell_d^2} \right) \\
&\quad - \frac{5s^2 \sqrt{5}}{3} \mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij} \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) \cdot \frac{\partial}{\partial \ell_k} \left(\frac{(x_{i1} - x_{j1})^2}{\ell_1^2} + \dots + \frac{(x_{id} - x_{jd})^2}{\ell_d^2} \right)^{\frac{1}{2}} \\
&= -5s^2 \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) \cdot \frac{1}{2} (\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij})^{-\frac{1}{2}} \cdot -2 \cdot \frac{(x_{ik} - x_{jk})^2}{\ell_k^3} \\
&\quad - \frac{10s^2}{3} \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) \cdot \frac{(x_{ik} - x_{jk})^2}{\ell_k^3} \\
&\quad - \frac{5s^2 \sqrt{5}}{3} \mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij} \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) \cdot \frac{1}{2} (\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij})^{-\frac{1}{2}} \cdot -2 \cdot \frac{(x_{ik} - x_{jk})^2}{\ell_k^3} \\
&= 5s^2 \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) \cdot \frac{(x_{ik} - x_{jk})^2}{\ell_k^3} \\
&\quad - \frac{10s^2}{3} \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) \cdot \frac{(x_{ik} - x_{jk})^2}{\ell_k^3} \\
&\quad + \frac{5s^2 \sqrt{5}}{3} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}} \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) \cdot \frac{(x_{ik} - x_{jk})^2}{\ell_k^3} \\
&= \frac{5s^2}{3} \exp\left(-\sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right) \frac{(x_{ik} - x_{jk})^2}{\ell_k^3} \left(1 + \sqrt{5} \sqrt{\mathbf{d}_{ij}^\top \mathbf{L}^{-2} \mathbf{d}_{ij}}\right). \tag{37}
\end{aligned}$$

References

C. E. Rasmussen and C. K. I. Williams. *Gaussian Processes for Machine Learning*. MIT Press, 2006.