

**Proposition 1.** Exponential kernel (also called Laplace kernel)  $k(\mathbf{x}, \mathbf{y}) = e^{-\gamma\|\mathbf{x}-\mathbf{y}\|}$ , where  $\gamma > 0$  and  $\mathbf{x}, \mathbf{y} \in \mathbb{R}^d$  is a kernel.

*Proof.* The idea is to represent exponential kernel as a non-negative linear combination of Gaussian RBF kernels. To do this formally, we resort to inverse Laplace transform:

**Lemma 2.** The inverse Laplace transform of  $e^{-\gamma\sqrt{s}}$  is

$$e^{-\gamma\sqrt{s}} = \int_0^{+\infty} \frac{\gamma e^{-\frac{\gamma^2}{4t}}}{2\sqrt{\pi}t^{3/2}} e^{-ts} dt. \quad (1)$$

This can be computed directly using Bromwich integral, which can be found in most complex analysis textbooks<sup>1</sup>.

Because the integrand is non-negative when  $t \in [0, +\infty)$ ,  $k(\mathbf{x}, \mathbf{y})$  can be expressed as a non-negative linear combination of Gaussian RBF kernels:

$$k(\mathbf{x}, \mathbf{y}) = \frac{\gamma}{2\sqrt{\pi}} \int_0^{+\infty} \frac{e^{-\frac{\gamma^2}{4t}}}{t^{3/2}} e^{-t\|\mathbf{x}-\mathbf{y}\|^2} dt. \quad (2)$$

Due to the closure property of kernels, exponential kernel is a kernel.

Strictly speaking, the ‘‘linear combination’’ is an integral. To be mathematically rigorous, one can further Fourier transform the Gaussian RBF kernel, as it only depends on  $\mathbf{x} - \mathbf{y}$ :

**Lemma 3.** The Fourier transform of a Gaussian is a Gaussian:

$$e^{-\alpha\|\mathbf{x}\|^2} = \frac{1}{(2\pi)^{d/2}} \int_{\mathbb{R}^d} \frac{e^{-\frac{\|\mathbf{y}\|^2}{4\alpha}}}{(2\alpha)^{d/2}} e^{-i\mathbf{x}\cdot\mathbf{y}} d\mathbf{y}. \quad (3)$$

It follows:

$$\sum_{i,j} c_i c_j k(\mathbf{x}_i, \mathbf{x}_j) = \sum_{i,j} c_i c_j e^{-\gamma\sqrt{\|\mathbf{x}_i-\mathbf{x}_j\|^2}} \quad (4)$$

$$= \frac{\gamma}{2\sqrt{\pi}} \int_0^{+\infty} dt \sum_{i,j} c_i c_j \frac{e^{-\frac{\gamma^2}{4t}}}{t^{3/2}} e^{-t\|\mathbf{x}_i-\mathbf{x}_j\|^2} dt \quad (5)$$

$$= \frac{\gamma}{2\sqrt{\pi}} \int_0^{+\infty} dt \sum_{i,j} c_i c_j \frac{e^{-\frac{\gamma^2}{4t}}}{t^{3/2}} \frac{1}{(2\pi)^{d/2}} \int_{\mathbb{R}^d} \frac{e^{-\frac{\|\mathbf{y}\|^2}{4t}}}{(2t)^{d/2}} e^{-i(\mathbf{x}_i-\mathbf{x}_j)\cdot\mathbf{y}} d\mathbf{y} \quad (6)$$

$$= \frac{\gamma}{2^{d+1}\pi^{(d+1)/2}} \int_0^{+\infty} dt \int_{\mathbb{R}^d} d\mathbf{y} \frac{e^{-\frac{\gamma^2}{4t}}}{t^{(d+3)/2}} \left| \sum_i c_i e^{-i\mathbf{x}_i\cdot\mathbf{y}} \right|^2 \geq 0. \quad (7)$$

Therefore the exponential kernel is positive semi-definite and is indeed a kernel. □

The key in the above proof is that the Fourier transform of the exponential kernel is non-negative. To obtain the Fourier transform of the exponential kernel, we first compute its inverse Laplace transform and represent the result as an integral over  $t$ . In this way, our proof can be easily generalized to the following proposition

**Proposition 4.** If a function  $\varphi(s)$  has a non-negative inverse Laplace transform, i.e.

$$\varphi(s) = \int_0^{+\infty} \mu(t) e^{-ts} dt, \quad (8)$$

$\mu(t) \geq 0$  for  $t \in [0, +\infty)$ , then  $k(\mathbf{x}, \mathbf{y}) = \varphi(\|\mathbf{x} - \mathbf{y}\|^2)$ , where  $\mathbf{x}, \mathbf{y} \in \mathbb{R}^d$  is a kernel.

<sup>1</sup>See also, for example [this Mathematics Stack Exchange page]

For example, because the inverse Laplace transform of  $\varphi(s) = (1 + s)^{-1}$  is  $\mu(t) = e^{-t}$ , using Proposition 4,

$$k(\mathbf{x}, \mathbf{y}) = \frac{1}{1 + \|\mathbf{x} - \mathbf{y}\|^2}, \quad (9)$$

is a kernel. Similarly, the inverse Laplace transform of  $\varphi(s) = (1 + \sqrt{s})^{-1}$  is also non-negative, because

$$\frac{1}{1 + \sqrt{s}} = \int_0^{+\infty} e^{-(1+\sqrt{s})t} dt = \int_0^{+\infty} dt e^{-t} \int_0^{+\infty} \frac{te^{-\frac{t^2}{4x}}}{2\sqrt{\pi}x^{3/2}} e^{-xs} dx \quad (10)$$

$$= \int_0^{+\infty} \underbrace{\left( \int_0^{+\infty} dt \frac{e^{-t} te^{-\frac{t^2}{4x}}}{2\sqrt{\pi}x^{3/2}} \right)}_{\geq 0} e^{-xs} dx, \quad (11)$$

where we have used Lemma 2. Therefore,

$$k(\mathbf{x}, \mathbf{y}) = \frac{1}{1 + \|\mathbf{x} - \mathbf{y}\|}, \quad (12)$$

is also a kernel.

Before concluding, we note that the inverse of Proposition 4 is also true, although the proof is much more non-trivial. We have the following characterizations of positive definite functions (note that all three theorems below are stated with “if and only if”), which were proved in the 1930s:

**Definition 1** (Positive Definite Function). A function  $\Phi(\mathbf{x}) : \mathbb{R}^d \rightarrow \mathbb{C}$  is positive definite if for all  $n \in \mathbb{N}$  and  $\mathbf{x}_1, \dots, \mathbf{x}_n \in \mathbb{R}^d$ , the matrix  $M$ , where  $M_{ij} = \Phi(\mathbf{x}_i - \mathbf{x}_j)$ , is positive semi-definite.

**Theorem 5** (Bochner). A function  $\Phi(\mathbf{x}) : \mathbb{R}^d \rightarrow \mathbb{C}$  is positive definite if and only if it has non-negative Fourier transform on  $\mathbb{R}^d$ .

The above theorem characterizes all translation invariant positive definite functions. For radial functions, i.e.  $\Phi(\mathbf{x}) = \varphi(\|\mathbf{x}\|)$ , we have further characterizations:

**Definition 2** (Complete Monotone Function). A function on  $\varphi : [0, +\infty) \rightarrow \mathbb{R}$  is complete monotone on  $[0, +\infty)$  if it is continuous on  $[0, +\infty)$ , infinitely differentiable on  $(0, +\infty)$ , and satisfies

$$(-1)^n \frac{d^n}{dr^n} \varphi(r) \geq 0, \quad (13)$$

for all  $n \in \mathbb{N}$  and  $r > 0$ .

**Theorem 6** (Schoenberg). A function  $\varphi(r)$  is complete monotone on  $[0, +\infty)$  if and only if  $\Phi = \varphi(\|\mathbf{x}\|^2)$  is positive definite on  $\mathbb{R}^d$  for all  $d$ .

**Theorem 7** (Hausdorff-Bernstein-Widder). A function  $\varphi : [0, +\infty) \rightarrow \mathbb{R}$  is complete monotone on  $[0, +\infty)$  if and only if it has non-negative Laplace transform on  $[0, +\infty)$ .