

## The Radial Basis Function Kernel

The **Radial basis function kernel**, also called the **RBF kernel**, or **Gaussian kernel**, is a kernel that is in the form of a radial basis function (more specifically, a Gaussian function). The RBF kernel is defined as

$$K_{\text{RBF}}(\mathbf{x}, \mathbf{x}') = \exp \left[ -\gamma \|\mathbf{x} - \mathbf{x}'\|^2 \right]$$

where  $\gamma$  is a parameter that sets the “spread” of the kernel.

### The RBF kernel as a projection into infinite dimensions

Recall a kernel is any function of the form:

$$K(\mathbf{x}, \mathbf{x}') = \langle \psi(\mathbf{x}), \psi(\mathbf{x}') \rangle$$

where  $\psi$  is a function that projects vectors  $\mathbf{x}$  into a new vector space. The kernel function computes the inner-product between two projected vectors.

As we prove below, the  $\psi$  function for an RBF kernel projects vectors into an infinite dimensional space. For Euclidean vectors, this space is an infinite dimensional Euclidean space.

That is, we prove that

$$\psi_{\text{RBF}} : \mathbb{R}^n \rightarrow \mathbb{R}^\infty$$

**Proof:**

Without loss of generality, let  $\gamma = \frac{1}{2}$ .

$$\begin{aligned}
K_{\text{RBF}}(\mathbf{x}, \mathbf{x}') &= \exp \left[ -\frac{1}{2} \|\mathbf{x} - \mathbf{x}'\|^2 \right] \\
&= \exp \left[ -\frac{1}{2} \langle \mathbf{x} - \mathbf{x}', \mathbf{x} - \mathbf{x}' \rangle \right] \\
&= \exp \left[ -\frac{1}{2} (\langle \mathbf{x}, \mathbf{x} - \mathbf{x}' \rangle - \langle \mathbf{x}', \mathbf{x} - \mathbf{x}' \rangle) \right] \\
&= \exp \left[ -\frac{1}{2} (\langle \mathbf{x}, \mathbf{x} - \mathbf{x}' \rangle - \langle \mathbf{x}', \mathbf{x} - \mathbf{x}' \rangle) \right] \\
&= \exp \left[ -\frac{1}{2} (\langle \mathbf{x}, \mathbf{x} \rangle - \langle \mathbf{x}, \mathbf{x}' \rangle - \langle \mathbf{x}', \mathbf{x} \rangle + \langle \mathbf{x}', \mathbf{x}' \rangle) \right] \\
&= \exp \left[ -\frac{1}{2} (\|\mathbf{x}\|^2 + \|\mathbf{x}'\|^2 - 2\langle \mathbf{x}, \mathbf{x}' \rangle) \right] \\
&= \exp \left[ -\frac{1}{2} \|\mathbf{x}\|^2 - \frac{1}{2} \|\mathbf{x}'\|^2 \right] \exp \left[ -\frac{1}{2} - 2\langle \mathbf{x}, \mathbf{x}' \rangle \right] \\
&= C e^{\langle \mathbf{x}, \mathbf{x}' \rangle} \qquad C := \exp \left[ -\frac{1}{2} \|\mathbf{x}\|^2 - \frac{1}{2} \|\mathbf{x}'\|^2 \right] \text{ is a constant} \\
&= C \sum_{n=0}^{\infty} \frac{\langle \mathbf{x}, \mathbf{x}' \rangle^n}{n!} \qquad \text{Taylor expansion of } e^x \\
&= C \sum_{n=0}^{\infty} \frac{K_{\text{poly}(n)}(\mathbf{x}, \mathbf{x}')}{n!}
\end{aligned}$$

We see that the RBF kernel is formed by taking an infinite sum over polynomial kernels.

As proven previously, recall that the sum of two kernels

$$K_c(\mathbf{x}, \mathbf{x}') := K_a(\mathbf{x}, \mathbf{x}') + K_b(\mathbf{x}, \mathbf{x}')$$

implies that the  $\psi_c$  function is defined so that it forms vectors of the form

$$\psi_c(\mathbf{x}) := (\psi_a(\mathbf{x}), \psi_b(\mathbf{x}))$$

That is, the vector  $\psi_c(\mathbf{x})$  is a tuple where the first element of the tuple is the vector  $\psi_a(\mathbf{x})$  and the second element is  $\psi_b(\mathbf{x})$ . The inner-product on the vector space of  $\psi_c$  is defined as

$$\langle \psi_c(\mathbf{x}), \psi_c(\mathbf{x}') \rangle := \langle \psi_a(\mathbf{x}), \psi_a(\mathbf{x}') \rangle + \langle \psi_b(\mathbf{x}), \psi_b(\mathbf{x}') \rangle$$

For Euclidean vector spaces, this means that  $\psi_c(\mathbf{x})$  is the vector formed by appending the elements of  $\psi_b(\mathbf{x})$  onto the  $\psi_a(\mathbf{x})$  and that

$$\begin{aligned} \langle \psi_c(\mathbf{x}), \psi_c(\mathbf{x}') \rangle &:= \sum_i^{\text{dimension}(a)} \psi_{a,i}(\mathbf{x}) \psi_{a,i}(\mathbf{x}') + \sum_j^{\text{dimension}(b)} \psi_{b,j}(\mathbf{x}) \psi_{b,j}(\mathbf{x}') \\ &= \sum_i^{\text{dimension}(a)+\text{dimension}(b)} \psi_{c,i}(\mathbf{x}) \psi_{c,i}(\mathbf{x}') \end{aligned}$$

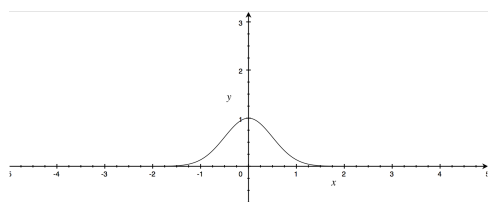
Since the RBF is an infinite sum over such appendages of vectors, we see that the projections is into a vector space with infinite dimension.

□

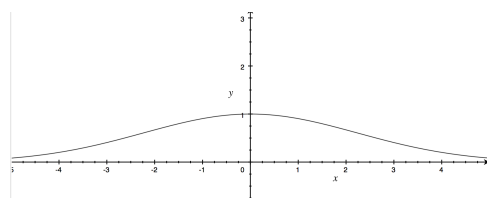
## The $\gamma$ parameter

Recall a kernel expresses a measure of similarity between vectors. The RBF kernel represents this similarity as a decaying function of the distance between the vectors (i.e. the squared-norm of their distance). That is, if the two vectors are close together then,  $\|\mathbf{x} - \mathbf{x}'\|$  will be small. Then, so long as  $\gamma > 0$ , it follows that  $-\gamma \|\mathbf{x} - \mathbf{x}'\|^2$  will be larger. Thus, closer vectors have a larger RBF kernel value than farther vectors. This function is of the form of a bell-shaped curve.

The  $\gamma$  parameter sets the width of the bell-shaped curve. The larger the value of  $\gamma$  the narrower will be the bell. Small values of  $\gamma$  yield wide bells. This is illustrated in Figure 1.



**(a)**



**(b)**

Figure 1: (a) Large  $\gamma$ . (b) Small  $\gamma$ .