

$$Q1) \quad k(x, x') = e^{-x^T x / 2\sigma^2} e^{x^T x' / \sigma^2} e^{-x'^T x' / 2\sigma^2}$$

$$\text{Let } c(x) = e^{-x^T x / 2\sigma^2}$$

$$\text{Then, } k(x, x') = c(x) \cdot e^{x^T x' / \sigma^2} \cdot c(x')$$

Consider the Taylor series expansion for $f(x) = e^x$ about $x=0$:

$$\textcircled{1} \quad e^x = \sum_{n=0}^{\infty} \frac{x^n}{n!}$$

$$\begin{aligned} \text{By } \textcircled{1}, \quad k(x, x') &= c(x) \cdot \left[\sum_{n=0}^{\infty} \frac{\left(\frac{x^T x'}{\sigma^2} \right)^n}{n!} \right] \cdot c(x') \\ &= c(x) \cdot \left[\sum_{n=0}^{\infty} \frac{K_{\text{poly}(n)}(x, x')}{\sigma^{2n} \cdot n!} \right] \cdot c(x') \end{aligned}$$

Thus, the Gaussian kernel is an infinite sum of all polynomial kernels.

As shown in class, each polynomial kernel, $K_{\text{poly}(n)}(x, x')$, can be broken down into some basis functions, say ϕ_n .

In the infinite dimensional space corresponding to the Gaussian kernel, each $\phi(x) \in \phi_n$, is multiplied by $\frac{c(x)}{\sqrt{\sigma^{2n} \cdot n!}}$.

By taking the dot product of 2 vectors in this basis space, we arrive at the kernel above.