

Assignment 4

Devansh Jain, 190100044

30th Oct 2021

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1 Clustering

1.1 CS 335 KMeans Implementation

2 Kernel design and Kernelized clustering

2.1 CS 337: Proving Kernel Validity

We are going to use the following property of kernels from Lecture slides (Lecture 11):

$$K(x, y) = \sum_{d=1}^r \alpha_d (x^T y)^d \text{ where } \alpha_d \geq 0 \text{ is a kernel (} r \text{ can be } \infty)$$

Another property of kernels which we are going to use is:

$$K(x, y) \text{ is a kernel} \implies K'(x, y) = f(x)f(y)K(x, y) \text{ is also a kernel}$$

Corresponding feature map, $\phi'(x) := f(x)\phi(x)$

An important property of exponential function which we are going to exploit is:

$$\exp(x) = \sum_{k=0}^{\infty} \frac{x^k}{k!}$$

Using the above properties, we can conclude that $\exp(\alpha x^T y)$ where $\alpha \geq 0$ is a kernel.

Now, if take $\alpha = \frac{1}{\sigma^2}$ and $f(x) = \exp\left(-\frac{x^T x}{2\sigma^2}\right)$, we get:

$$K_{\alpha}(x, y) = \exp\left(-\frac{\|x - y\|^2}{2\sigma^2}\right) = \exp\left(-\frac{x^T x}{2\sigma^2}\right) \exp\left(-\frac{y^T y}{2\sigma^2}\right) \exp\left(\frac{x^T y}{\sigma^2}\right) \text{ is a kernel.}$$

Hence, proved. □

2.2 CS 337: Simple Kernel Design

(i)

(ii)