

# STAT 542: Homework 3

Due: Mar. 8 midnight on Canvas

Please make sure that your solutions are readable and the file size is reasonable. Typing the answers is highly encouraged.

## Problem 1.

[3pts] Consider a regression problem with data  $(x_i, y_i)_{i=1}^n$ , where each  $x_i$  is one-dimensional. Consider a feature map  $\phi(x) = (1, \frac{x}{2}, \frac{x^2}{4}, \dots, \frac{x^{p-1}}{2^{p-1}})$ .

- Find an explicit expression for the kernel function  $k(x, x') = \langle \phi(x), \phi(x') \rangle$ , where  $\langle \cdot, \cdot \rangle$  is the inner product in  $\mathbb{R}^p$ .
- Suppose that we want to estimate the regression function by solving

$$\hat{\theta} := \arg \min_{\theta \in \mathbb{R}^p} \left\{ \sum_{i=1}^n |\langle \phi(x_i), \theta \rangle - y_i|^2 + \lambda \|\theta\|_2^2 \right\} \quad (1)$$

where  $\lambda > 0$ , and setting  $\hat{f}(x) = \langle \phi(x), \hat{\theta} \rangle$ . Find an expression of  $\hat{f}$  using only  $K = [k(x_i, x_j)]_{1 \leq i, j \leq n}$  (and not using  $\phi$  directly). Note that the resulting algorithm is more computationally efficient than directly solving (1) when  $n$  is much smaller than  $p$ .

- If  $\lambda \leq 0$ , is the expression of  $\hat{f}$  you found in the previous question still equivalent to (1)?

## Problem 2.

[2pts] Consider 3-NN in the setting of P6 in `s8_knn.pdf`. Show that as  $n \rightarrow \infty$ , the asymptotic error of 3-NN is upper bounded by

$$P_{3nn} \leq p(1 - p)(4p - 4p^2 + 1) \quad (2)$$

where  $p$  is the Bayes probability of error. Show that the bound can be weakened to  $P_{3nn} \leq 1.4p$ , and compare it with the case of 1-NN.

*Hint: Notes\_knn.pdf contains the main ingredients of the analysis.*

### Problem 3.

[2pts] Kernel functions can be defined over objects as diverse as graphs, sets, strings, and text documents. Consider, for instance, a fixed set and define a nonvectorial space consisting of all possible subsets of this set. If  $A_1$  and  $A_2$  are two such subsets then one simple choice of kernel would be

$$k(A_1, A_2) = 2^{|A_1 \cap A_2|} \quad (3)$$

where  $A_1 \cap A_2$  denotes the intersection of sets  $A_1$  and  $A_2$ , and  $|A|$  denotes the number of elements in  $A$ .

- Show that this is a valid kernel function, by constructing a feature map  $\phi$  such that  $k(A_1, A_2) = \langle \phi(A_1), \phi(A_2) \rangle$ .
- If  $k(A_1, A_2) = 4^{|A_1 \cap A_2|}$  instead, construct a  $\phi$  such that  $k(A_1, A_2) = \langle \phi(A_1), \phi(A_2) \rangle$ .
- [1 additional bonus point] Construction for  $k(A_1, A_2) = a^{|A_1 \cap A_2|}$ , where  $a > 1$  is arbitrary?

*Hint: Part 1) is taken from Ex 6.12 in Bishop's book: "<https://www.microsoft.com/en-us/research/uploads/prod/2006/01/Bishop-Pattern-Recognition-and-Machine-Learning-2006.pdf>". The constructions are not unique, but a convenient method for part 2), 3) is to use part 1) and then use the construction in P21 s7\_svm.pdf for a polynomial transform of a kernel.*