## Quiz 5: CS 215, Fall 2024

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3 October 2024, 08:30-08:50AM

The quiz will last for 20 minutes. A total of 18 points may be scored. There is no partial marking on questions marked multiple-options-correct.

#### 1 Valid Kernel (+2, -1)

Which of the following functions are valid Kernels?

1. K(x) = sin(x)

2. 
$$K(x) = \begin{cases} 2 & \text{if } |x| \le \frac{1}{2} \\ 0 & \text{otherwise} \end{cases}$$

3. 
$$K(x) = \frac{1}{2}e^{-|x|}$$

4. 
$$K(x) = \frac{1}{\pi(1+x^2)}$$

5. 
$$K(x) = \frac{3}{4} max\{1 - x^2, 0\}$$

Correct Answers: 3, 4, 5

## 2 Miscellaneous (+3, -2)

Which of the following statements are true?

- 1. The optimal bandwidth (h) for Kernel Density Estimation is of the order of  $O(n^{-\frac{4}{5}})$  (where n is the number of data points).
- 2. The optimal bandwidth (h, in this case bin-width) for density estimation using histograms is of the order of  $O(n^{-\frac{1}{3}})$  (where n is the number of data points).
- 3. For the optimal bandwidth, bias is minimized.
- 4. The mean squared error is given by Variance<sup>2</sup> + Bias.

**Correct Answer: 2** 

# 3 Bias and variance (+3, -1)

State which of the following are true about the histogram and/or KDE density estimator for a distribution f(x). Assume the histogram width is h

- 1. The bias of the histogram estimate when f(x) = U(0.1, 1) and histogram creates bins of width 0.2 between 0 and 1 is zero everywhere.
- 2. The bias of the KDE estimate reduces with increasing size of data sample.
- 3. The risk of the KDE reduces faster than histogram's with increasing data size.
- 4. If f(x) is a Gaussian distribution, then the KDE density with Gaussian Kernel will incur a bias of zero.

Correct Answer: 3

#### 4 Multivariate Kernel Density Estimation (+3, -1)

Suppose that the datapoints  $x_i \in \mathbb{R}^d$  (d-dimensional datapoints). The kernel density estimator  $\hat{f}(x) = \frac{1}{n} \sum_{i=1}^n \frac{1}{h^p} K(\frac{x-x_i}{h})$  (h = bandwidth). Which of the following statements are true?

- 1. The optimal bandwidth (h) for Kernel Density Estimation is of the order of  $O(n^{-\frac{1}{d+2}})$ .
- 2. The mean squared error is of the order of  $O(n^{-\frac{4}{d+4}})$ .
- 3. The bias is of the order of  $O(h^3)$ .
- 4. The Variance is of the order of  $O((nh^d)^{-1})$ .

Correct Answer: 2,4

## 5 Example Kernel (+2, -1)

Consider the following kernel:  $K(x) = N(1 - 4x^2)$  (defined for the range  $|x| \le \frac{1}{2}$  otherwise 0) (N is the normalization constant). We have four datapoints  $x_1 = 0.1$ ,  $x_2 = 0.4$ ,  $x_3 = 0.3$ ,  $x_4 = -0.3$ . The bandwidth is h = 0.3. What is the value of the kernel density estimator at x = 0.2 (Correct to two decimal places)?

Correct Answer: 1.39

## 6 Image Segmentation (+3, -1)

The mean-shift algorithm is often used in image segmentation and is based on Kernel Density Estimation (KDE). For this question we will use a Gaussian  $(N(0; \sigma))$  Kernel to compute the density function. The goal of the algorithm is to find the closest mode (peak) of the probability density function (pdf) estimated from the data samples  $\{x_i\}$  for each sample  $x_i$ . Essentially, a group of points will converge to the same mode after several iterations, and these points will form a cluster. Unlike K-means, where you need to define the number of clusters, mean-shift automatically determines the number of clusters. Which of the following statements are true about the mean-shift algorithm?

- 1. We move in the direction of gradient descent of the density function i.e. we move in the direction opposite to the gradient of the density function.
- 2. We should do the following iteratively:  $x = x + \epsilon \nabla f(x)$  where f(x) = Kernel Density Estimator Function
- 3.  $\hat{p}(x) = \frac{1}{N} \sum_{i=1}^{N} K(x x_i; \sigma)$  where  $K(x x_i; \sigma) = \frac{1}{(2\pi)^d \sigma^d} exp(-\frac{||x x_i||^2}{2\sigma^2})$  where  $d = \text{dimension of the data and } \hat{p}(x) = \text{kernel density estimator.}$
- 4.  $\nabla \hat{p}(x) = \frac{1}{\sigma^2} \hat{p}(x) m(x; \sigma)$  where  $m(x; \sigma) = \frac{\sum_{i=1}^{N} x_i exp(-\frac{||x-x_i||^2}{2\sigma^2})}{\sum_{i=1}^{N} exp(-\frac{||x-x_i||^2}{2\sigma^2})} x$

Correct Answer: 2, 4

## 7 CDF (+2, -1)

Consider an empirical estimate  $\hat{F}(x)$  of the CDF of 10 random variables whose actual CDF is F(x). The highest expected risk of the estimate is

**Correct Answer: 0.025** 

Explanation: The variance of the estimate is F(x)(1 - F(x))/10. This has a maximum value of 0.25/10