

## EECS545 Lecture 8 Quiz Solutions

1. Which of the following are true statements about kernels? **Choose all options that apply:**
- (a) A machine learning algorithm can be kernelized if it does not need explicit access to the feature vectors and instead only requires access to inner products of the feature vectors.
  - (b) A Gram/kernel matrix must be positive semidefinite.
  - (c) The product of two kernel functions is still a kernel function.
  - (d) The sum of two kernel functions is always a kernel function.

**Solution:** (a), (b), (c), (d). All of them are true.

2. What is the purpose of the kernel trick?
- (a) To transform the problem from regression to classification.
  - (b) To transform the data into a richer feature space without explicitly computing the feature vector.
  - (c) To transform the problem from supervised to unsupervised learning.
  - (d) To transform a linear regression model to SVM.

**Solution:** (b) Please check the slide 10 of Lecture 8 for an example.

3. (True/False) For any two documents  $\mathbf{x}$  and  $\mathbf{z}$ , define  $k(\mathbf{x}, \mathbf{z})$  to equal the number of unique words that occur in both  $\mathbf{x}$  and  $\mathbf{z}$  (i.e., the size of the intersection of the sets of words in the two documents  $\mathbf{x}$  and  $\mathbf{z}$ ). This function cannot be considered as a kernel.

**Solution:** False. We can make  $k$  as kernel by setting  $\phi(\mathbf{x})$  as a binary vector whose  $i$ -th entry is 1 when the document  $\mathbf{x}$  contains the  $i$ -th word and 0 if it doesn't.

4. (True/False)  $k(\mathbf{x}, \mathbf{y}) = k_1(\mathbf{x}, \mathbf{y}) - k_2(\mathbf{x}, \mathbf{y})$  is a kernel if  $k_1$  and  $k_2$  are valid kernels, and those kernels are defined as  $k_1(\mathbf{x}, \mathbf{y}) = (\mathbf{x} \diamond \mathbf{y} + 1)^2$  and  $k_2(\mathbf{x}, \mathbf{y}) = \mathbf{x} \diamond \mathbf{y}$  (Please assume  $\diamond$  is a predefined operation that can lead  $k_1$  and  $k_2$  valid kernel. For example,  $\diamond(\mathbf{x}, \mathbf{y}) = 2\mathbf{x} + \frac{\mathbf{y}}{2}$  could work as a candidate).

**Solution:** True.  $k(\mathbf{x}, \mathbf{y}) = (\mathbf{x} \diamond \mathbf{y})^2 + \mathbf{x} \diamond \mathbf{y} + 1$  with each term in the sum being a kernel.

5. (True/False) A Gaussian kernel will always have better classification performance at a testing time compared to a linear kernel.

**Solution:** False. Bad hyperparameter choice or small (finite) training data size setting could lead the kernel to overfit.