## CMSC5724: Exercise List 6

**Problem 1.** Prove the theorem on Slide 6 of the lecture notes on the kernel method without the interleaving assumption.

**Problem 2.** Consider the kernel function  $K(p,q) = (\mathbf{p} \cdot \mathbf{q} + 1)^3$ , where  $\mathbf{p} = (p[1], p[2])$  and  $\mathbf{q} = (q[1], q[2])$  are 2-dimensional vectors. Recall that there is a mapping function  $\phi$  from  $\mathbb{R}^2$  to  $\mathbb{R}^d$  for some integer d such that K(p,q) equals the dot product of  $\phi(p)$  and  $\phi(q)$ . Give the details of  $\phi$ .

**Problem 3.** Consider a set P of 2D points each labeled either -1 or 1. It is known that the points of the two labels can be linearly separated after applying the Kernel function  $K(p,q) = (\mathbf{p} \cdot \mathbf{q} + 1)^2$ . Prove that they can also be linearly separated by applying the kernel function  $K'(p,q) = (2\mathbf{p} \cdot \mathbf{q} + 3)^2$ .

**Problem 4.** Consider a set P of 2D points that has three label-1 points  $p_1(-2, -2)$ ,  $p_2(1, 1)$ ,  $p_3(3, 3)$ , and two label-(-1) points  $q_1(-2, 2)$ ,  $q_2(2, -2)$ . Answer the following questions:

- Use Perceptron to find a separation plane  $\pi$  using the Kernel function  $K(x, y) = (x \cdot y + 1)^2$ .
- According to  $\pi$ , what is the label of point (2,2)?

**Problem 5.** Same settings as in Problem 3. Calculate the distance from  $\phi(p_1)$  to the separation plane you find in the feature space.

**Problem 6.** Let P be a set of points in  $\mathbb{R}^d$ . Prove: the Gaussian kernel produces a kernel space where every point  $p \in P$  is mapped to a point  $\phi(p)$  satisfying  $|\phi(p)| = 1$ , namely,  $\phi(p)$  is on the surface of an infinite-dimensional sphere.

**Problem 7.** For any a *d*-dimensional sphere centered at the origin of  $\mathbb{R}^d$ , we know that any set of d+1 points on the sphere's surface can be shattered by the set of linear classifiers. Use this fact to prove that any finite set P of points in  $\mathbb{R}^d$  can be linearly separated in the kernel space produced by the Gaussian kernel. (Hint: use the conclusion of Problem 6 and use the fact that the Gaussian kernel produces a kernel space of infinite dimensionality.)