

## CS446: Machine Learning, Fall 2017, Homework 3

Name: Lanxiao Bai (lbai5)

*Worked individually*

### Problem 1

**Solution:**

$$w_t = w_0 + \sum_{i=1}^t a_i y_{i-1} x_{i-1} = \sum_{i=1}^t a_i y_{i-1} x_{i-1}$$

in which

$$a_i = \begin{cases} 1 & \text{sgn}(w_{i-1} \cdot x_i) \neq y_i \\ 0 & \text{otherwise} \end{cases}$$

### Problem 2

**Solution:** We notice that in kernelized ridge regression, we have

$$w = \sum_{i=1}^N \alpha_i \mathbf{x}_i$$

in which  $\alpha_i > 0$  for support vectors.

So the form of two algorithm is the same, while in kernelized perceptron the coefficient  $a_i$  is at most 1. Which means that perceptron does not maximize the margin.

### Problem 3

**Solution:**

$$\hat{y} = \text{sgn}(w \cdot x) = \text{sgn}\left(\sum_{i=1}^t \alpha_i x_{i-1} x\right) = \text{sgn}\left(\sum_{i=1}^t \alpha_i x_{i-1}^T x\right)$$

in which

$$\alpha_i = \begin{cases} y_i & \text{sgn}(w_{i-1} \cdot x_i) \neq y_i \\ 0 & \text{otherwise} \end{cases}$$

### Problem 4

**Solution:**

$$\hat{y} = \text{sgn}(w \cdot \phi(x)) = \text{sgn}\left(\sum_{i=1}^t \alpha_i \phi(x_{i-1}) \phi(x)\right) = \text{sgn}\left(\sum_{i=1}^t \alpha_i \kappa(x, x_{i-1})\right)$$

## Problem 5

**Solution:** The kernelized algorithm we got is

1. Initialize  $\alpha_0 = 0$ , where  $\alpha_0 \in \mathbb{R}^d$
2. For  $t$  in  $1 \cdots N$ :
  - (a) Predict  $\hat{y} = \text{sgn}(\sum_{i=1}^t \alpha_i \kappa(x, x_{i-1}))$
  - (b) If  $\hat{y}_t \neq y_t$ , update the weight:
    - i.  $\alpha_t = \alpha_{t-1} + y_t \phi(x_t)$

## References