

# Assignment 2:

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## 1 Theoretical question on K-means Clustering

Initially we have two datasets:  $D_1 = \{-2, \dots, -2\}$   $D_2 = \{0, \dots, 0, a\}$ , where  $D_1$  has  $m-2$  values and  $D_2$  has  $m-0$  and one  $a$ .

Let us consider other dataset, which may have lower sum-of-squared error. Given two sets has sum-of-squared error  $J$ .  $D_1 = \{-2, \dots, -2, 0, \dots, 0\}$   $D_2 = \{a\}$ , where  $D_1$  has one  $a$  values and  $D_2$  has  $m-2$  and  $m-0$ . This dataset will have low or 0 sum-of-squared error for dataset  $D_2$  and low sum-of-squared error for  $D_1$ . This datasets has sum-of-squared error  $J'$ .

Now  $J$  and  $J'$  will be computed:

First calculate  $J$ .

Calculate  $\mu_1$  and  $\mu_2$ :

$$\mu_1 = \sum_{x \in D_1} (x/m) = -2m/m = -2$$

$$\mu_2 = \sum_{x \in D_2} (x/m) = a/(m+1) + 0/(m+1) = a/(m+1)$$

$$\text{Now calculate } J: J = \sum \sum_{i: D_i \neq \emptyset, x \in D_i} (x - \mu_i)^2 = 0 + (0 - a)^2 * m/(m+1)^2 + (a - a/(m+1))^2 = a^2 m/(m+1)$$

Secondly calculate  $J'$ .

Calculate  $\mu_1$  and  $\mu_2$ :

$$\mu_1 = \sum_{x \in D_1} (x/m) = (-2m + 0m)/2m = -1$$

$$\mu_2 = \sum_{x \in D_2} (x/m) = a/1 = a$$

$$\text{Now calculate } J': J' = \sum \sum_{i: D_i \neq \emptyset, x \in D_i} (x - \mu_i)^2 = (-2 + 1)^2 * m + (0 - 1)^2 * m = 2m$$

Thus, our initial dataset will have minimum sum-of-squared error  $J < J'$ . So  $a^2 m/(m+1) < 2m$ , thus  $a^2 < 2(m+1)$

## 2 Theoretical question on SVM.

(I):

**a: No** Given margin doesn't have support vector from negative dataset. According to the definition, margin is computed based on perpendicular distances from both sides of the decision boundary. Thus, given margin cannot be produced by SVM described by formula (I).

**b: No** Formula (I) has condition:  $y_t \theta^T x_t > 1$  which means that  $\theta_0 = 0$ . Therefore, margin goes through the origin. As we see on the plot b, margin b doesn't go through the origin. So plot b cannot be generated by formula (I).

**c: No** There are two points from "+" dataset which doesn't satisfy condition  $y_t \theta^T x_t \geq 1$ . Those points has distance less than 1. Thus, margin c cannot be constructed.

(II):

**a: No** Given margin doesn't have support vector from negative dataset. According to the definition, margin is computed based on perpendicular distances from both sides of the decision boundary. Thus, given margin cannot be produced by SVM described by formula (II).

**b: Yes**

**c: No** There are two points from "+" dataset which doesn't satisfy condition  $y_t (\theta^T x_t + \theta_0) \geq 1$ . Those points has distance less than 1. Thus, margin c cannot be constructed.

(III):

**a: No** Given margin doesn't have support vector from negative dataset. According to the definition, margin is computed based on perpendicular distances from both sides of the decision boundary. Thus, given margin cannot be produced by SVM described by formula (I).

**b: No** Formula (III) has condition  $y_t \theta^T x_t \geq 1 - \epsilon_t$  which means that constant  $\theta_0 = 0$ , thus margins that are created by formula (III) goes through the origin. Margin on plot b doesn't satisfy this condition. Thus, b is not created by formula (III)

**c: Yes**