

Homework 3

CS - 4033

Roman Munoz

1)

$$- \alpha = [\alpha_1, \dots, \alpha_m]^T$$

$$- Y = [y_1, \dots, y_n]^T$$

- \tilde{K} is an $n \times m$ matrix, with $\tilde{K}_{ij} = K(x_i, x_j)$

From AKRR:

$$\tilde{\beta} = \sum_{i=1}^m \alpha_i \phi(x_i)$$

Now to solve:

$$\min_{\beta} \sum_{i=1}^n (\phi(x_i)^T \tilde{\beta} - y_i)^2 + \lambda \tilde{\beta}^T \tilde{\beta}$$

$$J(\beta) = \sum_{i=1}^n \left[\phi(x_i)^T \sum_{j=1}^m \alpha_j \phi(x_j) - y_i \right]^2 + \lambda \sum_{i=1}^n \alpha_i \phi(x_i)^T \cdot \sum_{j=1}^m \alpha_j \phi(x_j)$$

$$= \sum_{i=1}^n \left[\sum_{j=1}^m \alpha_j \phi(x_i)^T \cdot \phi(x_j) - y_i \right]^2 + \lambda \sum_{i=1}^n \sum_{j=1}^m \alpha_i \alpha_j \cdot \phi(x_i)^T \cdot \phi(x_j)$$

From Kernel trick: $\phi(x_i)^T \phi(x_j) = K(x_i, x_j)$, thus,

$$J(\beta) = \sum_{i=1}^n \left[\sum_{j=1}^m \alpha_j \cdot K(x_i, x_j) \right]^2 + \lambda \sum_{i=1}^n \sum_{j=1}^m \alpha_i \alpha_j \cdot K(x_i, x_j)$$

And since $\tilde{K}_{ij} = K(x_i, x_j)$:

$$J(\beta) = \sum_{i=1}^n (\tilde{K}_{i:} \alpha - y_i)^2 + \lambda \sum_{i,j} \alpha_i \alpha_j \cdot \tilde{K}_{ij}$$

Matrix form:

$$J(\beta) = (\tilde{K}_1 \cdot \alpha - Y)^T (\tilde{K}_1 \cdot \alpha - Y) + \lambda \alpha^T \tilde{K}_2 \alpha$$

$$= \alpha^T \tilde{K}_1^T \tilde{K}_1 \alpha - 2 \alpha^T \tilde{K}_1^T Y + Y^T Y + \lambda \alpha^T \tilde{K}_2 \alpha$$

Now if we take derivative:

$$\frac{\partial J(\beta)}{\partial \beta} = 2 \tilde{K}_1^T \tilde{K}_1 \alpha - 2 \tilde{K}_1^T Y + 2 \lambda \tilde{K}_2 \alpha$$

$$\alpha = (\tilde{K}_1^T \tilde{K}_1 + \lambda \tilde{K}_2)^{-1} \cdot \tilde{K}_1^T Y$$

2. From previous result, we can see that it takes t.c. $O(nmp)$ for \tilde{K}_1^T and it takes $O(m^2n)$ for $\tilde{K}_1^T \tilde{K}_1$ since \tilde{K}_1^T is an $m \times n$ matrix and \tilde{K}_1 is an $n \times m$ matrix. Additionally it takes $O(m^3)$ to take the inverse.

\therefore the total t.c. = $O(nmp) + O(m^2n) + O(m^3)$, and since $m \leq d$
then, t.c. = $O(nmp) + O(m^2n)$