

### Question 21

The least number of queries is  $N + 1$ .

$$RMSE(h) = 0 \rightarrow \sum_{n=1}^N (y_n - h(x_n))^2 = 0$$

Let  $k \in \mathbb{R}^N$  and  $query(k) = \sum_{n=1}^N (y_n - k_n)^2 = d$ .

Start with  $k_0 = \{0\}^N$ , then  $query(k_0) = \sum_{n=1}^N y_n^2 = d_0$

Construct a query  $query(k_i)$  then subtract from  $query(k_0)$ , we get :

$$\begin{aligned} query(k_0) - query(k_i) &= \sum_{n=1}^N y_n^2 - \sum_{n=1}^N (y_n - k_{in})^2 \\ &= \sum_{n=1}^N (2k_{in}y_n - k_{in}^2) \\ &= d_i \end{aligned}$$

which is a linear equation about  $y$ . If we want to solve  $y$ , we need at least  $N$  such linear equations about  $y$ .

e.g. Consider a reasonable  $K$  given below:

$$K = \begin{bmatrix} k_1 \\ \vdots \\ k_i \\ \vdots \\ k_N \end{bmatrix} = \begin{bmatrix} 1 & \cdots & 0 & \cdots & 0 \\ \vdots & \ddots & \vdots & & \vdots \\ 0 & \cdots & 1 & \cdots & 0 \\ \vdots & & \vdots & \ddots & \vdots \\ 0 & \cdots & 0 & \cdots & 1 \end{bmatrix}$$

$$query(K) = \begin{bmatrix} query(k_1) \\ \vdots \\ query(k_i) \\ \vdots \\ query(k_N) \end{bmatrix}$$

$$query(k_0) - query(K) = \begin{bmatrix} query(k_0) - query(k_1) \\ \vdots \\ query(k_0) - query(k_i) \\ \vdots \\ query(k_0) - query(k_N) \end{bmatrix} = \begin{bmatrix} 2y_1 - 1 \\ \vdots \\ 2y_i - 1 \\ \vdots \\ 2y_N - 1 \end{bmatrix} = \begin{bmatrix} d_1 \\ \vdots \\ d_i \\ \vdots \\ d_N \end{bmatrix}$$

Then we can solve it easily. So the total number of queries we need is at least  $N + 1$ .

### Question 23

$$\min_{w_1, w_2, \dots, w_K} RMSE(H) \rightarrow \min_{w_1, w_2, \dots, w_K} \sum_{n=1}^N (y_n - H(x_n))^2$$

Let

$$\begin{aligned} f(w) &= \sum_{n=1}^N (y_n - H(x_n))^2 \\ &= \sum_{n=1}^N y_n^2 - 2 \sum_{n=1}^N y_n H(x_n) + \sum_{n=1}^N H^2(x_n) \end{aligned}$$

Where  $H(x_n) = \sum_{k=1}^K w_k h_k(x_n)$  If  $f$  is minimized, the partial derivative of all  $w_i$  should be 0.

$$\begin{aligned} \frac{\partial f}{\partial w_i} &= -2 \sum_{n=1}^N y_n h_i(x_n) + 2 \sum_{n=1}^N H(x_n) h_i(x_n) \\ &= -2 h_i^T y + 2 \sum_{n=1}^N (h_i(x_n) \cdot \sum_{k=1}^K w_k h_k(x_n)) \\ &= -2 h_i^T y + 2 \sum_{k=1}^K (\sum_{n=1}^N h_i(x_n) h_k(x_n)) \cdot w_k \\ &= -2 h_i^T y + 2 \sum_{k=1}^K h_i^T h_k w_k \\ &= 0 \end{aligned}$$

That means we should solve such linear equations:

$$\begin{aligned} h_1^T h_1 w_1 + h_1^T h_2 w_2 + \cdots + h_1^T h_K w_K &= h_1^T y \\ h_2^T h_1 w_1 + h_2^T h_2 w_2 + \cdots + h_2^T h_K w_K &= h_2^T y \\ &\vdots \\ h_K^T h_1 w_1 + h_K^T h_2 w_2 + \cdots + h_K^T h_K w_K &= h_K^T y \end{aligned}$$

Followed by question 22, as we have already known  $h_i^T h_j$ , the least number of queries to get  $h_1^T y \sim h_K^T y$  is  $K + 1$ , so the least number of queries would be  $K + 1$  to solve  $\min_{w_1, w_2, \dots, w_K} RMSE(H)$ .