

# Comp 6321 - Machine Learning - Assignment 2

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## Question 1:

### 1.a Doodles

$$\arg \min_w \frac{1}{2}(\Phi \mathbf{w} - \mathbf{y})^T(\Phi \mathbf{w} - \mathbf{y}) + \frac{\lambda}{2} \sum_{k=0}^{K-1} |w_k| \quad (1)$$

Which is equivalent to finding:

$$\arg \min_w (\Phi \mathbf{w} - \mathbf{y})^T(\Phi \mathbf{w} - \mathbf{y}) + \lambda \sum_{k=0}^{K-1} |w_k| \quad (2)$$

And expands to:

$$\arg \min_w \mathbf{w}^T \Phi^T \Phi \mathbf{w} - 2\mathbf{y}^T \Phi \mathbf{w} + \mathbf{y}^T \mathbf{y} + \lambda \sum_{k=0}^{K-1} |w_k| \quad (3)$$

And for which we can remove the constant term  $\mathbf{y}^T \mathbf{y}$ , yielding:

$$\arg \min_w \mathbf{w}^T \Phi^T \Phi \mathbf{w} - 2\mathbf{y}^T \Phi \mathbf{w} + \lambda \sum_{k=0}^{K-1} |w_k| \quad (4)$$

Matlab's `quadprog(H, f, A, b)` function, gives the optimal  $\mathbf{x}$  corresponding to the expression  $\arg \min_{\mathbf{x}} \mathbf{x}^T \mathbf{H} \mathbf{x} + \mathbf{f}^T \mathbf{x}$ , subject to constraints  $\mathbf{A} \mathbf{x} \leq \mathbf{b}$ . We can thus take  $\mathbf{H} := \Phi^T \Phi$ , then  $\mathbf{f} := 2\mathbf{y}^T \Phi$ ,  $\mathbf{A} := \lambda \mathbf{P}$ , where for a system with  $n$  variables,  $\mathbf{P}$  is the matrix with  $2^n$  permutations of  $[b_1, b_2, \dots, b_n]$ ,  $b \in \{-1, 1\}$  and lastly  $\mathbf{b} := c \cdot \mathbf{1}$  places an upper bound to the expression  $\lambda \sum_{k=0}^{K-1} |w_k|$ , such that  $\sum_{k=0}^{K-1} |w_k| \leq \frac{c}{\lambda}$  TODO develop relationship between lambda, c and whatnot, too tired to think now...

This here, on the other hand, is something else...

$$\begin{aligned} \min_w J_D(\mathbf{w}) &= \min_w (\Phi \mathbf{w} - \mathbf{y})^T(\Phi \mathbf{w} - \mathbf{y}) \\ s.t. \sum_{i=1}^n |w_i| &\leq \eta \end{aligned} \quad (5)$$