

# Modern Statistical Methods — Example Sheet 1

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November 2, 2020

**Question 7.** Suppose we have a matrix of predictors  $X \in \mathbb{R}^{n \times p}$  where  $p \gg n$ . Explain how to obtain the fitted values of the following ridge regression using the kernel trick:

$$\begin{aligned} & \text{Minimise over } \beta \in \mathbb{R}^p, \vartheta \in \mathbb{R}^{p(p-1)/2}, \gamma \in \mathbb{R}^p, \\ & \sum_{i=1}^n \left( Y_i - \sum_{k=1}^p X_{ik} \beta_k - \sum_{k=1}^p \sum_{j=1}^{k-1} X_{ik} X_{ij} \vartheta_{jk} - \sum_{k=1}^p X_{ik}^2 \gamma_k \right)^2 + \lambda_1 \|\beta\|^2 + \lambda_2 \|\vartheta\|^2 + \lambda_3 \|\gamma\|^2. \end{aligned}$$

Note that we have indexed  $\vartheta$  with two numbers for convenience.

*Proof.* Our linear model has predictors

$$\{X_{ik} \mid 1 \leq k \leq p\} \cup \{X_{ij} X_{ik} \mid 1 \leq j < k \leq p\} \cup \{X_{ik}^2 \mid 1 \leq k \leq p\}$$

for  $i = 1, \dots, n$ . We want to assume that  $\lambda_1 = \lambda_2 = \lambda_3$ : to this end, define  $\xi = \frac{\lambda_1}{\lambda_2}$  and  $\eta = \frac{\lambda_3}{\lambda_1}$ , and replace  $\vartheta$  by  $\xi \vartheta$  and  $\gamma$  by  $\eta \gamma$ . This means we will have to replace our predictors by

$$Z_i := (X_{ik} \mid 1 \leq k \leq p)^\top \cup (\xi X_{ij} X_{ik} \mid 1 \leq j < k \leq p)^\top \cup (\eta X_{ik}^2 \mid 1 \leq k \leq p)^\top$$

We use the kernel trick: note that

$$\begin{aligned} \langle Z_i, Z_j \rangle &= \sum_k X_{ik} X_{jk} + \xi^2 \sum_{k < \ell} X_{ik} X_{i\ell} X_{jk} X_{j\ell} + \eta^2 \sum_k X_{ik}^2 X_{jk}^2 \\ &= \left( \frac{1}{\sqrt{2\xi}} + \frac{\xi}{\sqrt{2}} \sum_k X_{ik} X_{jk} \right)^2 + \left( \eta^2 - \frac{\xi^2}{2} \right) \sum_k X_{ik}^2 X_{jk}^2 - \frac{1}{2\xi^2}. \end{aligned}$$

We have therefore rewritten  $\langle Z_i, Z_j \rangle$  into something that can be computed in  $O(p)$ , which is exactly the aim of the kernel trick, since this allows computation of the kernel matrix  $K$  in  $O(n^2 p)$ .  $\square$