

# MATH60005/70005: Optimisation (Autumn 24-25)

## Chapter 3: exercises

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### Polynomial fit and denoising

We will replicate the linear regression and regularized linear least squares examples from this week, with a different model. First, we will generate a noisy dataset of 200 samples coming from

$$v_i = u_i^2 + \mathcal{N}(0, 0.04), \quad i = 1, \dots, 200,$$

where the  $u_i$ 's are uniformly sampled in  $[-1, 1]$ , and  $\mathcal{N}(0, 0.04)$  means adding Gaussian noise of mean 0 and variance 0.04 for **each** sample.

1. Generate the pairs  $(u_i, v_i)$  using suitable random generators. Make a plot illustrating  $(u_i, v_i)$ . What is the model you can identify to express  $v$  as a function of  $u$ ?
2. Write a linear regression problem for finding the optimal parameters in a model

$$v(u) = au^2 + bu + c$$

3. Compute the least squares solution and compute the total least squares error in the  $\ell_2$  norm.
4. Now, instead of solving a regression problem, use the  $v_i$  values to recover a denoised signal using regularized least squares using the same total variation regularization described in the lecture notes (denoising part).

