

## SECTION B: INVERSE PROBLEMS

## Question 3 (15 marks)

Consider the standard Tikhonov form of the regularised least squares problem

$$\min_{\mathbf{x}} \|\mathbf{Ax} - \mathbf{y}\|_2^2 + \lambda \|\mathbf{x}\|_2^2$$

where  $\mathbf{x} \in \mathbb{R}^n$ ,  $\mathbf{y} \in \mathbb{R}^m$ ,  $\lambda \in \mathbb{R}$  and  $\mathbf{A}$  is an  $m \times n$  matrix.

- i) By differentiating the objective function with respect to  $\mathbf{x}$ , derive the following corresponding normal equations that characterise solutions to the optimisation problem above:

$$(\mathbf{A}^T \mathbf{A} + \lambda \mathbf{I}) \mathbf{x} = \mathbf{A}^T \mathbf{y}.$$

[5 marks]

- ii) Show how to derive an *iterative solution method* to the above normal equations. Note: you do NOT need to give conditions under which your scheme converges.

[5 marks]

- iii) Suppose that you decided to set the explicit Tikhonov regularisation parameter  $\lambda \in \mathbb{R}$  to zero in your problem. Explain how you could *still* obtain a regularised solution just using an iterative scheme.

[5 marks]

**Question 4 (10 marks)**

- i) Consider the general reduced form of the SVD of a matrix  $\mathbf{A}$ :

$$\mathbf{A} = \mathbf{U}_r \mathbf{\Sigma}_r \mathbf{V}_r^T,$$

where  $r$  is the rank of  $\mathbf{A}$ .

- a. Express the pseudoinverse  $\mathbf{A}^+$  in terms of the SVD component matrices given above.

[1 mark]

- b. Use your answer to (a) to derive an expression for the *model (parameter) resolution operator*  $\mathbf{R}_m$  in terms of the SVD component matrices.

[2 marks]

- c. Use your answer to (a) to derive an expression for the *data resolution operator*  $\mathbf{R}_d$  in terms of the SVD component matrices.

[2 marks]

- ii) Briefly explain how to obtain regularised solutions to a linear inverse problem by using the SVD. Include a brief intuitive explanation of *why* this works.

[5 marks]