

# ACM41020 Assignment 3

## Optimisation

**DUE: 10AM FRIDAY 8<sup>TH</sup> NOVEMBER 2024**

At several points in this assignment we will consider the particular case where

$$S = \begin{bmatrix} 5 & -3 \\ -3 & 5 \end{bmatrix}, \quad A = \begin{bmatrix} 1 \\ 3 \end{bmatrix}.$$

### 1. Constrained Optimisation

We wish to minimise the quadratic function  $F(\mathbf{x}) = \frac{1}{2} \mathbf{x}^T S \mathbf{x}$  subject to  $A^T \mathbf{x} = b$ .

- What is the Lagrangian  $L(\mathbf{x}, \lambda)$  for this problem?
- What are the equations “derivative of  $L = \text{zero}$ ”?
- Solve those equations to find the location  $\mathbf{x}^* = (x_1^*, x_2^*)$  of the minimum  $F_{\min} = F(\mathbf{x}^*)$ .  
Solve also for the multiplier  $\lambda^*$ .
- Draw or plot the constraint line  $A^T \mathbf{x} = b$  tangent to the ellipse  $F(\mathbf{x}) = F_{\min}$ .
- Verify that the derivative of the minimum cost is  $\frac{\partial F_{\min}}{\partial b} = \lambda^*$ .

### 2. Unconstrained Optimisation Using Gradient Descent

- Derive an expression in terms of  $S$  and  $\mathbf{x}$  for the optimal step size for gradient descent applied to find the unconstrained minimum of a quadratic function  $F(\mathbf{x}) = \frac{1}{2} \mathbf{x}^T S \mathbf{x}$  with a generic symmetric and positive-definite matrix  $S$ .
- By considering the worst case scenario (worst  $\mathbf{x}_i$ ) for the decrease in the error  $e_k = ||\mathbf{x}_k - \mathbf{x}^*||$  from one step to the next, derive an expression for the optimal step size in that case in terms of the eigenvalues of  $S$ .
- What condition must  $\mathbf{x}_i$  satisfy to be in this worst-case scenario?
- Working with the specific matrix  $S$  above and starting from  $\mathbf{x}_0 = (3, 5)$  compute six steps using gradient descent with the optimal step size.
- Sketch/plot the contours defined by the ellipses  $F(\mathbf{x}) = F(\mathbf{x}_i)$  for  $i = 0, 1, \dots, 6$ . On the same plot indicate the points  $\mathbf{x}_0, \mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3, \mathbf{x}_4, \mathbf{x}_5, \mathbf{x}_6$  obtained in part (d) and the lines defined by the set of worst-case points identified in part (c).
- Compute the error  $e_k$  in each step. By considering how the error decreases over 3 steps verify that the iteration is converging linearly to the solution  $\mathbf{x}^*$ .

### 3. Gradient Descent with Momentum

- Working with the matrix  $S$  above and starting from  $\mathbf{x}_0 = (3, 5)$  compute six steps using gradient descent with momentum. Use the values for the optimal step size  $\alpha$  and momentum parameter  $\beta$  in terms of the eigenvalues of  $S$  as given in the lectures.
- Sketch/plot the contours defined by the ellipses  $F(\mathbf{x}) = F(\mathbf{x}_i)$  for  $i = 0, 1, \dots, 6$ . On the same plot indicate the points  $\mathbf{x}_0, \mathbf{x}_1, \mathbf{x}_2, \mathbf{x}_3, \mathbf{x}_4, \mathbf{x}_5, \mathbf{x}_6$ .
- Compute the error  $e_k$  in each step.
- Briefly (1-2 sentence) compare and contrast the results with and without momentum.

**Note:** The questions in this assignment can be completed with by-hand calculations, but if you wish you may use a computer to help with any number-crunching.