Optimisation methods – Homework 1

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1 Exercise 1

1.1 Gradient and Hessian

The gradient and the Hessian for f are the following:

$$\nabla f = \begin{bmatrix} 2x_1 + x_2 \cdot \cos(x_1) \\ 9x_2^2 + \sin(x_1) \end{bmatrix}$$
$$H_f = \begin{bmatrix} 2 - x_2 \cdot \sin(x_1) & \cos(x_1) \\ \cos(x_1) & 18x_2 \end{bmatrix}$$

1.2 Taylor expansion

$$\begin{split} f(h) &= 0 + \left\langle \begin{bmatrix} 0+0\\ 0+0 \end{bmatrix}, \begin{bmatrix} h_1\\ h_2 \end{bmatrix} \right\rangle + \frac{1}{2} \left\langle \begin{bmatrix} 2-0 & 1\\ 1 & 0 \end{bmatrix} \begin{bmatrix} h_1\\ h_2 \end{bmatrix}, \begin{bmatrix} h_1\\ h_2 \end{bmatrix} \right\rangle + O(\|h\|^3) \\ f(h) &= \frac{1}{2} \left\langle \begin{bmatrix} 2h_1+h_2\\ h_1 \end{bmatrix}, \begin{bmatrix} h_1\\ h_2 \end{bmatrix} \right\rangle + O(\|h\|^3) \\ f(h) &= \frac{1}{2} \left(2h_1^2 + 2h_1h_2 \right) + O(\|h\|^3) \\ f(h) &= h_1^2 + h_1h_2 + O(\|h\|^3) \end{split}$$

2 Exercise 2

2.1 Gradient and Hessian

For A symmetric, we have:

$$\frac{d}{dx}\langle b, x \rangle = \langle b, \cdot \rangle = b$$
$$\frac{d}{dx}\langle Ax, x \rangle = 2\langle Ax, \cdot \rangle = 2Ax$$

Then:

$$\nabla J = Ax - b$$
$$H_J = \frac{d}{dx} \nabla J = A$$

2.2 First order necessary condition

It is a necessary condition for a minimizer x^* of J that:

$$\nabla J(x^*) = 0 \Leftrightarrow Ax^* = b$$

2.3 Second order necessary condition

It is a necessary condition for a minimizer x^* of J that:

$$\nabla^2 J(x^*) \ge 0 \Leftrightarrow A$$
 is positive semi-definite

2.4 Sufficient conditions

It is a sufficient condition for x^* to be a minimizer of J that the first necessary condition is true and that:

$$\nabla^2 J(x^*) > 0 \Leftrightarrow A$$
 is positive definite

2.5 Does $\min_{x \in \mathbb{R}^n} J(x)$ have a unique solution?

Not in general. If for example we consider A and b to be only zeros, then J(x) = 0 for all $x \in \mathbb{R}^n$ and thus J would have an infinite number of minimizers.

However, for if A would be guaranteed to have full rank, the minimizer would be unique because the first order necessary condition would hold only for one value x^* . This is because the linear system $Ax^* = b$ would have one and only one solution (due to A being full rank).

3 Exercise 3

3.1 Quadratic form

f(x, y) can be written in quadratic form in the following way:

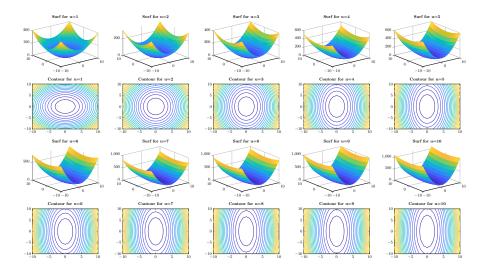
$$f(v) = \frac{1}{2} \left\langle \begin{bmatrix} 2 & 0 \\ 0 & 2\mu \end{bmatrix} v, v \right\rangle + \left\langle \begin{bmatrix} 0 \\ 0 \end{bmatrix}, x \right\rangle$$

where:

$$v = \begin{bmatrix} x \\ y \end{bmatrix}$$

3.2 Matlab implementation with surf and contour

The graphs generated by MATLAB are shown below:



Isolines get stretched along the y axis as μ increases. For a large μ , points well far away from the axes could be a problem since picking search directions and steps using a naive gradient based method iterations will zig-zag to the minimizer reaching it slowly.