

NUMERICAL OPTIMISATION ASSIGNMENT 1: SOLUTION

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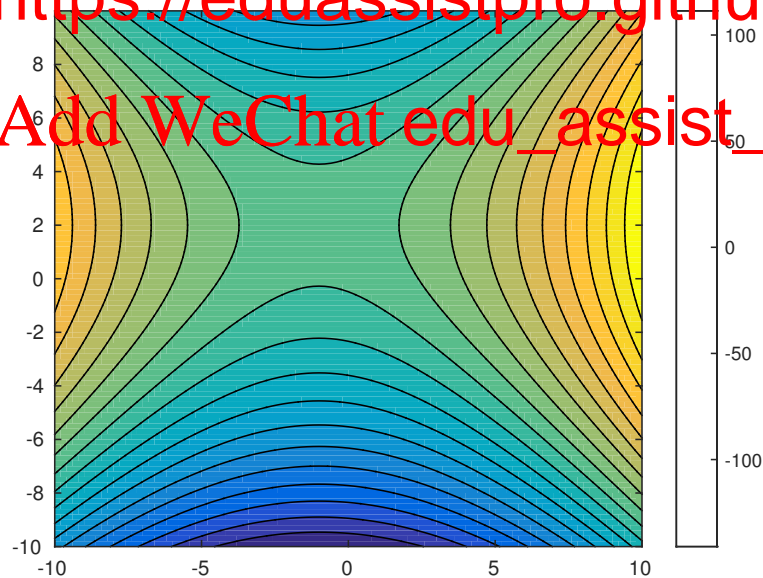
EXERCISE 1.

(a) Function and its contours.

```
n = 100;  
x = linspace(-1,1,n+1); y = x;  
[X,Y] = meshgrid(x,y);  
  
% Plot the function of two variables  
alpha = 10;  
figure, contourf(alpha*X, alpha*Y, f(alpha*X,alpha*Y), 20)  
colorbar();  
saveas(gcf, 'c
```

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(b)

$$f(x, y) = 2x + 4y + x^2 - 2y^2 = (x + 1)^2 - 2(y - 1)^2 + 1$$

For $c \in \mathbb{R}$ we have the contour sets

$$S_c = \{(x, y) : (x + 1)^2 - 2(y - 1)^2 + 1 = c\}.$$

For $c = 1$ we have the point

$$S_1 = \{(-1, 1)\}.$$

Finally, for $c \neq 1$ we have that the contour set is a hyperbola that can be represented implicitly by

$$S_c = \{(x, y) : \frac{(x+1)^2}{c-1} - \frac{(y-1)^2}{\frac{1}{2}(c-1)} = 1\}.$$

(c) Gradient.

$$\frac{\partial f}{\partial x} = 2x + 2, \quad \frac{\partial f}{\partial y} = -4y + 4$$

The only stationary point is $x_0 = (-1, 1)$. To classify it we consider the Hessian at this point.

$$\nabla^2 f(x_0) = \begin{pmatrix} 2 & 0 \\ 0 & -4 \end{pmatrix}$$

Since the eigenvalues of the Hessian are $\{2, -4\}$, we have a saddle point.

EXERCISE 2.

(a) We need to show that the eigenvalues of A are non-negative. Consider an eigenvalue λ . The eigenvalue problem has the form

$$Ax = \lambda x, \quad x \neq 0,$$

with λ the eigenvalue and $x \neq 0$ the corresponding eigenvector. We left multiply each side of this equation by x^T , to obtain

Dividing by $x^T x$ we obtain

$$\lambda = \frac{x^T Ax}{x^T x}$$

Plugging in $A = B^T B$ we obtain

$$\lambda = \frac{x^T B^T B x}{x^T x} = \frac{\|Bx\|^2}{\|x\|^2} \geq 0$$

which implies that all eigenvalues are non-negative and hence A is positive semidefinite.

(b) We would like to show that

$$f(y + \alpha(x - y)) - \alpha f(x) - (1 - \alpha)f(y) \leq 0$$

We have $f(x) = x^T Q x$ for Q symmetric positive semidefinite:

$$\begin{aligned} (y + \alpha(x - y))^T Q (y + \alpha(x - y)) - \alpha x^T Q x - (1 - \alpha)y^T Q y &= \\ = y^T Q y - (1 - \alpha)y^T Q y - \alpha x^T Q x + 2\alpha x^T Q y - 2\alpha y^T Q y - \alpha^2 x^T Q x - \alpha^2 x^T Q y + \alpha^2 y^T Q y &= \\ = y^T Q y \alpha(\alpha - 1) + x^T Q x \alpha(\alpha - 1) + 2\alpha x^T Q y(1 - \alpha) &= \\ = \alpha(\alpha - 1)[y^T Q y + x^T Q x - 2x^T Q y] &= \\ = \alpha(\alpha - 1)(x - y)^T Q (x - y). \end{aligned}$$

Now, since $\alpha \in [0, 1]$ and Q is positive semidefinite, we have

$$\underbrace{\alpha(\alpha - 1)}_{\leq 0} \underbrace{(x - y)^T Q (x - y)}_{\geq 0} \leq 0.$$