

Seminar 9/2.

Positive definiteness requires that $\mathbf{x}'\mathbf{A}\mathbf{x} > 0$ and positive semidefiniteness requires $\mathbf{x}'\mathbf{B}\mathbf{x} \geq 0$ for all \mathbf{x} . Therefore

$$\begin{aligned}\mathbf{x}'\mathbf{A}\mathbf{x} + \mathbf{x}'\mathbf{B}\mathbf{x} &> 0 \\ \mathbf{x}'(\mathbf{A} + \mathbf{B})\mathbf{x} &> 0 \\ \mathbf{x}'(\mathbf{A} + \mathbf{B})\mathbf{x} &> 0\end{aligned}$$

The sum of \mathbf{A} and \mathbf{B} is therefore positive definite.

How about the product of a positive definite matrix and a positive semidefinite matrix? Let:

$$\mathbf{A} = \begin{bmatrix} a_1 & A \\ A & a_2 \end{bmatrix} \quad \text{and} \quad \mathbf{B} = \begin{bmatrix} b_1 & B \\ B & b_2 \end{bmatrix}$$

In order for \mathbf{A} to be positive definite, a_1 and $|\mathbf{A}|$ must be strictly positive. A can be negative as long as A^2 is less than $a_1 a_2$. If we now calculate \mathbf{AB} we find that:

$$\mathbf{AB} = \begin{bmatrix} a_1 b_1 + AB & a_1 B + A b_2 \\ A b_1 + a_2 B & AB + a_2 b_2 \end{bmatrix}$$

Clearly \mathbf{AB} is not symmetric and therefore not definite in any way. However the quadratic form $\mathbf{x}'\mathbf{AB}\mathbf{x}$ has an equivalent form with a matrix that is symmetric.

Compendium problem 1-13.

a) We have

$$\begin{aligned}
& \left(\mathbf{X} + \frac{1}{2}\mathbf{A}^{-1}\mathbf{B}'\right)' \mathbf{A} \left(\mathbf{X} + \frac{1}{2}\mathbf{A}^{-1}\mathbf{B}'\right) - \frac{1}{4}\mathbf{B}\mathbf{A}^{-1}\mathbf{B}' \\
& \left(\mathbf{X}' + \left(\frac{1}{2}\mathbf{A}^{-1}\mathbf{B}'\right)'\right) \left(\mathbf{A}\mathbf{X} + \frac{1}{2}\mathbf{B}'\right) - \frac{1}{4}\mathbf{B}\mathbf{A}^{-1}\mathbf{B}' \\
& \left(\mathbf{X}' + \left(\frac{1}{2}\mathbf{B}(\mathbf{A}^{-1})'\right)\right) \left(\mathbf{A}\mathbf{X} + \frac{1}{2}\mathbf{B}'\right) - \frac{1}{4}\mathbf{B}\mathbf{A}^{-1}\mathbf{B}' \\
& \mathbf{X}'\mathbf{A}\mathbf{X} + \frac{1}{2}\mathbf{X}'\mathbf{B}' + \left(\frac{1}{2}\mathbf{B}(\mathbf{A}^{-1})'\right)\mathbf{A}\mathbf{X} + \left(\frac{1}{2}\mathbf{B}(\mathbf{A}^{-1})'\right)\frac{1}{2}\mathbf{B}' - \frac{1}{4}\mathbf{B}\mathbf{A}^{-1}\mathbf{B}' \\
& \mathbf{X}'\mathbf{A}\mathbf{X} + \frac{1}{2}\mathbf{X}'\mathbf{B}' + \frac{1}{2}\mathbf{B}\mathbf{X} + \frac{1}{4}\mathbf{B}\mathbf{A}^{-1}\mathbf{B}' - \frac{1}{4}\mathbf{B}\mathbf{A}^{-1}\mathbf{B}' \\
& \mathbf{X}'\mathbf{A}\mathbf{X} + \mathbf{B}\mathbf{X}
\end{aligned}$$

b) We have

$$\begin{aligned}
\frac{d}{d\mathbf{X}}(\mathbf{X}'\mathbf{A}\mathbf{X} + \mathbf{B}\mathbf{X}) &= \mathbf{X}'\mathbf{A} + \mathbf{A}\mathbf{X} + \mathbf{B} \\
&= \mathbf{A}\mathbf{X} + \mathbf{A}\mathbf{X} + \mathbf{B} \\
&= 2\mathbf{A}\mathbf{X} + \mathbf{B} = \mathbf{0} \\
&\Downarrow \\
\mathbf{X} &= -\frac{1}{2}\mathbf{A}^{-1}\mathbf{B}
\end{aligned}$$

That this is a minimum follows from \mathbf{A} being positive definite.

Exam 2008 problem 1(c)

$$\mathbf{A} = \begin{bmatrix} 0 & 2 & 0 \\ 2 & 3 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

We have that:

$$D_1 = |0|$$

This implies that A is not positive definite. We have that:

$$D_2 = \det \begin{vmatrix} 0 & 2 \\ 2 & 3 \end{vmatrix} = -4$$

This is one of the principal subdeterminants, so A is not positive semidefinite. As $D_1 = 0$ we can rule out negative definiteness. For $r = 1$,

the principal subdeterminants are all non-negative so $(-1)^1 \Delta_r = [0 \ -3 \ 0]$ so we can also rule out negative semidefiniteness.

Alternatively: Note that \mathbf{A} is symmetric. Then we can use eigenvalues. It is straight forward to calculate that the eigenvalues for \mathbf{A} are given by -1, 0 and 4. With eigenvalues of different signs we now that A is indefinite.

Now we add the constraint that

$$\mathbf{D} = \begin{bmatrix} 0 & 0 & 16 & 0 & 1 \\ 0 & 0 & 1 & 2 & 0 \\ 16 & 1 & 0 & 2 & 0 \\ 0 & 2 & 2 & 3 & 0 \\ 1 & 0 & 0 & 0 & 0 \end{bmatrix}$$

There are 3 variables and two constraints. We must therefore calculate the following determinant.

$$D_3 = \begin{vmatrix} 0 & 0 & 16 & 0 & 1 \\ 0 & 0 & 1 & 2 & 0 \\ 16 & 1 & 0 & 2 & 0 \\ 0 & 2 & 2 & 3 & 0 \\ 1 & 0 & 0 & 0 & 0 \end{vmatrix} = -5$$

Fist we check if it is positive definite. We note that there are $m = 2$ constraints and calculate that $(-1)^m D_3 = -5$. Thus the constrained quadratic form is not positive definite. Then we calculate $(-1)^3 D_3 = 5$ and the constrained quadratic form is negative definite.

Differentiation

We have that $g: \mathbb{R}^m \rightarrow \mathbb{R}$. We have that the Hessian is $\mathbf{H}(\mathbf{y})$ where the element in the i th row and the j th column is $g_{i,j}(\mathbf{y})$. We are asked to find the Hessian of $f(\mathbf{x}) = g(\mathbf{Ax})$. Don't think too hard about this stuff. Just use the chain rule. First we get the gradient.

$$\begin{aligned} \mathbf{h}(\mathbf{x}) = \nabla f(\mathbf{Ax}) &= \left[\sum_{i=1}^m g'_i(\mathbf{Ax})a_{i,1}, \sum_{i=1}^m g'_i(\mathbf{Ax})a_{i,2}, \dots, \sum_{i=1}^m g'_i(\mathbf{Ax})a_{i,n} \right] = \underbrace{\left[\nabla g(\mathbf{Ax}) \right]}_{1 \times m} \underbrace{\mathbf{A}}_{m \times n} \\ &= \underbrace{\mathbf{A}^T}_{n \times m} \underbrace{\nabla g(\mathbf{Ax})}_{m \times 1} \end{aligned}$$

Then take the derivative of this expression.

$$\begin{aligned} \underbrace{\mathbf{h}'(\mathbf{x})}_{n \times n} &= \begin{bmatrix} \frac{\partial h_1}{\partial x_1} & \frac{\partial h_1}{\partial x_2} & \dots & \frac{\partial h_1}{\partial x_n} \\ \frac{\partial h_2}{\partial x_1} & \frac{\partial h_2}{\partial x_2} & \dots & \frac{\partial h_2}{\partial x_n} \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\partial h_n}{\partial x_1} & \frac{\partial h_n}{\partial x_2} & \dots & \frac{\partial h_n}{\partial x_n} \end{bmatrix} = \mathbf{A}^T \underbrace{\begin{bmatrix} g''_{11}(\mathbf{Ax}) & g''_{12}(\mathbf{Ax}) & \dots & g''_{1m}(\mathbf{Ax}) \\ g''_{21}(\mathbf{Ax}) & g''_{22}(\mathbf{Ax}) & \dots & g''_{2m}(\mathbf{Ax}) \\ \vdots & \vdots & \ddots & \vdots \\ g''_{m1}(\mathbf{Ax}) & g''_{m2}(\mathbf{Ax}) & \dots & g''_{m2}(\mathbf{Ax}) \end{bmatrix}}_{m \times m} \mathbf{A} \\ &= \mathbf{A}^T \mathbf{H}(\mathbf{Ax}) \mathbf{A} \end{aligned}$$

Problem 2-07

Let $f(x, y) = (\ln x)^a (\ln y)^b$, defined for $x > 1$, $y > 1$. Assume that $a > 0$, $b > 0$ and $a + b < 1$. Compute the Hessian matrix \mathbf{H} of f and show that f is strictly concave.

We have that

$$\begin{aligned}
f''_{x,x}(x,y) &= \frac{-(1-a)a(\ln x)^{a-2}(\ln y)^b}{x^2} - \frac{a(\ln x)^{a-1}(\ln(y))^b}{x^2} \\
f''_{y,y}(x,y) &= \frac{-(1-b)b(\ln x)^a(\ln y)^{b-2}}{y^2} - \frac{b(\ln x)^a(\ln y)^{b-1}}{y^2} \\
f''_{x,y}(x,y) &= \frac{ab(\ln x)^{a-1}(\ln y)^{b-1}}{xy}
\end{aligned}$$

$f''_{x,x}$ is clearly negative. The Hessian is easy to form after having calculated these expressions. Noting that $f''_{x,x}(x,y)$ and $f''_{y,y}(x,y)$ has common factors simplify the calculations. The determinant of the Hessian is

$$\begin{aligned}
|\mathbf{H}| &= \frac{ab(\ln x)^{2a-2}(\ln y)^{2b-2}(1-a+\ln x)(1-b+\ln y)}{x^2y^2} - \frac{a^2b^2(\ln x)^{2a-2}(\ln y)^{2b-2}}{x^2y^2} \\
&= \frac{ab}{x^2y^2}(\ln x)^{2a-2}(\ln y)^{2b-2} \left((1-a+\ln x)(1-b+\ln y) - ab \right) \\
&= \frac{ab}{x^2y^2}(\ln x)^{2a-2}(\ln y)^{2b-2} \underbrace{\left(1 - (a+b) + (1-a)\ln y + (1-b)\ln x + \ln(x+y) \right)}_{\text{positive by assumption}} > 0
\end{aligned}$$