

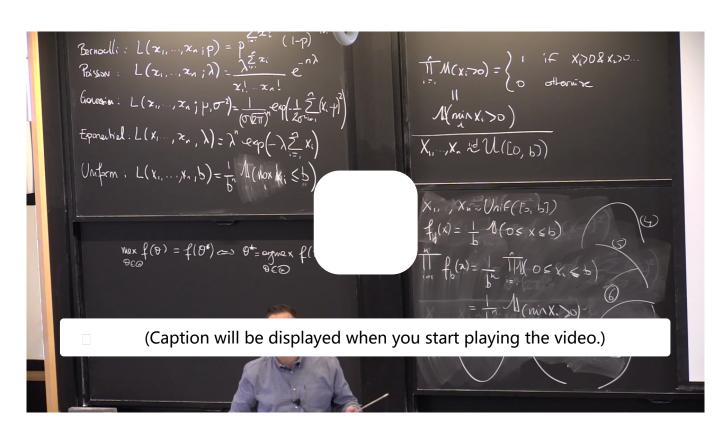
<u>Lecture 9: Introduction to</u>

8. Review: Gradients and Hessians;

Concavity in Higher dimensions

课程 □ Unit 3 Methods of Estimation □ Maximum Likelihood Estimation

8. Review: Gradients and Hessians; Concavity in Higher dimensions Concavity in Higher Dimensions: Gradients, Hessians, Semi-Definiteness



Start of transcript. Skip to the end.

So here, you can see we have one example where

the function we're looking at is actually a function of two parameters-- one living on the real line

and one living on the positive real line.

And so we're going to have to talk about convex and concave

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Multivariable Calculus Review: Compute the Gradient

1/1 point (graded) Let

denote a **differentiable** function. The **gradient** of $m{f}$ is the vector-valued function

Consider $f(\theta)=-c_1\theta_1^2-c_2\theta_2^2-c_3\theta_3^2$ where $c_1,c_2,c_3>0$ are positive real numbers.

Compute the gradient ∇f .

(Enter your answer as a vector, e.g., type [3,2,x] for the vector $\begin{pmatrix} 3 \\ 2 \\ x \end{pmatrix}$. Note the square brackets, and commas as separators. Enter **c_i** for

 c_i , theta_i for $heta_i$.)

$$\nabla f = \begin{bmatrix} -2*c_1*theta_1, -2*c_2*theta_2, -2*c_3*theta_3 \end{bmatrix}$$
 \Box Answer: $[-2*c_1*theta_1, -2*c_2*theta_2, -2*c_3*theta_3]$

STANDARD NOTATION

Solution:

$$egin{aligned} f\left(heta
ight) &=& -c_1 heta_1^2 - c_2 heta_2^2 - c_3 heta_3^2 \ \left.egin{aligned} rac{\partial f}{\partial heta_1} \ rac{\partial f}{\partial heta_2} \ rac{\partial f}{\partial heta_3} \end{aligned}
ight)
ight|_{ heta} &=& egin{pmatrix} -2c_1 heta_1 \ -2c_2 heta_2 \ -2c_3 heta_3 \end{pmatrix}. \end{aligned}$$

提交

你已经尝试了1次(总共可以尝试3次)

☐ Answers are displayed within the problem

Multivariable Calulus Review: Compute the Hessian Matrix

1/1 point (graded) As above, let

$$f\colon \, \mathbb{R}^d \, o \, \mathbb{R} \, \, heta = egin{pmatrix} heta_1 \ heta_2 \ dots \ heta_d \end{pmatrix} \, \mapsto \, f(heta) \, .$$

denote a twice-differentiable function.

The **Hessian** of f is the matrix

$$\mathbf{H} f: \mathbb{R}^d \to \mathbb{R}^{d \times d}$$

whose entry in the i-th row and j-th column is defined by

$$(\mathbf{H}\,f)_{ij} \;:=\; rac{\partial^2}{\partial heta_i\partial heta_j}f, \quad 1\leq i,j\leq d.$$

Consider the same function $f(\theta)=-c_1\theta_1^2-c_2\theta_2^2-c_3\theta_3^2$ where $c_1,\,c_2,\,c_3>0$ as in the previous problem. Compute the Hessian matrix ${\bf H}f$.

(Enter your answer as a matrix, e.g. by typing **[[1,2],[5*x,y-1]]** for the matrix $\begin{pmatrix} 1 & 2 \\ 5x & y-1 \end{pmatrix}$. Note the square brackets, and commas as separaters.)

$$\mathbf{H}f = [[-2*c_1,0,0],[0,-2*c_2,0],[0,0,-2*c_3]]$$

Answer: [[-2*c 1,0,0],[0,-2*c 2,0],[0,0,-2*c 3]]

STANDARD NOTATION

Solution:

Recall from the previous problem:

One way to compute the Hessian is to start will in j-th column of the Hessian matrix by the gradient of the j-th component of ∇f . We obtain:

提交

你已经尝试了2次(总共可以尝试3次)

☐ Answers are displayed within the problem

Semi-Definiteness

2/3 points (graded)

A symmetric (real-valued) d imes d matrix ${f A}$ is **positive semi-definite** if

$$\mathbf{x}^T \mathbf{A} \mathbf{x} \geq 0 \quad \text{for all } \mathbf{x} \in \mathbb{R}^d.$$

If the inequality above is strict, i.e. if $\mathbf{x}^T \mathbf{A} \mathbf{x} > \mathbf{0}$ for all non-zero vectors $\mathbf{x} \in \mathbb{R}^d$, then \mathbf{A} is **positive definite** .

Analogously, a symmetric (real-valued) $d \times d$ matrix \mathbf{A} is negative semi-definite (resp. negative definite) if $\mathbf{x}^T \mathbf{A} \mathbf{x}$ is non-positive (resp. negative) for all $\mathbf{x} \in \mathbb{R}^d - \{\mathbf{0}\}$.

Note that by definition, positive (or negative) definiteness implies positive (or negative) semi-definiteness.

Consider the same function as in the problems above:

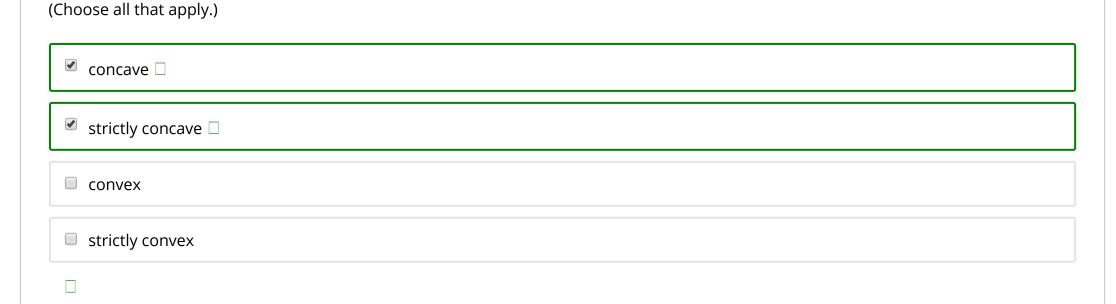
$$f(heta) \; = \; -c_1 heta_1^2 - c_2 heta_2^2 - c_3 heta_3^2 \quad ext{where}\, c_1, c_2, c_3 > 0.$$

Compute
$$\mathbf{x}^T$$
 $(\mathbf{H}f)$ \mathbf{x} where $\mathbf{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$.

$$\mathbf{x}^{T} \ (\mathbf{H}f) \ \mathbf{x} = \begin{bmatrix} -2*c_{1}*x_{1}^{2} - 2*c_{2}*x_{2}^{2} - 2*c_{3}*x_{3}^{2} \\ -2 \cdot c_{1} \cdot x_{1}^{2} - 2 \cdot c_{2} \cdot x_{2}^{2} - 2 \cdot c_{3} \cdot x_{3}^{2} \end{bmatrix}$$
 \(\text{Answer: } -2*c_{1}*x_{1}^{2} - 2*c_{2}*x_{2}^{2} - 2*c_{3}*x_{3}^{2} \)

The matrix $\mathbf{H}f$ is (Choose all that apply.)

- positive semi-definitepositive definite
 - \square negative semi-definite \square
 - ✓ negative definite □



Solution:

Hence, the function $m{f}$ is

Recall from the previous problem that

$$\mathbf{H}f(heta) \;=\; egin{pmatrix} -2c_1 & 0 & 0 \ 0 & 2c_2 & 0 \ 0 & 0 & -2c_3 \end{pmatrix}.$$

Then

$$egin{array}{lll} \mathbf{x}^T & (\mathbf{H}f) \; \mathbf{x} \; = \; (x_1 \quad x_2 \quad x_3 \,) egin{pmatrix} -2c_1 & 0 & 0 & \ 0 & 2c_2 & 0 & \ 0 & 0 & 2c_3 \end{pmatrix} egin{pmatrix} x_1 \ x_2 \ x_3 \end{pmatrix} \ & = \; -2c_1x_1^2 - 2c_2x_2^2 - 2c_3x_3^2 \, < \, 0. \end{array}$$

Since $c_1, c_2, c_3 > 0$, this means the $\mathbf{H}f$ is negative definite, (also negative semi-definite), and hence f is strictly concave (also concave).

提交

你已经尝试了2次(总共可以尝试2次)

☐ Answers are displayed within the problem

讨论

主题: Unit 3 Methods of Estimation:Lecture 9: Introduction to Maximum Likelihood Estimation / 8. Review: Gradients and Hessians; Concavity in Higher dimensions

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显示讨论