

Multivariate Calculus & Constrained Optimization

Lecture 14

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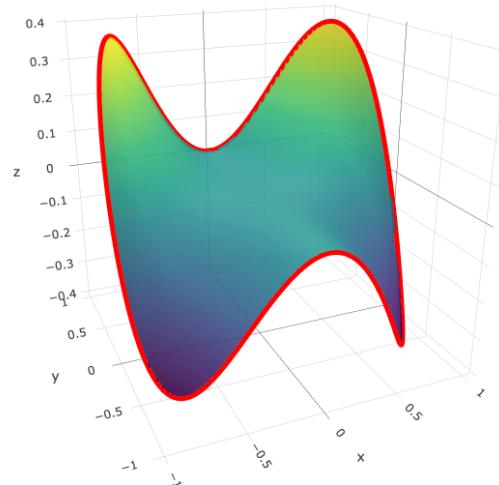
1

the problem

maximize $f(x, y) = x^2y$

on the set of all values $\underbrace{x^2 + y^2 = 1}_{\text{unit circle}}$

when projected onto the surface

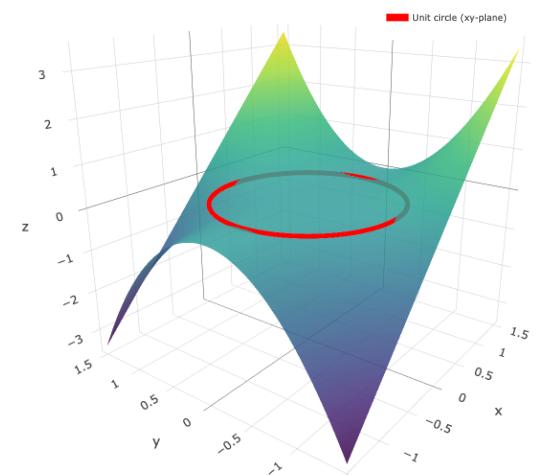


3

the problem

maximize $f(x, y) = x^2y$

on the set of all values $\underbrace{x^2 + y^2 = 1}_{\text{unit circle}}$



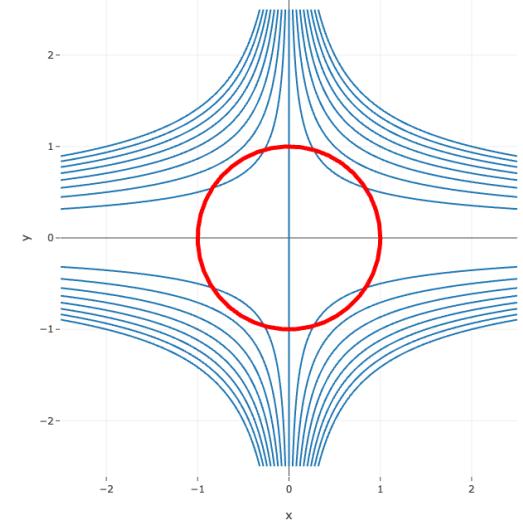
2

the problem

maximize $f(x, y) = x^2y$

on the set of all values $\underbrace{x^2 + y^2 = 1}_{\text{unit circle}}$

easier to work with contour map
which limit the view to the input space



4

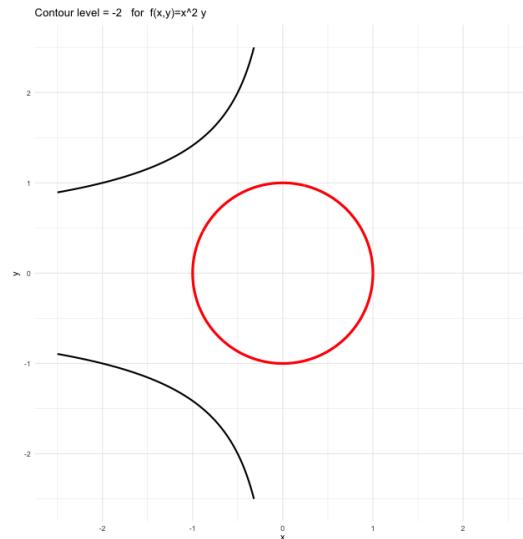
the problem

$$\text{maximize } f(x, y) = x^2 y$$

on the set of all values $\underbrace{x^2 + y^2 = 1}_{\text{unit circle}}$

easier to work with contour maps
which limit the view to the input space

$$\text{contour level } c \iff \{(x, y) : f(x, y) = c\}$$



5

the problem

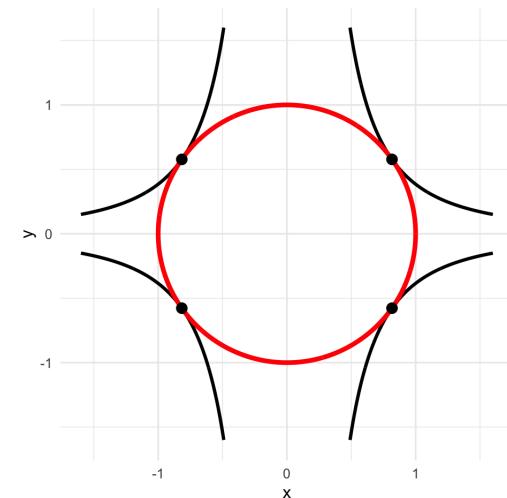
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easier to work with contour maps
which limit the view to the input space

$$\text{contour level } c \iff \{(x, y) : f(x, y) = c\}$$

use the gradient



6

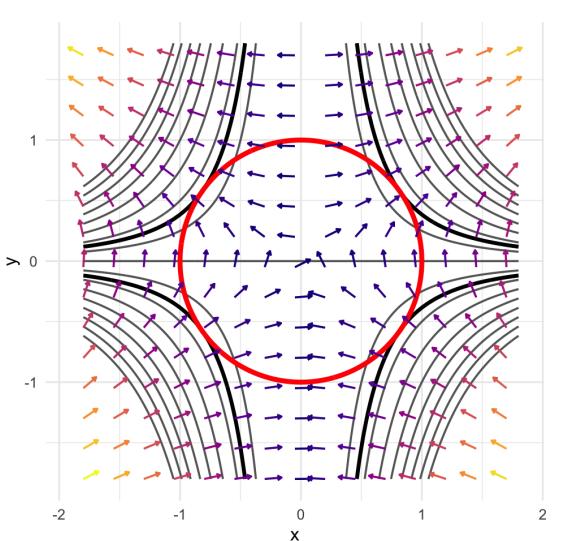
the problem

$$\text{maximize } f(x, y) = x^2 y$$

on the set of all values $\underbrace{x^2 + y^2 = 1}_{\text{unit circle}}$

$$f(x, y) = c$$

gradient vector are perpendicular
when crossing a contour line



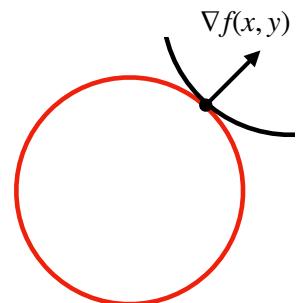
7

the problem

$$\text{maximize } f(x, y) = x^2 y$$

on the set of all values $\underbrace{x^2 + y^2 = 1}_{\text{unit circle}}$

$$f(x, y) = c$$



8

the problem

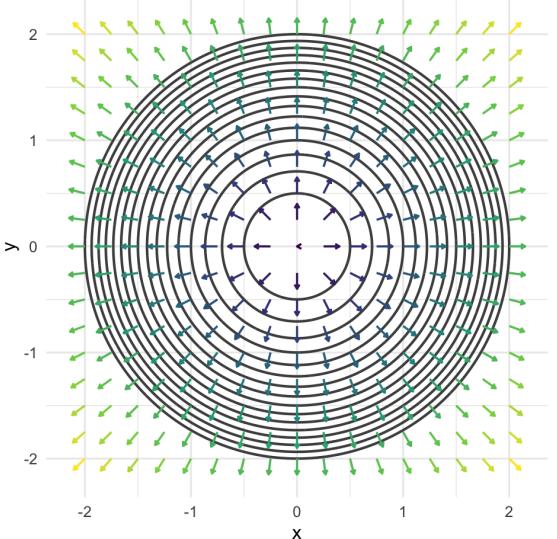
$$\text{maximize } f(x, y) = x^2 y$$

on the set of all values $\underbrace{x^2 + y^2 = 1}_{\text{unit circle}}$

$$f(x, y) = c$$

$$\text{now let } g(x, y) = x^2 + y^2$$

gradient vector are perpendicular
when crossing a contour line



9

the problem

$$\text{maximize } f(x, y) = x^2 y$$

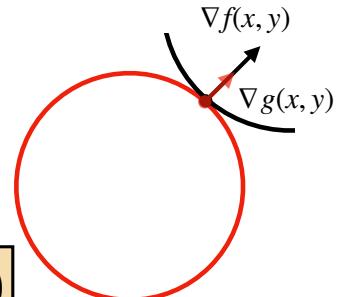
on the set of all values $\underbrace{x^2 + y^2 = 1}_{\text{unit circle}}$

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

∇g is proportional to ∇f

$$\nabla f(x_{\max}, y_{\max}) = \lambda \nabla g(x_{\max}, y_{\max})$$

where λ is the Lagrange multiplier



10

solution to the problem

$$\nabla f(x_{\max}, y_{\max}) = \lambda \nabla g(x_{\max}, y_{\max})$$

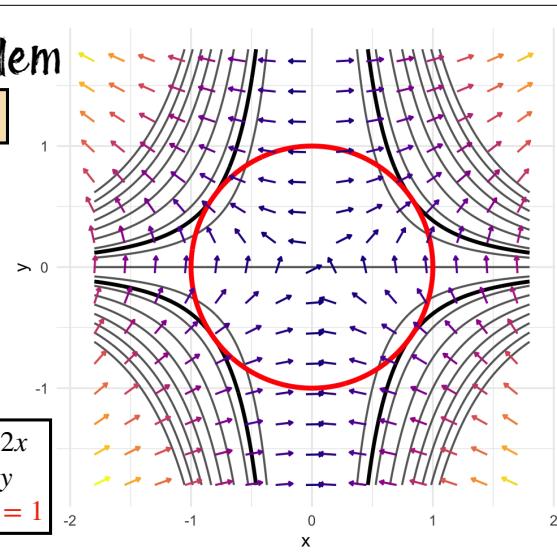
$$f(x, y) = x^2 y$$

$$\nabla f = \left(\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right) = \begin{bmatrix} 2xy \\ x^2 \end{bmatrix}$$

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

$$\nabla g = \left(\frac{\partial g}{\partial x}, \frac{\partial g}{\partial y} \right) = \begin{bmatrix} 2x \\ 2y \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} 2xy \\ x^2 \end{bmatrix} = \lambda \begin{bmatrix} 2x \\ 2y \end{bmatrix} \Rightarrow \begin{cases} 2xy = \lambda 2x \\ x^2 = \lambda 2y \\ x^2 + y^2 = 1 \end{cases}$$



11

solution to the problem

$$\nabla f(x_{\max}, y_{\max}) = \lambda \nabla g(x_{\max}, y_{\max})$$

$$f(x, y) = x^2 y$$

$$\nabla f = \left(\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right) = \begin{bmatrix} 2xy \\ x^2 \end{bmatrix}$$

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

$$\nabla g = \left(\frac{\partial g}{\partial x}, \frac{\partial g}{\partial y} \right) = \begin{bmatrix} 2x \\ 2y \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} 2xy \\ x^2 \end{bmatrix} = \lambda \begin{bmatrix} 2x \\ 2y \end{bmatrix}$$

assuming $x \neq 0$:

$$2xy = \lambda 2x \implies 2y = \lambda 2 \implies y = \lambda$$

$$x^2 = \lambda 2y \implies x^2 = 2y^2$$

$$x^2 + y^2 = 1 \implies 2y^2 + y^2 = 1$$

$$3y^2 = 1$$

$$y = \pm \sqrt{1/3}$$

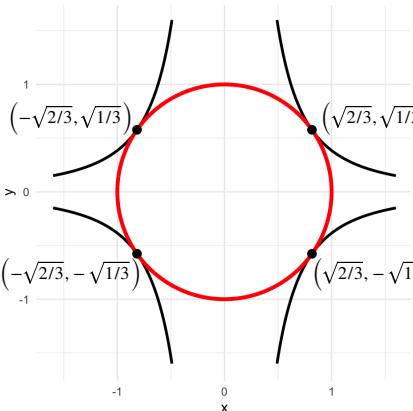
$$\implies x = 2y^2 = 2(\sqrt{1/3})^2 \implies x = \pm \sqrt{2/3}$$

check: if $x = 0$ then $y = 0$, constraint is not satisfied

⇒ four possible points satisfying the constraint

12

solution to the problem



assuming $x \neq 0$:

$$\begin{aligned} 2xy = \lambda 2x &\implies 2y = \lambda 2 \implies y = \lambda \\ x^2 = \lambda 2y &\implies x^2 = 2y^2 \\ x^2 + y^2 = 1 &\implies 2y^2 + y^2 = 1 \\ &\implies 3y^2 = 1 \\ y &= \pm\sqrt{1/3} \end{aligned}$$

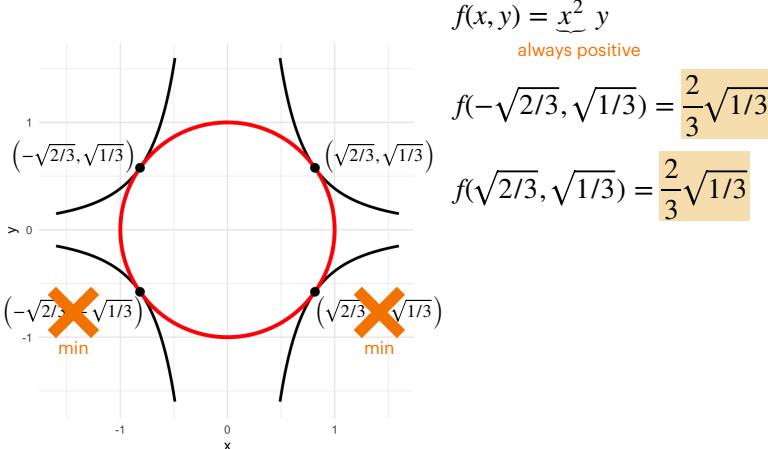
$$x = 2y^2 = 2(\sqrt{1/3})^2 \implies x = \pm\sqrt{2/3}$$

check: if $x = 0$ then $y = 0$, constraint is not satisfied

\implies four possible points satisfying the constraint

13

solution to the problem



$$f(x, y) = \underbrace{x^2}_y$$

always positive

$$f(-\sqrt{2/3}, \sqrt{1/3}) = \frac{2}{3}\sqrt{1/3}$$

$$f(\sqrt{2/3}, \sqrt{1/3}) = \frac{2}{3}\sqrt{1/3}$$

14

The Lagrangian

Let $f : \mathbb{R}^n \rightarrow \mathbb{R}$ be the objective function,

and $g_i : \mathbb{R}^n \rightarrow \mathbb{R}$ for $i = 1, \dots, m$ be constraint functions.

The Lagrangian is the function $\mathcal{L} : \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}$ defined by

$$\mathcal{L}(x, \lambda) = f(x_1, \dots, x_n) + \sum_{i=1}^m \lambda_i g_i(x_1, \dots, x_n).$$

where

$x = (x_1, x_2, \dots, x_n) \in \mathbb{R}^n$ are the primal variables,

$\lambda = (\lambda_1, \dots, \lambda_m) \in \mathbb{R}^m$ are the Lagrange multipliers (dual variables).

Note: the primal variables are the original variables of the optimization problem

(e.g. while the Lagrange multipliers are auxiliary variables introduced to enforce the constraints).

15

Lagrangian: Equality Constraint

If the problem is

maximize/minimize $f(x)$ subject to $g(x) = 0$,

the Lagrangian is

$$\mathcal{L}(x, \lambda) = f(x) + \lambda g(x)$$

for example: $x = (x_1, x_2) = (x, y) \implies \mathcal{L}(x, y, \lambda) = f(x, y) + \lambda g(x, y)$.

16

First-order (Lagrange multiplier) condition

Let $x = (x_1, \dots, x_n) \in \mathbb{R}^n$, $\lambda = (\lambda_1, \dots, \lambda_m) \in \mathbb{R}^m$.

A point (x^*, λ^*) is a stationary point of the Lagrangian if

$$\nabla_x \mathcal{L}(x^*, \lambda^*) = 0, \quad \nabla_\lambda \mathcal{L}(x^*, \lambda^*) = 0,$$

which explicitly means

$$\nabla f(x^*) + \sum_{i=1}^m \lambda_i^* \nabla g_i(x^*) = 0, \quad g_i(x^*) = 0 \quad \text{for } i = 1, \dots, m$$

for example: $x = (x_1, x_2) = (x, y) \implies \nabla f(x^*, y^*) + \lambda^* \nabla g(x^*, y^*) = 0, \quad g(x^*, y^*) = 0$.

At an optimum under constraints, the gradient of f lies in the span of the gradients of the constraints
i.e. level sets are tangent.

17

First-order (Lagrange multiplier) condition

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for example: $x = (x_1, x_2) = (x, y) \implies \nabla f(x^*, y^*) + \lambda^* \nabla g(x^*, y^*) = 0, \quad g(x^*, y^*) = 0$.

Note:

$\mathcal{L}(x, y, \lambda) = f(x, y) + \lambda g(x, y)$ with stationary points $\nabla f(x^*) + \lambda^* \nabla g(x^*) = 0$

is equivalent to

$\mathcal{L}(x, y, \lambda) = f(x, y) - \lambda g(x, y)$ with stationary points $\nabla f(x^*) - \lambda^* \nabla g(x^*) = 0$.

18

The Lagrangian

example

$$\text{maximize } f(x, y) = x^2 y$$

subject to the constraint $g(x, y) = x^2 + y^2 = 1$

$$\text{Lagrangian: } \mathcal{L}(x, y, \lambda) = f(x, y) - \lambda[g(x, y) - 1] \implies \nabla \mathcal{L} = \begin{bmatrix} \frac{\partial \mathcal{L}}{\partial x} \\ \frac{\partial \mathcal{L}}{\partial y} \\ \frac{\partial \mathcal{L}}{\partial \lambda} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$$\frac{\partial \mathcal{L}}{\partial x} = \frac{f(x, y) - \lambda[g(x, y) - 1]}{\partial x} = 0$$

$$\frac{\partial \mathcal{L}}{\partial y} = \frac{f(x, y) - \lambda[g(x, y) - 1]}{\partial y} = 0$$

generally: $g(x, y) = c$

$$\frac{\partial \mathcal{L}}{\partial \lambda} = \frac{f(x, y) - \lambda[g(x, y) - 1]}{\partial \lambda} = 0 - [g(x, y) - 1] = 0 \implies g(x, y) = x^2 + y^2 = 1$$

The Lagrangian

$$\max_{x, y} f(x, y) \quad \text{subject to} \quad g(x, y, c) = 0,$$

where: c is now a parameter (not a choice variable) and x, y are the primal variables.

Define the Lagrangian $\mathcal{L}(x, y, \lambda; c) = f(x, y) + \lambda g(x, y, c)$.

At an optimum (x^*, y^*, λ^*) , the first-order conditions are $\nabla_{x,y} \mathcal{L}(x^*, y^*, \lambda^*; c) = 0, \quad g(x^*, y^*, c) = 0$.

Define the optimal value: $M^* = f(x^*, y^*)$.

Since x^* and y^* depend on c we can write this explicitly as

$$M^*(c) = f(x^*(c), y^*(c)).$$

The envelope theorem states that $\lambda^* = \frac{dM^*}{dc} = \lambda^* \frac{\partial g(x^*, y^*, c)}{\partial c}$

The Lagrange multiplier equals the derivative of the value function with respect to the constraint parameter.

19

20

The Lagrangian

exercise

Suppose you run a small workshop. Labor costs €10 per hour and steel costs €40 per ton.

Revenue is modeled as $R(h, s) = 100 h^{1/2} s^{1/2}$, where h is labor hours and s is tons of steel.

If your budget is €400, what is the maximum possible revenue? Is the constraint valuable to relax?