

Lecture 24

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Recap from last lecture

1. Unconstrained optimization

$$\text{Min } f(x)$$

where

- $x = (x_1, x_2, \dots, x_n)$
- $f(x)$ is continuous and differentiable

1. Unconstrained optimization

Necessary condition of optimality

If \bar{x} is a global minimizer of f , then

$$\nabla f(\bar{x}) = 0.$$

1. Unconstrained optimization

Necessary and sufficient condition of optimality

Suppose f is continuous, differentiable, and **convex**.

\bar{x} is a global minimizer of f if and only if

$$\nabla f(\bar{x}) = 0.$$

Convex functions

Definition (convex function)

A function f is **convex** if:

for any two points a and b , and any number λ with $0 \leq \lambda \leq 1$,

the following inequality is satisfied:

$$f(\lambda a + (1 - \lambda)b) \leq \lambda f(a) + (1 - \lambda)f(b).$$

i>clicker

Q: Which functions are convex?

A. f_1 only

B. f_1 and f_2

C. f_1, f_2 , and f_3

D. f_1, f_2, f_3 , and f_4

E. f_1, f_2, f_3, f_4 , and f_5

Q: Which functions are convex?

A. f_1 only

B. f_1 and f_2

C. f_1, f_2 , and f_3

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E. f_1, f_2, f_3, f_4 , and f_5

Lecture 24

2. Constrained Optimization

$$\begin{array}{ll} \min & f(x) \\ s.t. & g_i(x) \geq 0 \quad \forall i \in \{1, \dots, p\} \quad (NLP) \\ & h_j(x) = 0 \quad \forall j \in \{1, \dots, q\} \end{array}$$

Example:

Nonlinear constrained optimization

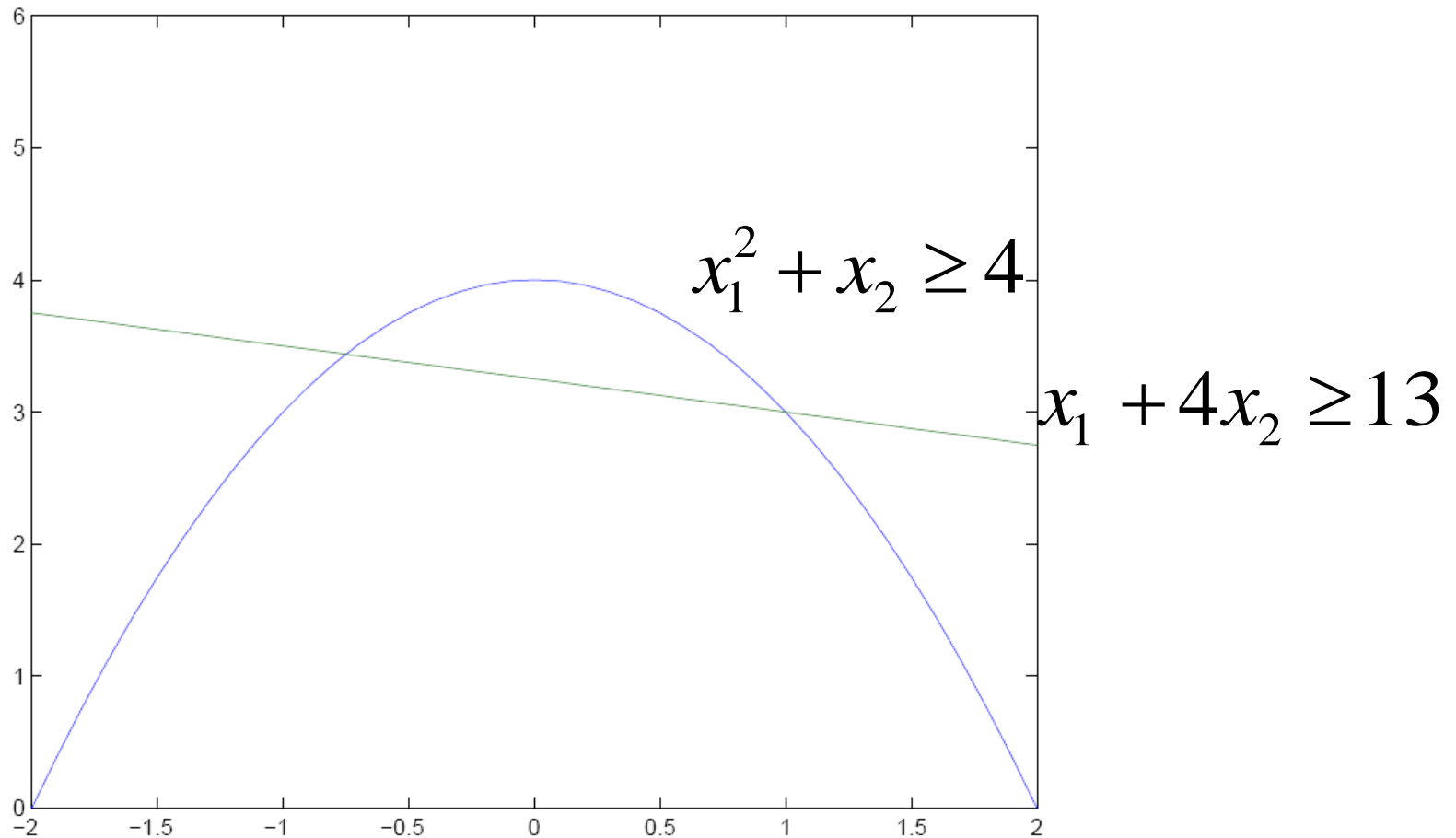
$$\min \quad x_1^2 + x_2^2$$

$$s.t. \quad x_1^2 + x_2 \geq 4$$

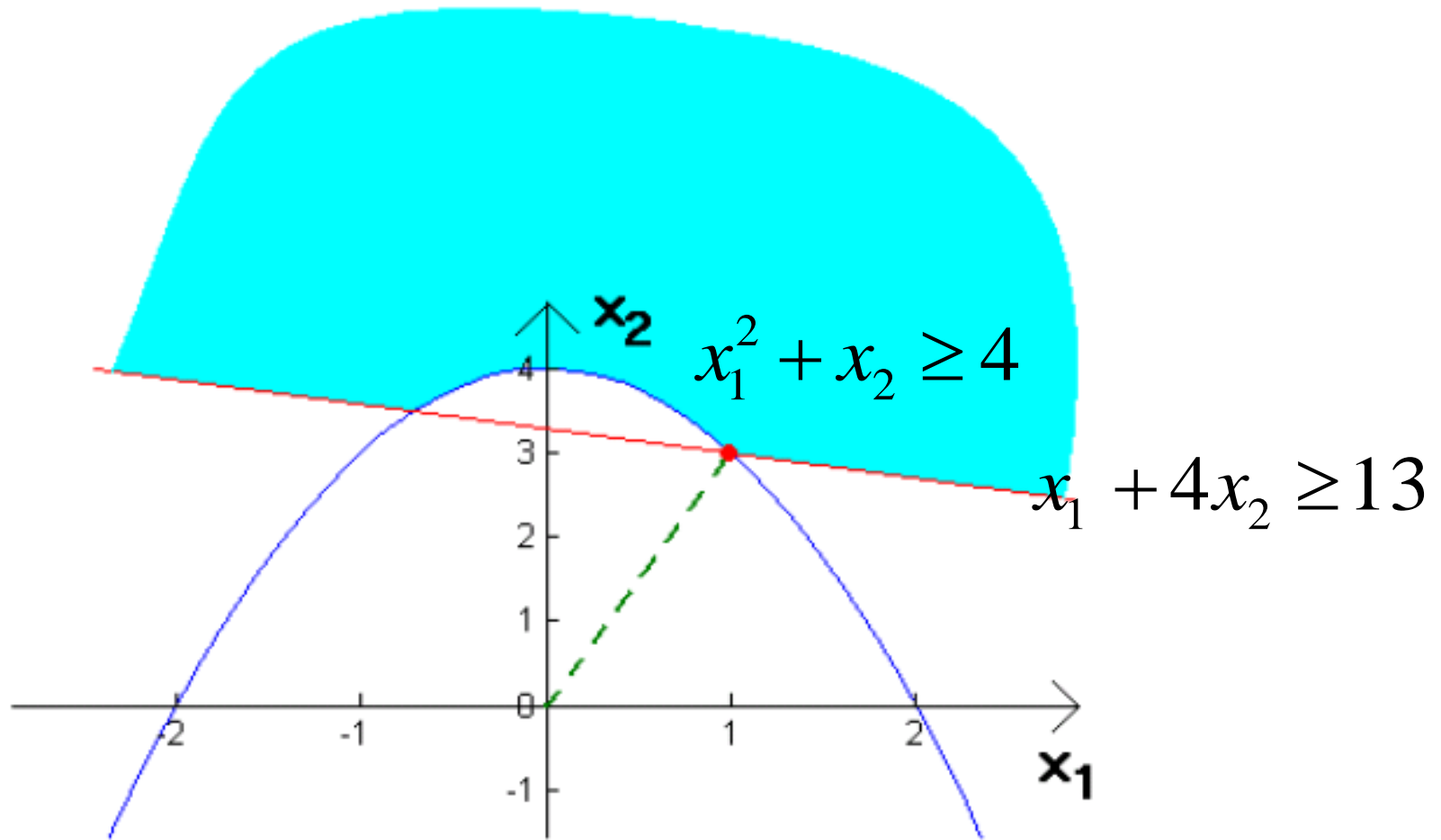
$$x_1 + 4x_2 \geq 13$$

Example:

Nonlinear constrained optimization

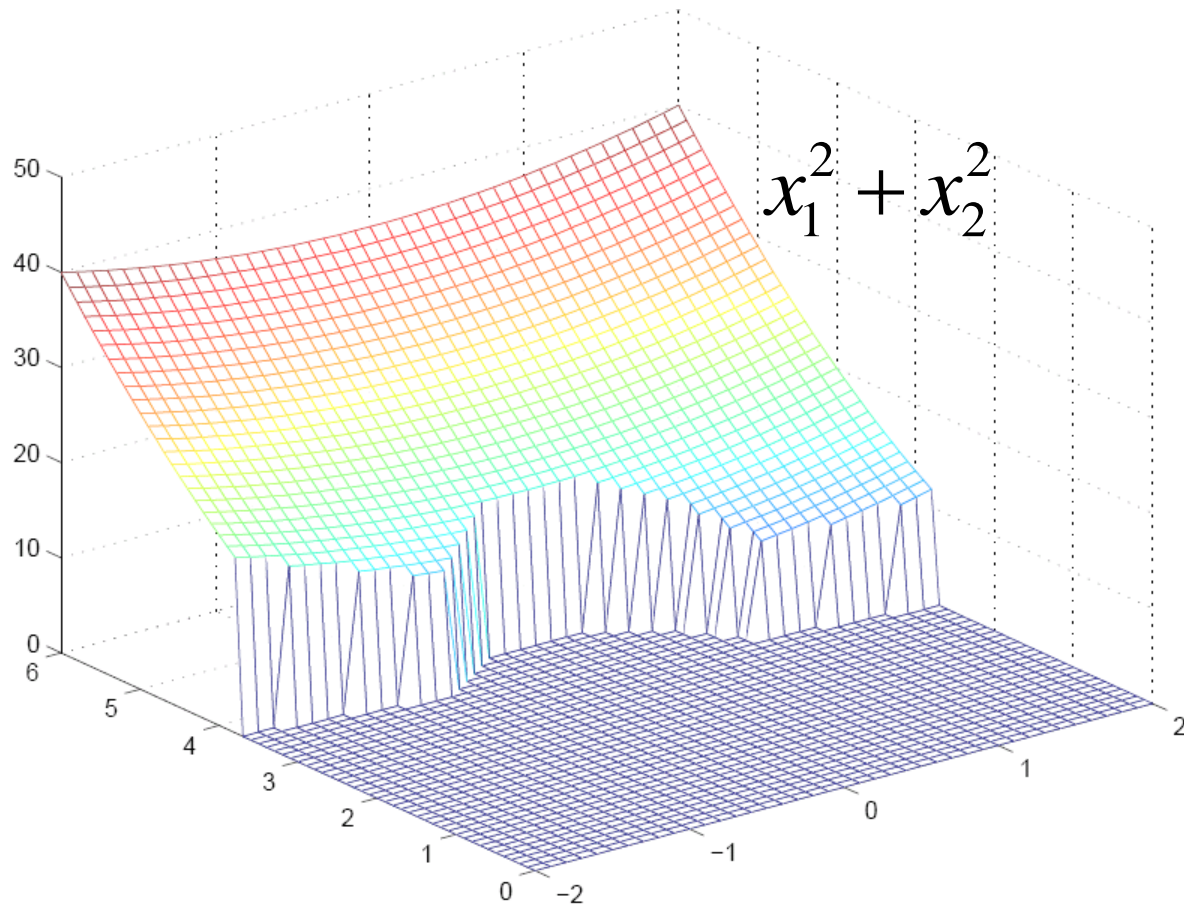


Example: Nonlinear constrained optimization



Example:

Nonlinear constrained optimization



Example:

Nonlinear constrained optimization

$$\min \quad x_1^2 + x_2^2$$

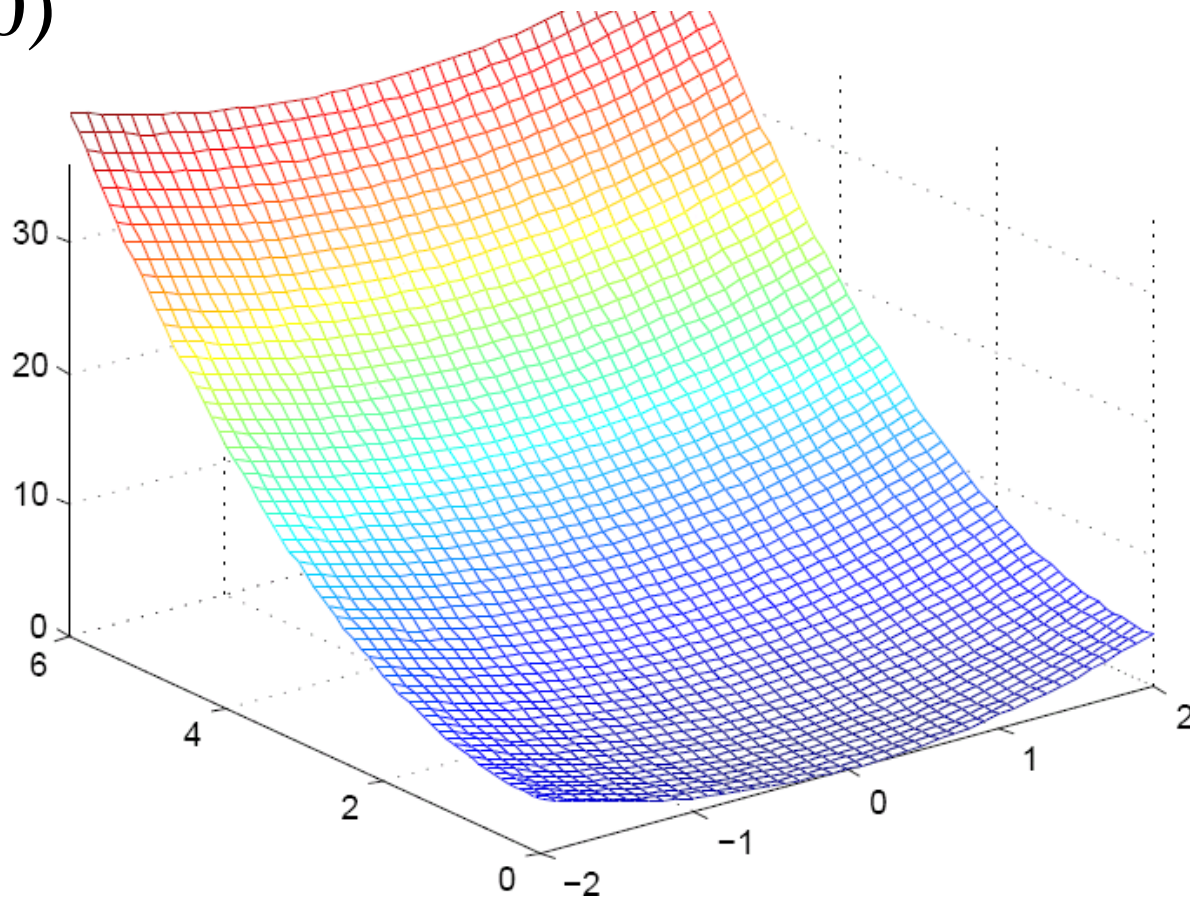
$$s.t. \quad x_1^2 + x_2 \geq 4$$

$$x_1 + 4x_2 \geq 13$$

Example:

$$L_y(x) = f(x) - y_1 g_1(x) - y_2 g_2(x)$$

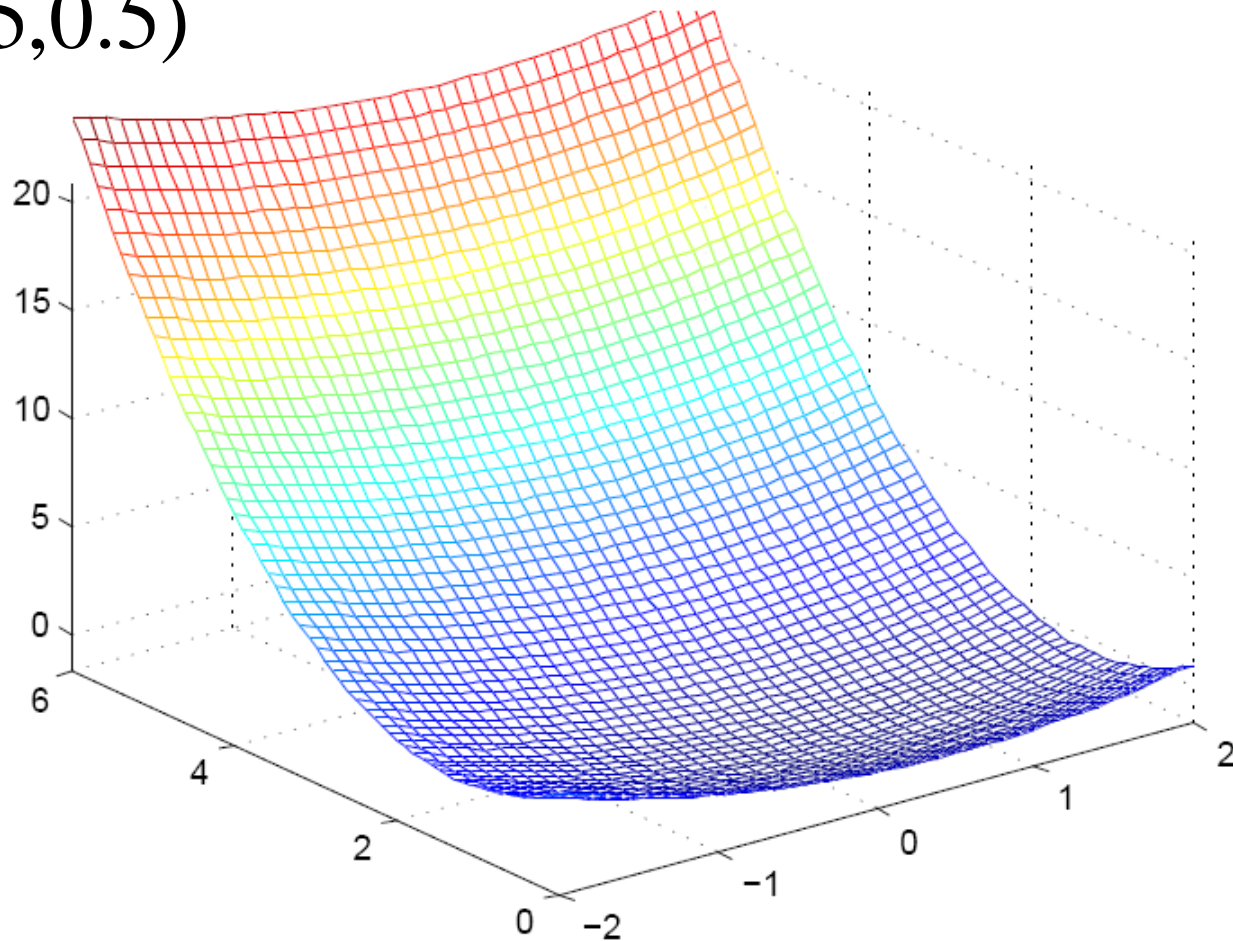
$$y = (0,0)$$



Example:

$$L_y(x) = f(x) - y_1 g_1(x) - y_2 g_2(x)$$

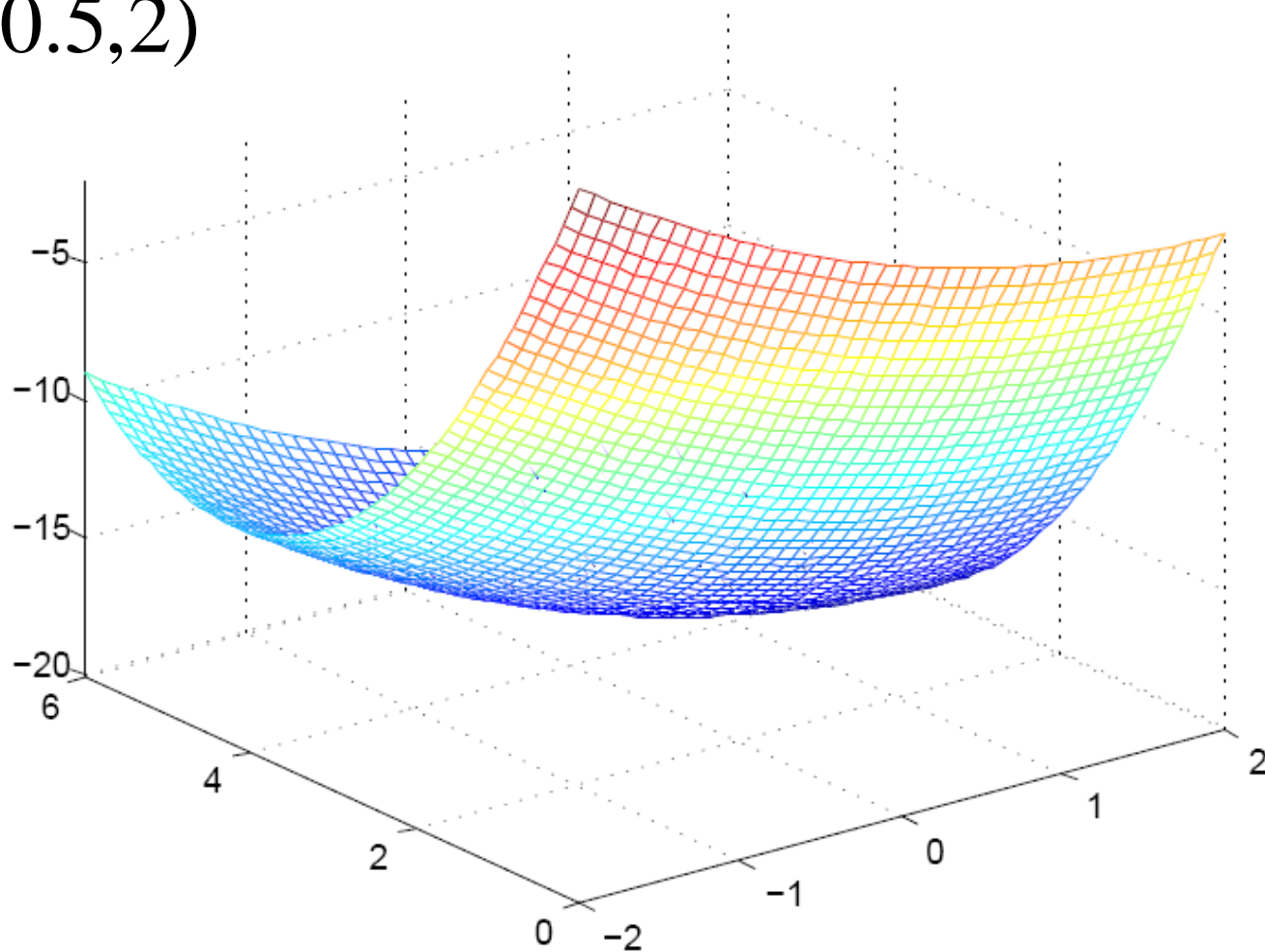
$$y = (0.5, 0.5)$$



Example:

$$L_y(x) = f(x) - y_1 g_1(x) - y_2 g_2(x)$$

$$y = (0.5, 2)$$



1. Unconstrained optimization

Necessary condition of optimality

If \bar{x} is a global minimizer of f , then

$$\nabla f(\bar{x}) = 0.$$

2. Constrained Optimization

$$\begin{array}{ll} \min & f(x) \\ s.t. & g_i(x) \geq 0 \quad \forall i \in \{1, \dots, p\} \quad (NLP) \\ & h_j(x) = 0 \quad \forall j \in \{1, \dots, q\} \end{array}$$

i>clicker

Q: Which solution is optimal for our NLP example?

A. $x^A = (2, 0)$

B. $x^B = (10, 20.25)$

C. $x^C = (1, 3)$

D. $x^B = (10, 20.25)$ and $x^C = (1, 3)$

E. None of the above

Q: Which solution is optimal for our NLP example?

A. $x = (2, 0)$

B. $x = (10, 20.25)$

C. $x = (1, 3)$

D. $x = (10, 20.25)$ and $x = (1, 3)$

E. None of the above

AMPL