

# MATH529 – Fundamentals of Optimization

## Constrained Optimization III

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## Constrained Optimization

A general model of a constrained optimization problem is:

$$\min_{x \in \mathbb{R}^n} f(x)$$

subject to

$$c_i(x) = 0, \quad i \in \mathcal{E}$$

$$c_i(x) \geq 0, \quad i \in \mathcal{I}$$

where  $f$  is called the *objective function*, the functions  $c_i(x)$ ,  $i \in \mathcal{E}$  are the *equality constraints*, and the functions  $c_i(x)$ ,  $i \in \mathcal{I}$  are the *inequality constraints*.

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## Example: The diet problem

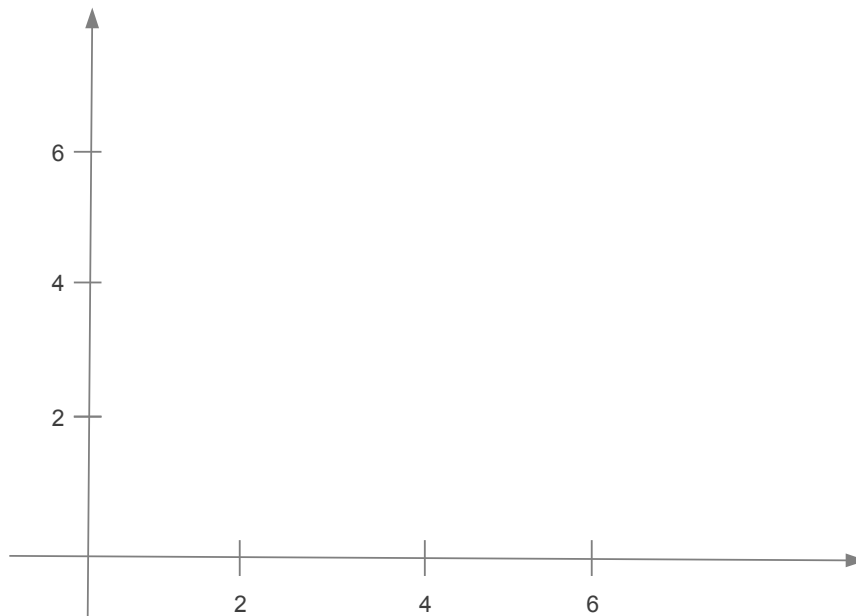
Food 1: \$0.6 cts per 100 grms.

Food 2: \$1 cts per 100 grms.

Nutrient	Food 1	Food 2	Minimum Daily Requirement
Calcium	10	4	20
Protein	5	5	20
Vitamins	2	6	12

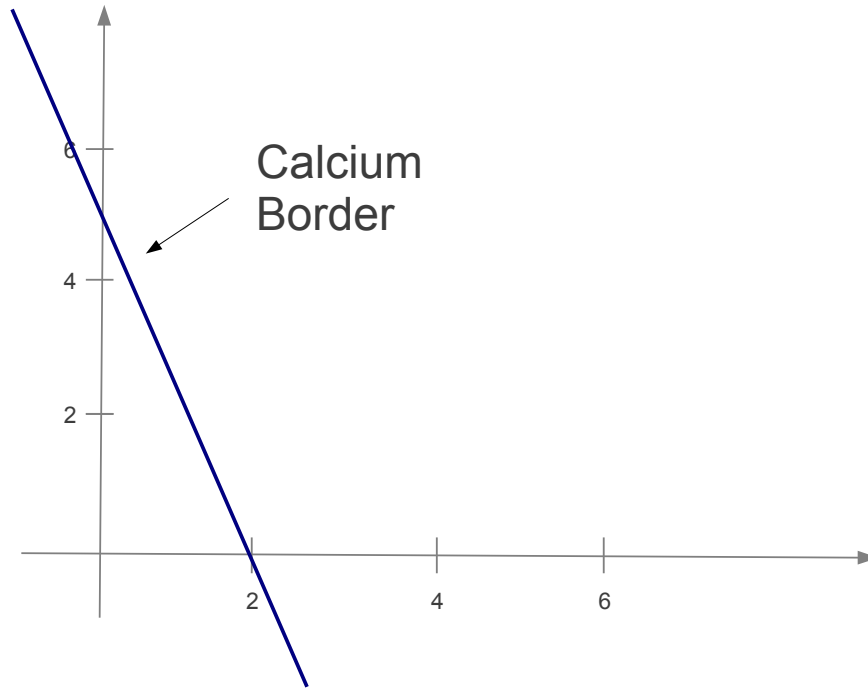
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## Graphical Solution



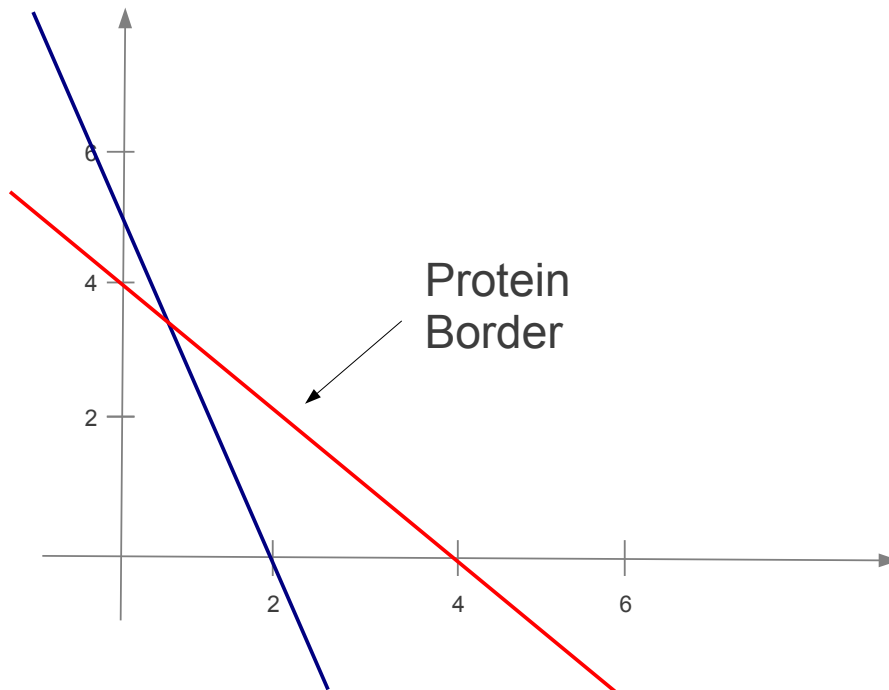
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## Graphical Solution



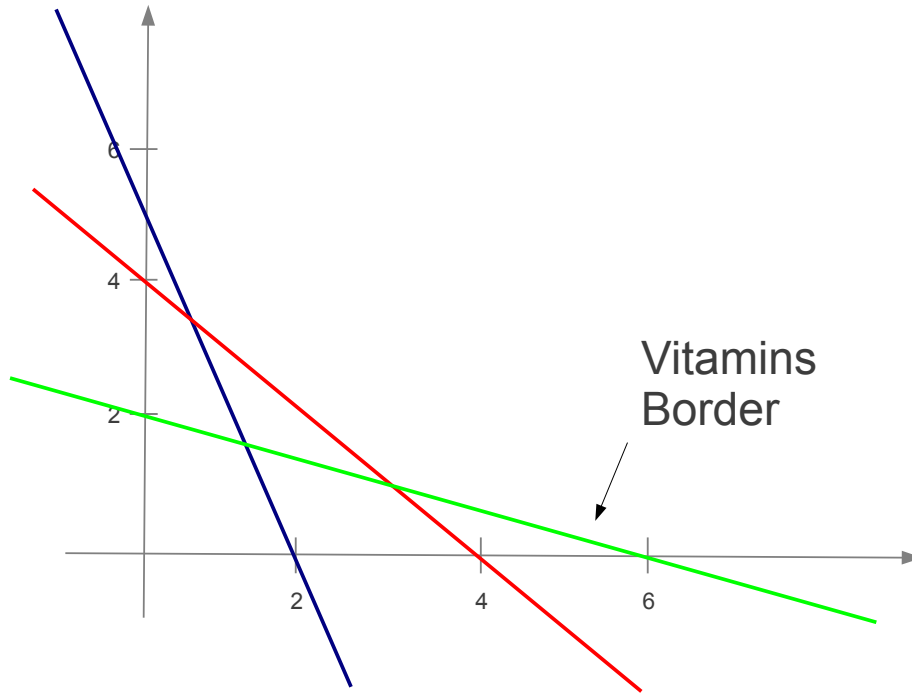
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## Graphical Solution



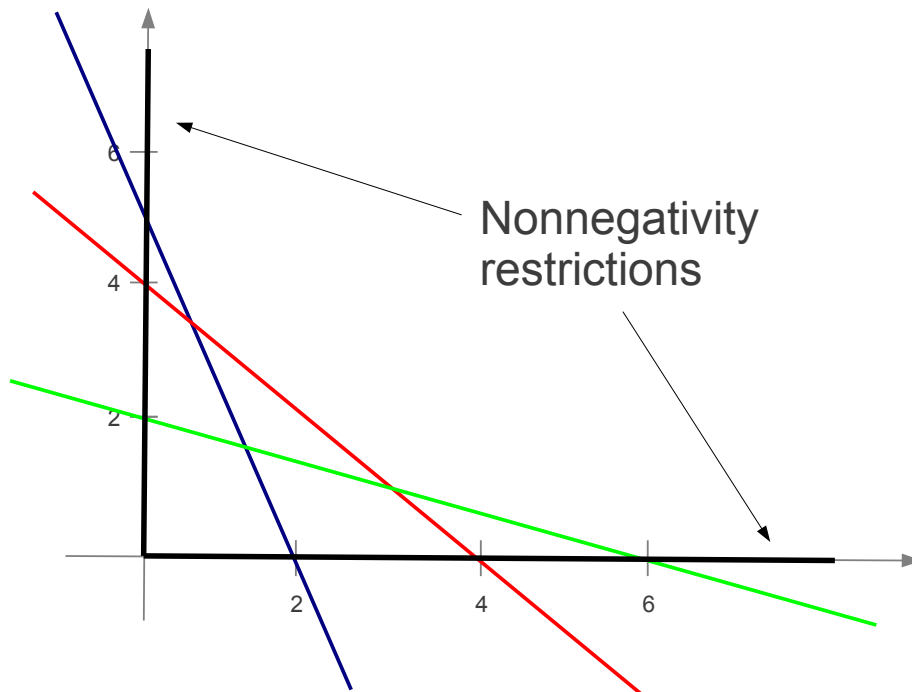
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## Graphical Solution



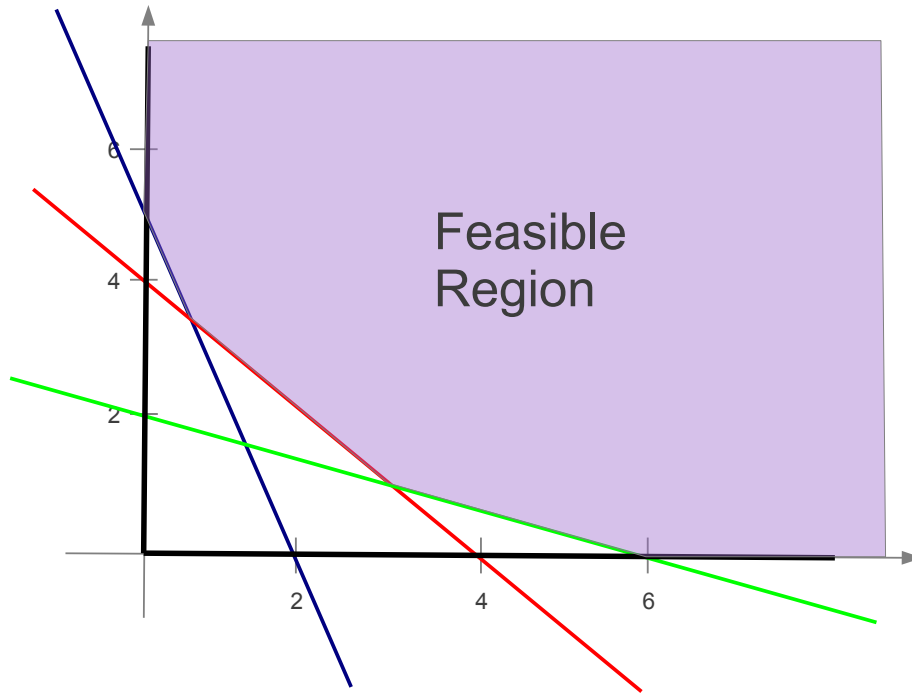
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## Graphical Solution



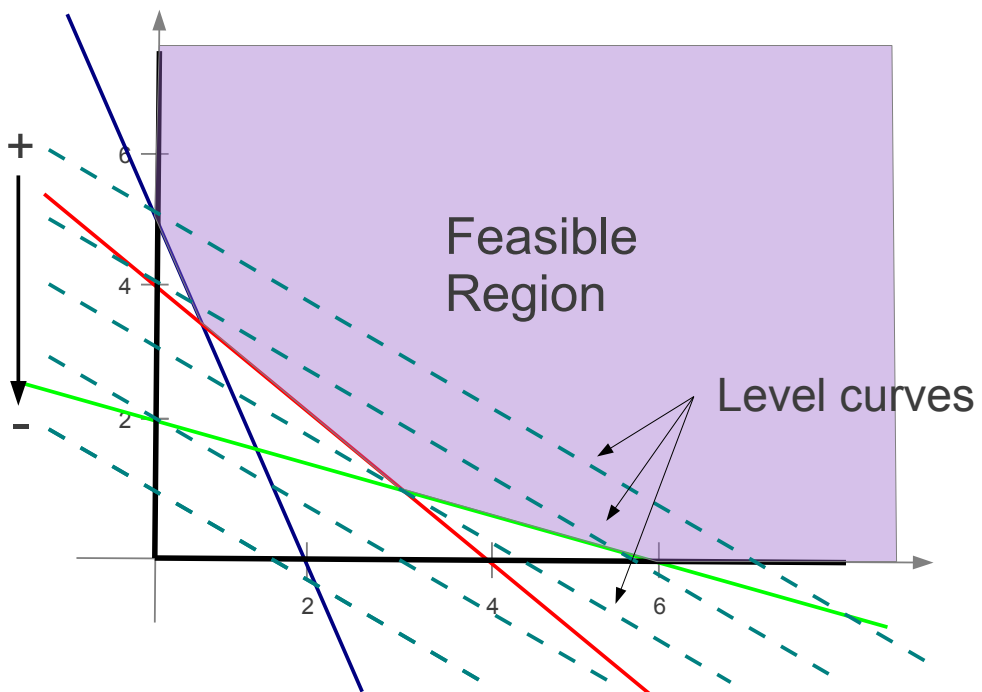
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## Graphical Solution



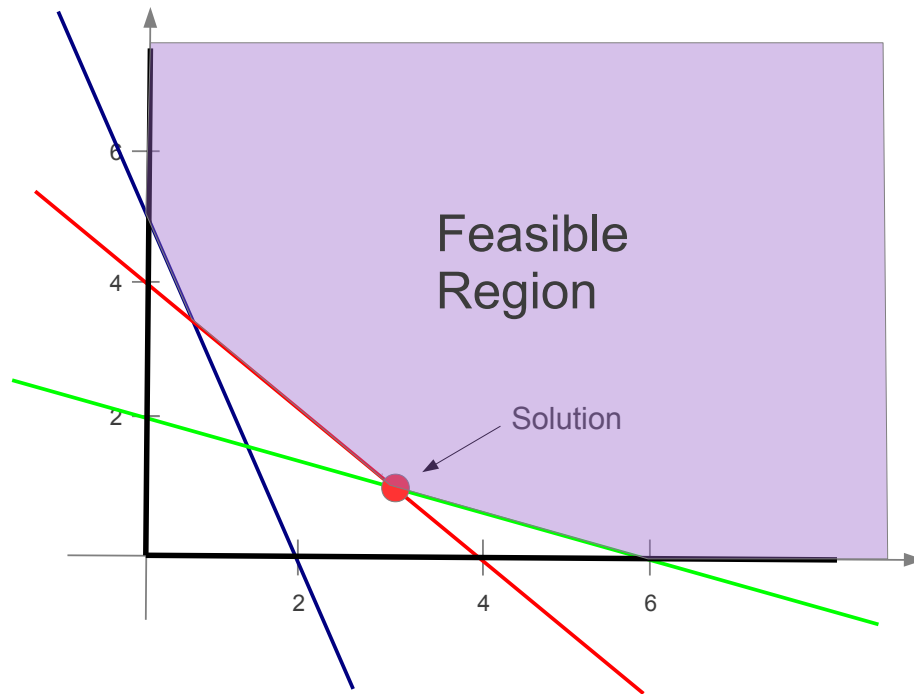
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## Graphical Solution



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## Graphical Solution



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## Equality vs. Inequality Constraints

### Equality

minimize  $x_1 + x_2$

$$10x_1 + 4x_2 = 20$$

$$5x_1 + 5x_2 = 20$$

$$2x_1 + 12x_2 = 12$$

### Inequality

minimize  $x_1 + x_2$

$$10x_1 + 4x_2 \geq 20$$

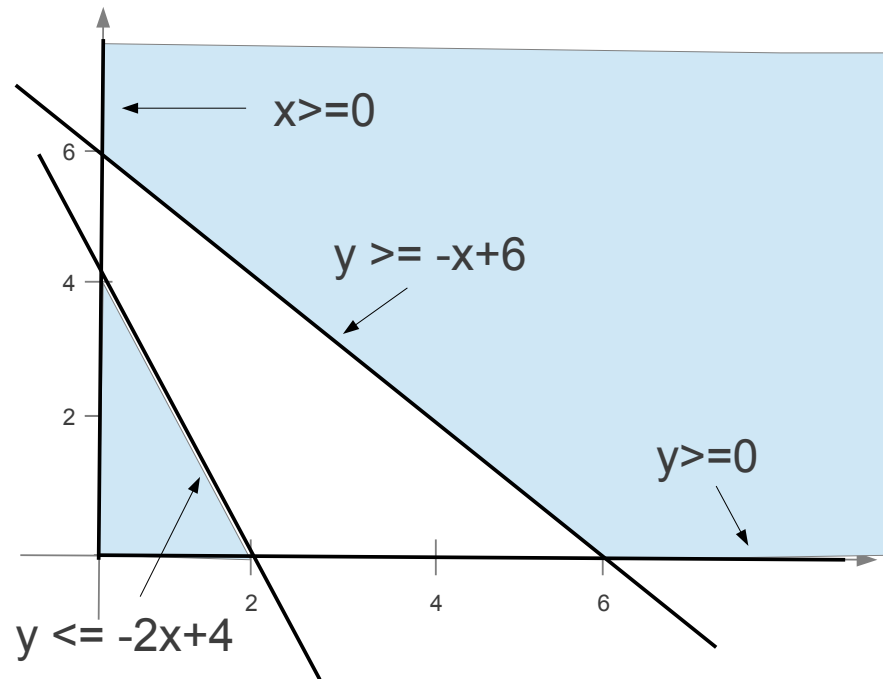
$$5x_1 + 5x_2 \geq 20$$

$$2x_1 + 12x_2 \geq 12$$

$$x_1, x_2 \geq 0$$

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## What can go wrong?



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## Active Set and Constraints

### Definition

An *active* inequality constraint  $i$  at a point  $x$  is one for which  $c_i(x) = 0$ . An *inactive* inequality constraint  $j$  at  $x$  is one for which  $c_j(x) > 0$ .

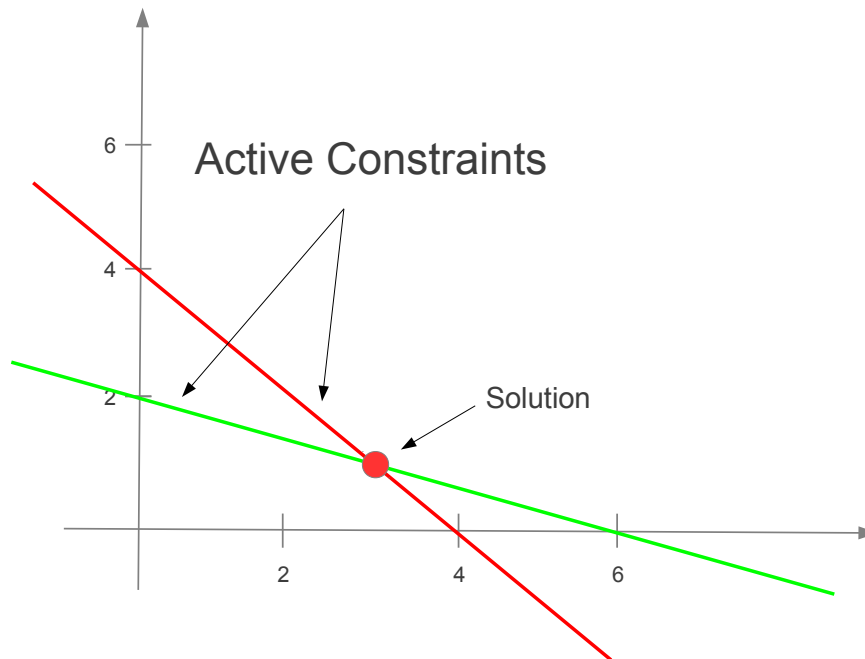
### Definition

The active set  $\mathcal{A}(x)$  at any feasible point  $x$  is the set of indices of all the equality constraints together with the indices  $i$  of the active inequality constraints.

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## In our example

At the solution  $x = (3, 1)^T$ ,  $\mathcal{A}(x) = \{2, 3\}$ .



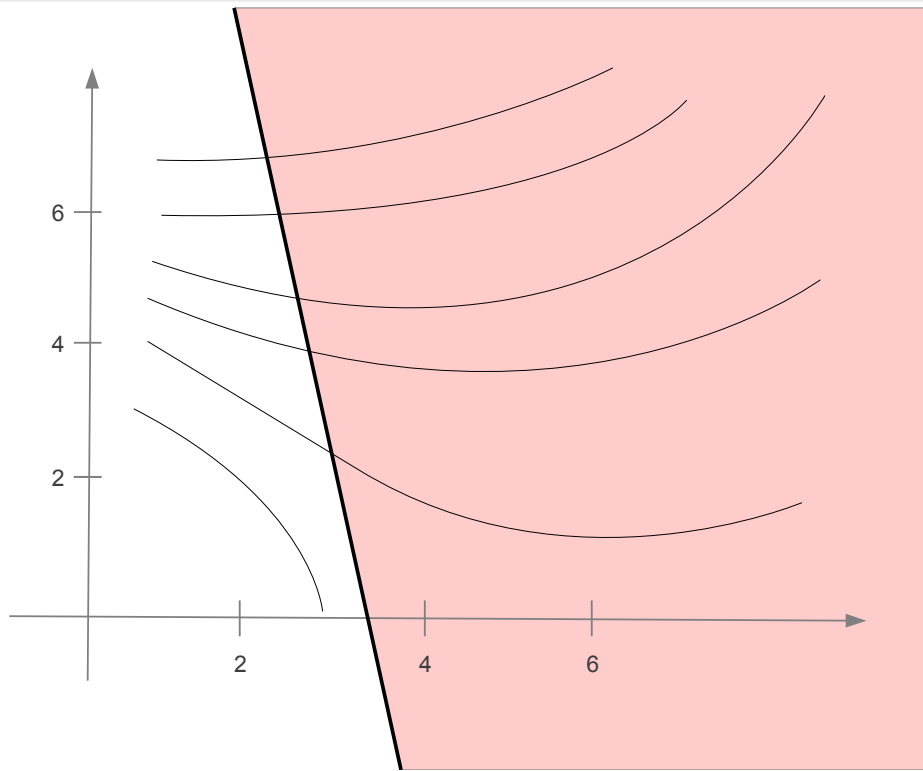
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## Karush-Kuhn-Tucker (KKT) Conditions

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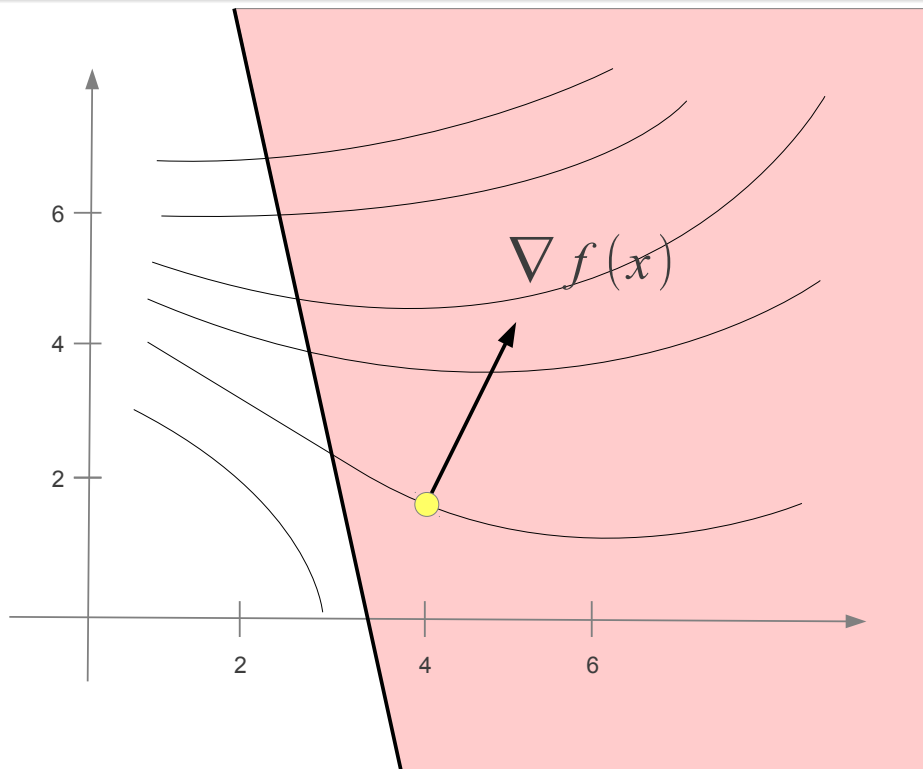


## Decrease Conditions: Interior Point



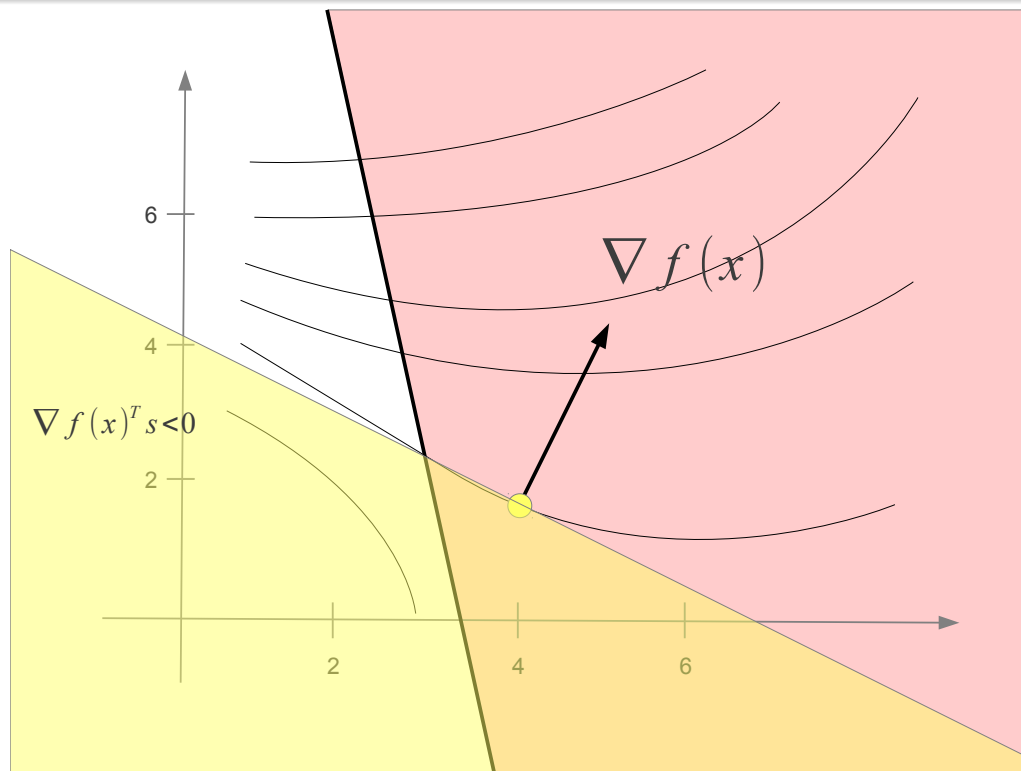
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## Decrease Conditions: Interior Point



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## Decrease Conditions: Interior Point



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## Decrease Conditions: Interior Points

At a point  $x$ , a step  $s$  that would result in an objective function should satisfy:

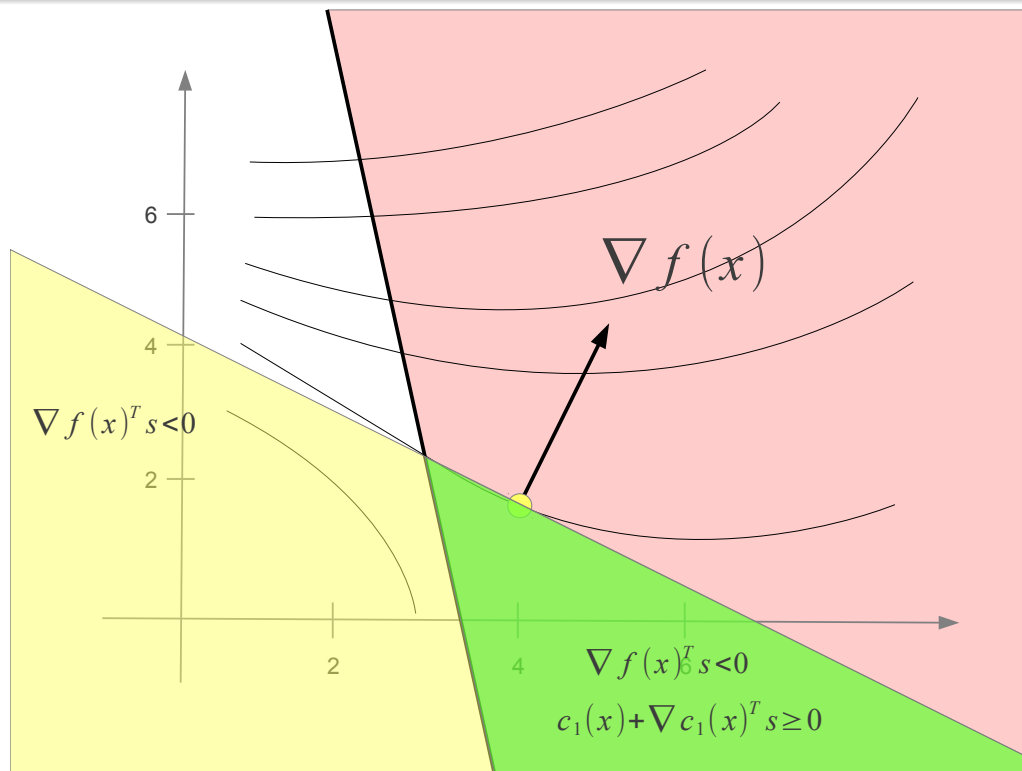
$$\nabla f(x)^T s < 0 \text{ (objective function decrease)*}$$

$$0 \leq c(x + s) \approx c(x) + \nabla c(x)^T s \text{ (feasibility)}$$

\* Unless  $\nabla f(x) = 0$

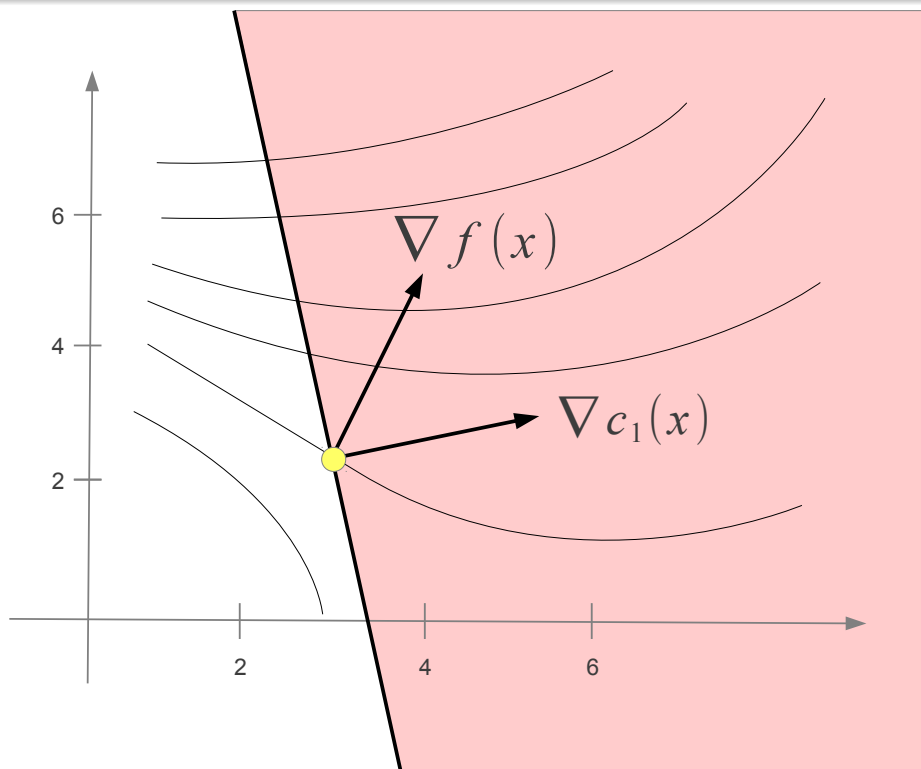
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## Decrease Conditions: Border Point



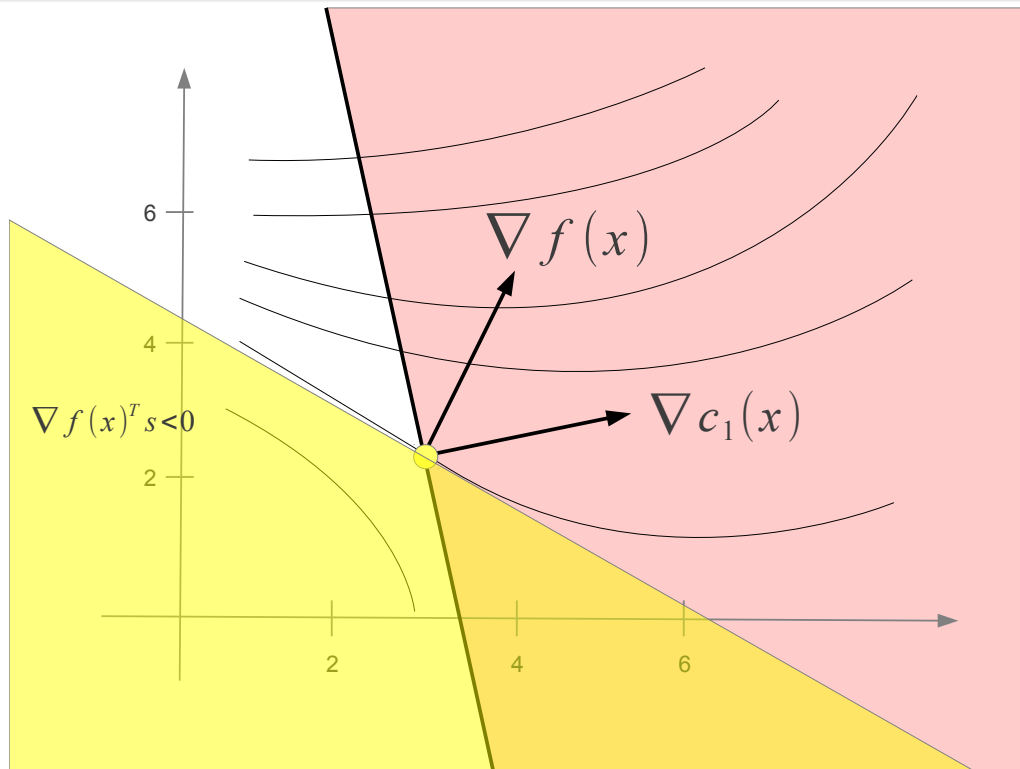
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## Decrease Conditions: Border Point



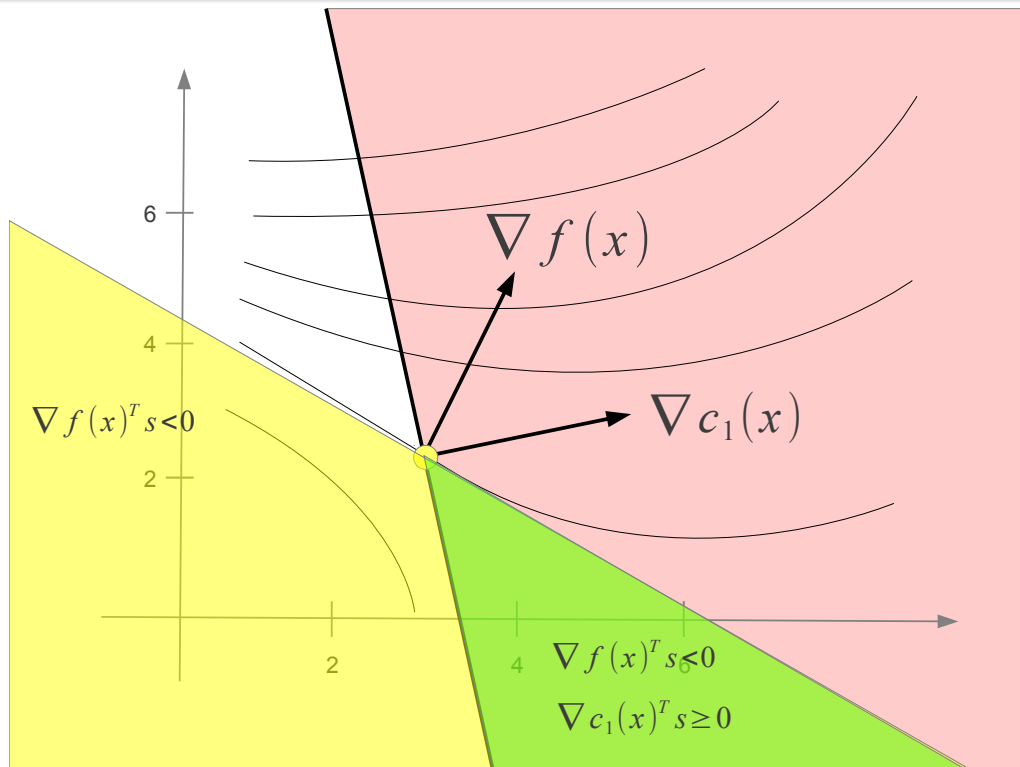
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## Decrease Conditions: Border Point



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## Conditions: Border Point



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## Decrease Conditions: Border Points

At a point  $x$ , a step  $s$  that would result in an objective function should satisfy:

$$\nabla f(x)^T s < 0 \text{ (objective function decrease)}$$

$$0 \leq c(x + s) \approx c(x) + \nabla c(x)^T s \text{ (feasibility)}$$

However, since  $x$  is a border point,  $c(x) = 0$ ,

$$0 \leq \nabla c(x)^T s \text{ (feasibility)}$$

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## Conditions: General

When no step  $s$  can be found, the point  $x^*$  is a candidate for a solution. The conditions for interior and border cases, can be summarized as follows:

$$\nabla_x L(x^*, \lambda^*) = 0,$$

$$c_i(x) = 0, \quad \text{for all } i \in \mathcal{E}$$

$$c_i(x) \geq 0, \quad \text{for all } i \in \mathcal{I}$$

$$\lambda^* c_i(x^*) = 0, \quad \text{for all } i \in \mathcal{E} \cup \mathcal{I} \text{ (complementarity condition), and}$$

$$\lambda^* \geq 0.$$

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## Example

$$\text{minimize } x^2 + 2y^2$$

subject to:

$$x + y \geq 0$$

$$y - x^2 \geq 1$$

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## Example

$$\text{maximize } x + y^2$$

subject to:

$$x - y = 5$$

$$x^2 + 9y^2 \leq 36$$

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