

# Fundamentals of optimization

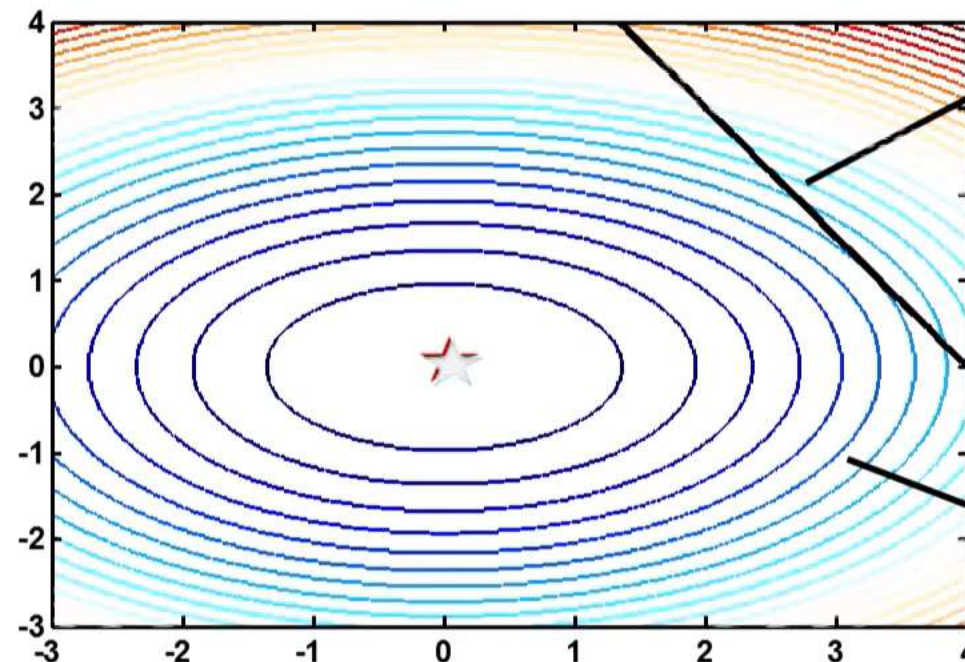
## Multivariate optimization with constraints

$$\min_{x_1, x_2} 2x_1^2 + 4x_2^2$$

*st*

$$3x_1 + 2x_2 \leq 12$$

- ★ Constrained minimum
- ★ Unconstrained minimum



$$3x_1 + 2x_2 = 12$$

Unconstrained minimum is the same as constrained minimum

feasible region

# General formulation

## Multivariate optimization

$$\min_{\bar{x}} f(\bar{x})$$

st

$$h_i(\bar{x}) = \bar{0}, i = 1, \dots, m$$

$$g_j(\bar{x}) \leq \bar{0}, j = 1, 2, \dots, l$$

Necessary condition for  $\bar{x}^*$  to be the minimizer

KKT conditions has to be satisfied

Sufficient condition

$\nabla^2 L(\bar{x}^*)$  has to be positive definite

# Summary – KKT conditions

## Multivariate optimization

When both equality and inequality constraints are present, at the optimum we have

### KKT (Karush-Kuhn-Tucker) conditions

$$\nabla f(\bar{x}^*) + \sum_{i=1}^l [\nabla h_i(\bar{x}^*)] \lambda_i^* + \sum_{j=1}^m [\nabla g_j(\bar{x}^*)] \mu_j^* = 0$$

$$h_i(x^*) = 0, i = 1 \dots l$$

$$\lambda_i \in \mathbb{R}, i = 1 \dots l$$

$$g_j(\bar{x}^*) \leq 0, j = 1 \dots m$$

$$\mu_j^*(g_j(\bar{x}^*)) = 0$$

$$\mu_j^* \geq 0, j = 1 \dots m$$

Gradient of the “Lagrangian function” at  $x^*$

$$L(\bar{x}^*, \lambda^*, \mu^*) = f(\bar{x}^*) + \sum_{i=1}^l \lambda_i h(\bar{x}^*) + \sum_{j=1}^m \mu_j g_j(\bar{x}^*)$$

Ensures that the optimum satisfies equality constraints

Ensures that the optimum is in the feasible region

Complementary slackness

➤ No possibility of improvement near the active constraints



# Summary – KKT conditions

## Multivariate optimization

- In general it is difficult to use the KKT conditions to solve for the optimum of an inequality constrained problem (than for a problem with equality constraints only) because we do not know a priori which constraints are active at the optimum.
- Makes this a combinatorial problem
- KKT conditions are used to verify that a point we have reached is a candidate optimal solution.
- Given a point, it is easy to check which constraints are binding.



# Fundamentals of optimization

## Multivariate optimization-quadratic programming

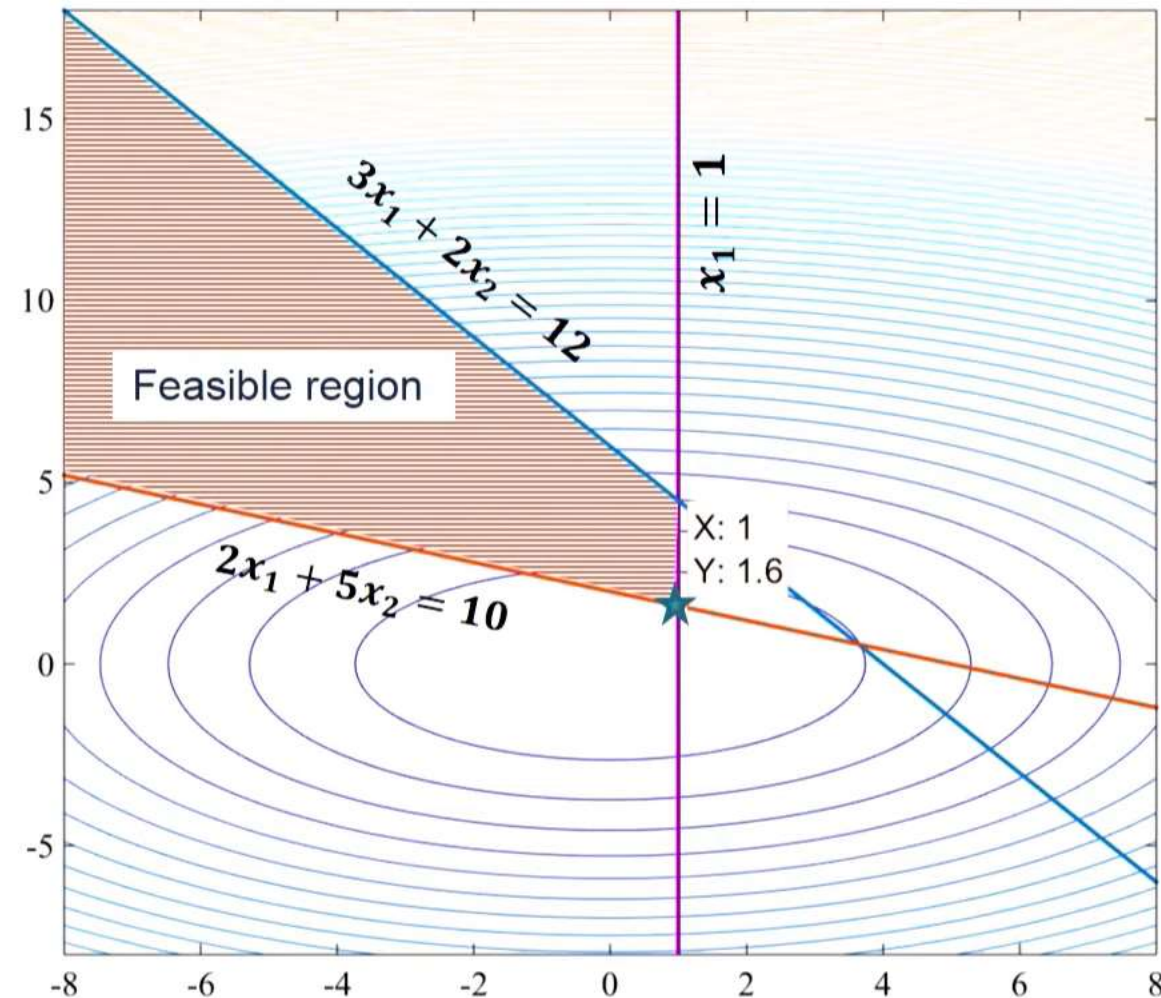
$$\min_{x_1, x_2} 2x_1^2 + 4x_2^2$$

st

$$3x_1 + 2x_2 \leq 12$$

$$2x_1 + 5x_2 \geq 10$$

$$x_1 \leq 1$$



# Fundamentals of optimization

## Multivariate optimization-quadratic programming

$$\begin{array}{ll}
 \min_{x_1, x_2} & 2x_1^2 + 4x_2^2 \\
 \text{st} & \\
 3x_1 + 2x_2 \leq 12 & \Rightarrow (a) \\
 2x_1 + 5x_2 \geq 10 & \Rightarrow (b) \\
 x_1 \leq 1 & \Rightarrow (c)
 \end{array}$$

- Lagrangian

$$\begin{aligned}
 L(x_1, x_2, \mu_1, \mu_2, \mu_3) = & 2x_1^2 + 4x_2^2 + \mu_1(3x_1 + 2x_2 - 12) \\
 & + \mu_2(10 - 2x_1 - 5x_2) + \mu_3(x_1 - 1)
 \end{aligned}$$

- First order KKT conditions

$$4x_1 + 3\mu_1 - 2\mu_2 + \mu_3 = 0$$

$$8x_2 + 2\mu_1 - 5\mu_2 = 0$$

$$\mu_1(3x_1 + 2x_2 - 12) = 0$$

$$\mu_2(10 - 2x_1 - 5x_2) = 0$$

$$\mu_3(x_1 - 1) = 0$$

$$\mu_i \geq 0$$



# Fundamentals of optimization

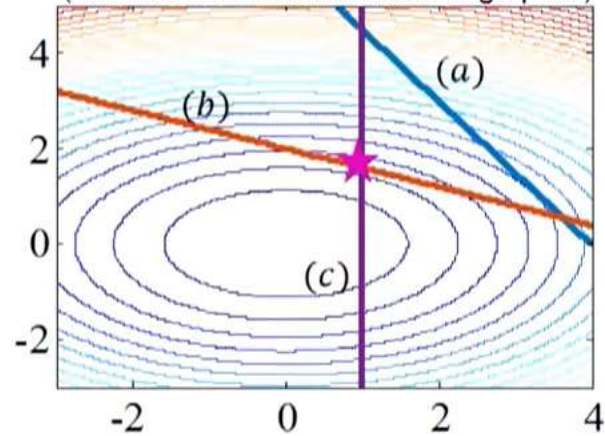
## Multivariate optimization-quadratic programming

Sl.no	Active (A) /Inactive (I) constraints			Solution ( $x, \mu$ )	Possible optima (Y/N)	Remark
(a)	(b)	(c)				
1	A	A	A	Infeasible	N	Equations do not have a valid solution.
2	A	A	I	$x = [3.6364 \quad 0.5455]$ $\mu = [-5.2 \quad -1.45 \quad 0]$	N	$x_1 \leq 1$ is not satisfied, $\mu_1 < 0$ , $\mu_2 < 0$
3	A	I	A	$x = [1 \quad 4.5]$ $\mu = [-18 \quad 0 \quad 50]$	N	$\mu_1 < 0$
4	I	A	A	$x = [1 \quad 1.6]$ $\mu = [0 \quad 2.56 \quad 1.12]$	Y	All constraints and KKT conditions satisfied
5	A	I	I	$x = [3.27 \quad 1.09]$ $\mu = [-4.36 \quad 0 \quad 0]$	N	$x_1 \leq 1$ is not satisfied
6	I	A	I	$x = [1.21 \quad 1.51]$ $\mu = [0 \quad 2.45 \quad 0]$	N	$x_1 \leq 1$ is not satisfied
7	I	I	A	$x = [1 \quad 0]$ $\mu = [0 \quad 0 \quad -4]$	N	$2x_1 + 5x_2 \geq 10$ is not satisfied
8	I	I	I	$x = [0 \quad 0]$ $\mu = [0 \quad 0 \quad 0]$	N	$2x_1 + 5x_2 \geq 10$ is not satisfied

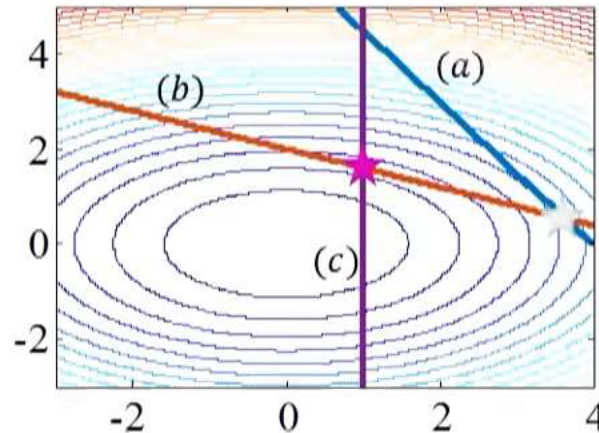
# Fundamentals of optimization

★ Solution for each case  
★ Actual Optima

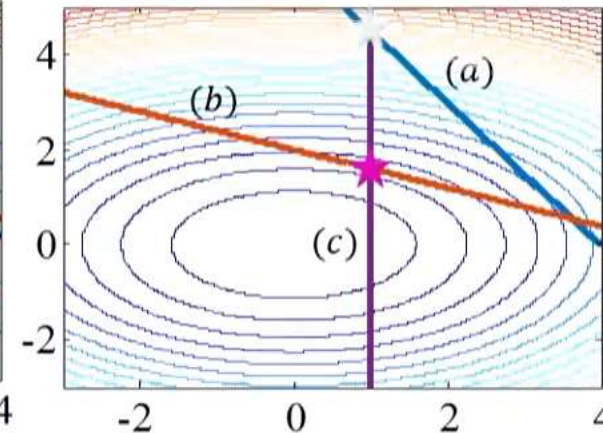
**Case 1: No solution**  
(All 3 lines do not intersect at a single point)



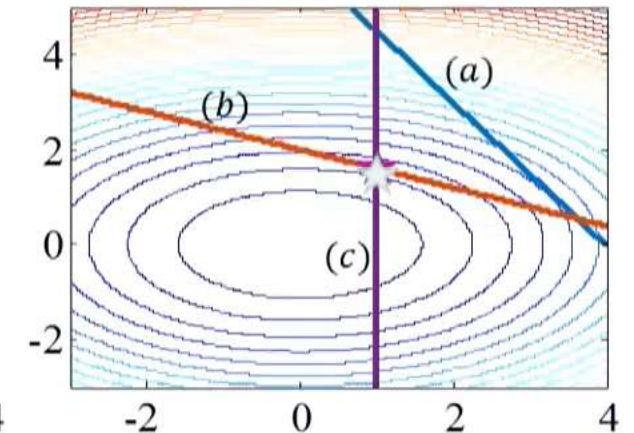
**Case 2**



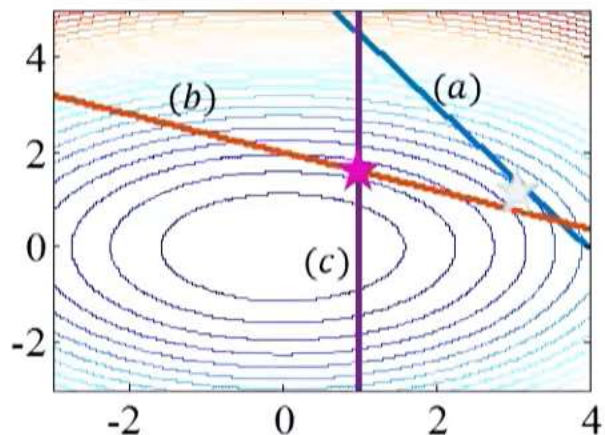
**Case 3**



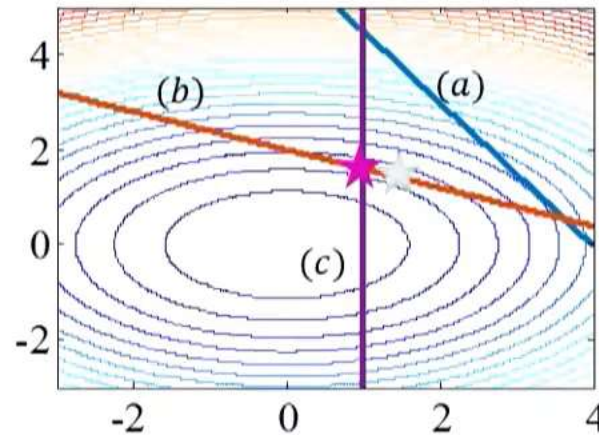
**Case 4**



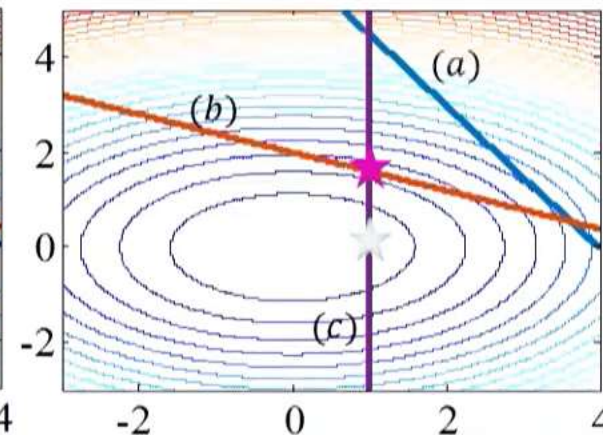
**Case 5**



**Case 6**



**Case 7**



**Case 8**

