

# Constraint Optimization Problems

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# Example 1

A farm uses at least 800 kg of special feed daily. The special feed is a mixture of corn and soybean meal with the following compositions.

Feedstuff	kg per kg of feedstuff		Cost (\$ per kg)
	Protein	Fiber	
Corn	0.09	0.02	0.3
Soyabean meal	0.6	0.06	0.9

The dietary requirements of the special feed are at least 30% protein and at most 5% fiber. The goal is to determine the daily minimum-cost of feed mix.

# Example 1

Let  $x_1$  = kg of corn in the daily mix

$x_2$  = kg of soybean meal in the daily mix

**Objective: determine minimum-cost feed mix.**

Minimize  $z = 0.3 x_1 + 0.9 x_2$

Subject to,

$$x_1 + x_2 \geq 800 \quad \text{Daily demand}$$

$$0.09 x_1 + 0.6 x_2 \geq 0.3(x_1 + x_2) \quad \text{Protein requirement in total mix}$$

$$-0.21x_1 + 0.3 x_2 \geq 0$$

$$0.02 x_1 + 0.06 x_2 \leq 0.05(x_1 + x_2) \quad \text{Fiber requirement in total mix}$$

$$-0.03 x_1 + 0.01 x_2 \leq 0$$

$$x_1, x_2 \geq 0 \quad \text{Lower bounds of the decision variables}$$

Feedstuff	kg per kg of feedstuff		Cost (\$ per kg)
	Protein	fiber	
Corn	0.09	0.02	0.3
Soyabean meal	0.6	0.06	0.9
Dietary requirements	At least 30%	At most 5%	

farm uses at least 800 lb  
of special feed daily

# Example 2

- Reddy Mikks company produces interior and exterior paints from raw materials, M1 and M2.
- Daily demand for interior paint cannot exceed that for exterior paint by more than 1 unit.
- Maximum daily demand for the interior paint is 2 units.
- Determine optimum quantity of interior and exterior paints that maximizes total daily profit.

	Exterior paint	Interior paint
M1	6	4
M2	1	2

# Example 2

Let  $x_1$  = Units of exterior paint produced daily

$x_2$  = Units of interior paints produced daily

Maximize Profit,  $Z = 5x_1 + 4x_2$

Subject to

$$6x_1 + 4x_2 \leq 24$$

$$x_1 + 2x_2 \leq 6$$

$$x_2 \leq x_1 + 1$$

$$x_1, x_2 \geq 0$$

$$x_2 \leq 2$$

Raw material  
constraints

Demand constraint

Constraints

Bound  
constraints

Decision variables

Objective function

	Ext. paint	Int. paint	Availability
M1	6	4	24
M2	1	2	6
Profit	5	4	

Daily demand for interior paint cannot exceed that for exterior paint by more than 1 Unit.

# Example 3

A farmer has 2400 m of fencing and wants to fence off a rectangular field that borders a straight river. No fence is required along the river. What are the dimensions of the field that has the largest area?

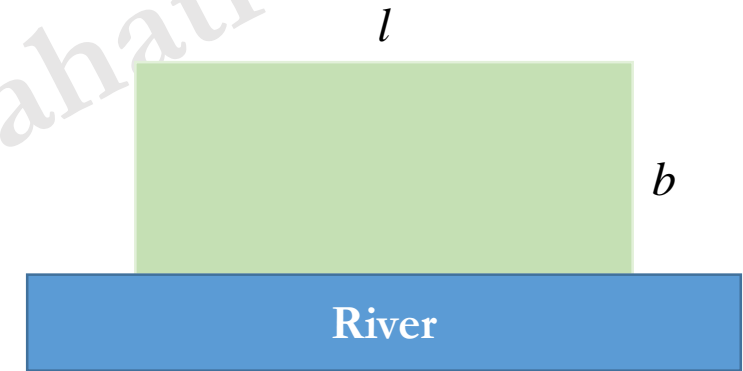
Area of a rectangular region =  $lb$

Objective: Maximize the area,  $\max f = lb$

Constraint: Perimeter should not exceed 2400m

$$l + 2b \leq 2400$$

Bounds:  $l, b > 0$



# Constraints

➤ Inequality constraints (usually resource/requirement constraints)

➤ In general denoted by  $g(x) \leq 0$

➤ Conversion from one form to the other by multiplying by -1

➤ Equality Constraints (usually material and energy balances)

➤ In general denoted by  $g(x) = 0$

➤ Feasible solution: Satisfies all the constraints

➤ Infeasible solution: Does not satisfy at least one constraint

# Constraints

- Any feasible solution is preferred to any infeasible solution
- Among two feasible solutions, the one having a better objective function value is preferred
- Among two infeasible solutions, the one having a smaller constraint violation is preferred

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# Constraints

- Any feasible solution is preferred to any infeasible solution
- Among two feasible solutions, the one having a better objective function value is preferred
- Among two infeasible solutions, the one having a smaller constraint violation is preferred
- A solution  $\mathbf{x}^{(i)}$  is said to be *constrain – dominate* a solution  $\mathbf{x}^{(j)}$ :
  - Solution  $\mathbf{x}^{(i)}$  is feasible and solution  $\mathbf{x}^{(j)}$  is not.
  - Solutions  $\mathbf{x}^{(i)}$  and  $\mathbf{x}^{(j)}$  are both infeasible, but solution  $\mathbf{x}^{(i)}$  has a smaller constraint violation.
  - Solutions  $\mathbf{x}^{(i)}$  and  $\mathbf{x}^{(j)}$  are feasible and solution  $\mathbf{x}^{(i)}$  dominates solution  $\mathbf{x}^{(j)}$  with respect to objective function.

# Constraints

- Any feasible solution is preferred to any infeasible solution
- Among two feasible solutions, the one having a better objective function value is preferred
- Among two infeasible solutions, the one having a smaller constraint violation is preferred
- A solution  $x^{(i)}$  is said to be *constrain – dominate* a solution  $x^{(j)}$ :
  - Solution  $x^{(i)}$  is feasible and solution  $x^{(j)}$  is not.
    - (S1 constrain-dominate S2)
  - Solutions  $x^{(i)}$  and  $x^{(j)}$  are both infeasible, but solution  $x^{(i)}$  has a smaller constraint violation.
    - (S3 constrain-dominate S2)
  - Solutions  $x^{(i)}$  and  $x^{(j)}$  are feasible and solution  $x^{(i)}$  dominates solution  $x^{(j)}$  with respect to objective function.
    - (S4 constrain-dominate S1)

	S1	S2	S3	S4
$x$	0.7	0.4	0.65	0.8
$f$	0.35	0.2	0.325	0.4
violation	0	0.29	0.0275	0

$$\begin{array}{ll}
 \text{Min} & f = \frac{x}{2} \\
 \text{s.t.} & x^2 \geq 0.45 \\
 & 0 < x < 1
 \end{array}$$

# Preserving feasibility of solutions

$$\text{Minimize } 3x_1 + 2x_2 + x_3$$

$$\text{s.t. } 2x_1 + x_2 + x_3 \geq 5$$

$$x_1 + x_2 + 0.5x_3 \leq 4$$

$$x_1 - x_2 = 2$$

$$x_1, x_2, x_3 \in \mathbb{R}$$



$$x_2 = x_1 - 2$$



$$\text{Minimize } 5x_1 + x_3 - 4$$

$$\text{s.t. } 3x_1 + x_3 \geq 7$$

$$2x_1 + 0.5x_3 \leq 6$$

$$x_1, x_3 \in \mathbb{R}$$

# Preserving feasibility of solutions

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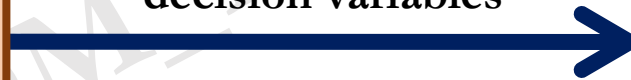
$$2x_1 + 0.5x_3 \leq 6$$

$$x_1, x_3 \in \mathbb{R}$$

## Algorithm

- 1) Generate randomly a single solution/a set of solutions – Partial decision variables
- 2) Based on the fitness of the current solution/set of solutions, suggest other solutions

Partial set of  
decision variables



## Problem Statement

- 1) Generate the remaining set of decision variables
- 2) Evaluate Objective function
- 3) Check for Constraint Satisfaction
- 4) Evaluation of fitness (based on 1,2 & 3)

Fitness



# Constraint Handling

## ➤ Case 1: Unconstrained optimization problem

- Fitness function = Objective function

Solution	Objective function	Fitness function
$S_1$	5	5
$S_2$	7	7

## ➤ Case 2: Constrained optimization problem (Minimization case)

- Fitness function = Objective function +  $\lambda$  (Penalty)

Solution	Objective function	Feasibility
$S_1$	5	Yes
$S_2$	7	No

$$\begin{aligned} &\text{Min } f(x) \\ &s.t. \ g(x) \geq b \end{aligned}$$

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## ➤ Case 2: Constrained optimization problem (Minimization case)

- Fitness function = Objective function +  $\lambda$  (Penalty)

Solution	Objective function	Feasibility	Violation	Fitness function
$S_1$	5	Yes	0	5
$S_2$	7	No	2	47

$$\begin{aligned} &\text{Min } f(x) \\ &s.t. \ g(x) \geq b \end{aligned}$$

$$\lambda = 20$$

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Solution	Objective function	Feasibility	Violation	Fitness function
$S_1$	5	Yes	0	5
$S_2$	7	No	2	47
$S_3$	2	No	0.1	4

$$\begin{aligned} &\text{Min } f(x) \\ &s.t. \ g(x) \geq b \end{aligned}$$

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- Fitness function = Objective function +  $\lambda$  (Penalty)

$$\begin{aligned} &\text{Min } f(x) \\ &s.t. \ g(x) \geq b \end{aligned}$$

Solution	Objective function	Feasibility	Violation	Fitness function $\lambda = 20$	Fitness function $\lambda = 200$
$S_1$	5	Yes	0	5	5
$S_2$	7	No	2	47	407
$S_3$	2	No	0.1	4	22



# Constraint Handling

- Given Problem

$$\begin{array}{ll} \text{Min} & f(x) \\ \text{s.t.} & g_j(x) \geq b \quad j = 1, 2, \dots, J \\ & x_l \leq x_i \leq x_u \quad i = 1, 2, \dots, I \end{array}$$

- Normalize Constraints

$$\underline{g}_j(x) \geq 0 \quad j = 1, 2, \dots, J$$

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- Normalize Constraints

$$\underline{g}_j(x) \geq 0 \quad j = 1, 2, \dots, J$$

- Estimate the extent of violation

$$\omega_j(x) = \begin{cases} |\underline{g}_j(x)|, & \text{if } \underline{g}_j(x) < 0 \\ 0, & \text{otherwise} \end{cases}$$

# Constraint Handling

- Given Problem
$$\begin{array}{ll} \text{Min} & f(x) \\ \text{s.t.} & g_j(x) \geq b \quad j = 1, 2, \dots, J \\ & x_l \leq x_i \leq x_u \quad i = 1, 2, \dots, I \end{array}$$
- Normalize Constraints
$$\underline{g}_j(x) \geq 0 \quad j = 1, 2, \dots, J$$
- Estimate the extent of violation
$$\omega_j(x) = \begin{cases} |\underline{g}_j(x)|, & \text{if } \underline{g}_j(x) < 0 \\ 0, & \text{otherwise} \end{cases}$$
- Estimate the total penalty
$$\Omega(x_i) = \sum_{j=1}^J \omega_j(x)$$
- Estimate the fitness function
$$F(x_i) = f(x) + R_m \Omega(x)$$

Penalty factor ( $R_m$ ): Usually an order of magnitude than the objective function value

# Evaluation of Fitness function

$$\begin{aligned}
 &\text{Minimize} && f(x) = \frac{1+x_2}{x_1} \\
 &s.t. && g_1(x) = 9x_1 + x_2 \geq 6 \\
 & && g_2(x) = 9x_1 - x_2 \geq 1 \\
 & && 0.1 \leq x_1 \leq 1, \quad 0 \leq x_2 \leq 5
 \end{aligned}$$

Normalized Constraints

$$\underline{g}_1(x) = \frac{9x_1 + x_2}{6} - 1 \geq 0$$

$$\underline{g}_2(x) = 9x_1 - x_2 - 1 \geq 0$$

$$R_m = 20$$

$$\omega_j(x) = \begin{cases} \left| \underline{g}_j(x) \right|, & \text{if } \underline{g}_j(x) < 0 \\ 0, & \text{otherwise} \end{cases}$$

$$\Omega(x) = \sum_{j=1}^J \omega_j(x)$$

$$F(x) = f(x) + R_m \Omega(x)$$

Solution	Decision Variables		Objective function (f)
	$x_1$	$x_2$	
<i>S1</i>	0.31	0.89	6.10
<i>S2</i>	0.38	2.73	9.82
<i>S3</i>	0.22	0.56	7.09
<i>S4</i>	0.59	3.63	7.85
<i>S5</i>	0.46	2.9	8.48
<i>S6</i>	0.66	4.11	7.74

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$$\Omega(x) = \sum_{j=1}^J \omega_j(x)$$

$$F(x) = f(x) + R_m \Omega(x)$$

Solution	Decision Variables		Objective function (f)	Feasibility	
	$x_1$	$x_2$		$g_1$	$g_2$
<i>S1</i>	0.31	0.89	6.10	✗	✓
<i>S2</i>	0.38	2.73	9.82	✓	✗
<i>S3</i>	0.22	0.56	7.09	✗	✓
<i>S4</i>	0.59	3.63	7.85	✓	✓
<i>S5</i>	0.46	2.9	8.48	✓	✓
<i>S6</i>	0.66	4.11	7.74	✓	✓

# Evaluation of Fitness function

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$$\underline{g}_1(x) = \frac{9x_1 + x_2}{6} - 1 \geq 0$$

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$$R_m = 20$$

$$\omega_j(x) = \begin{cases} \left| \underline{g}_j(x) \right|, & \text{if } \underline{g}_j(x) < 0 \\ 0, & \text{otherwise} \end{cases}$$

$$\Omega(x) = \sum_{j=1}^J \omega_j(x)$$

$$F(x) = f(x) + R_m \Omega(x)$$

Solution	Decision Variables		Objective function (f)	Feasibility		Violations		Total penalty $\Omega$	Fitness (F)
	$x_1$	$x_2$		$\underline{g}_1$	$\underline{g}_2$	$\omega_1$	$\omega_2$		
<b>S1</b>	0.31	0.89	6.10	✗	✓	0.39	0.00	0.39	13.90
<b>S2</b>	0.38	2.73	9.82	✓	✗	0	0.31	0.31	16.02
<b>S3</b>	0.22	0.56	7.09	✗	✓	0.58	0.00	0.58	18.69
<b>S4</b>	0.59	3.63	7.85	✓	✓	0	0	0	7.85
<b>S5</b>	0.46	2.9	8.48	✓	✓	0	0	0	8.48
<b>S6</b>	0.66	4.11	7.74	✓	✓	0	0	0	7.74

# Example

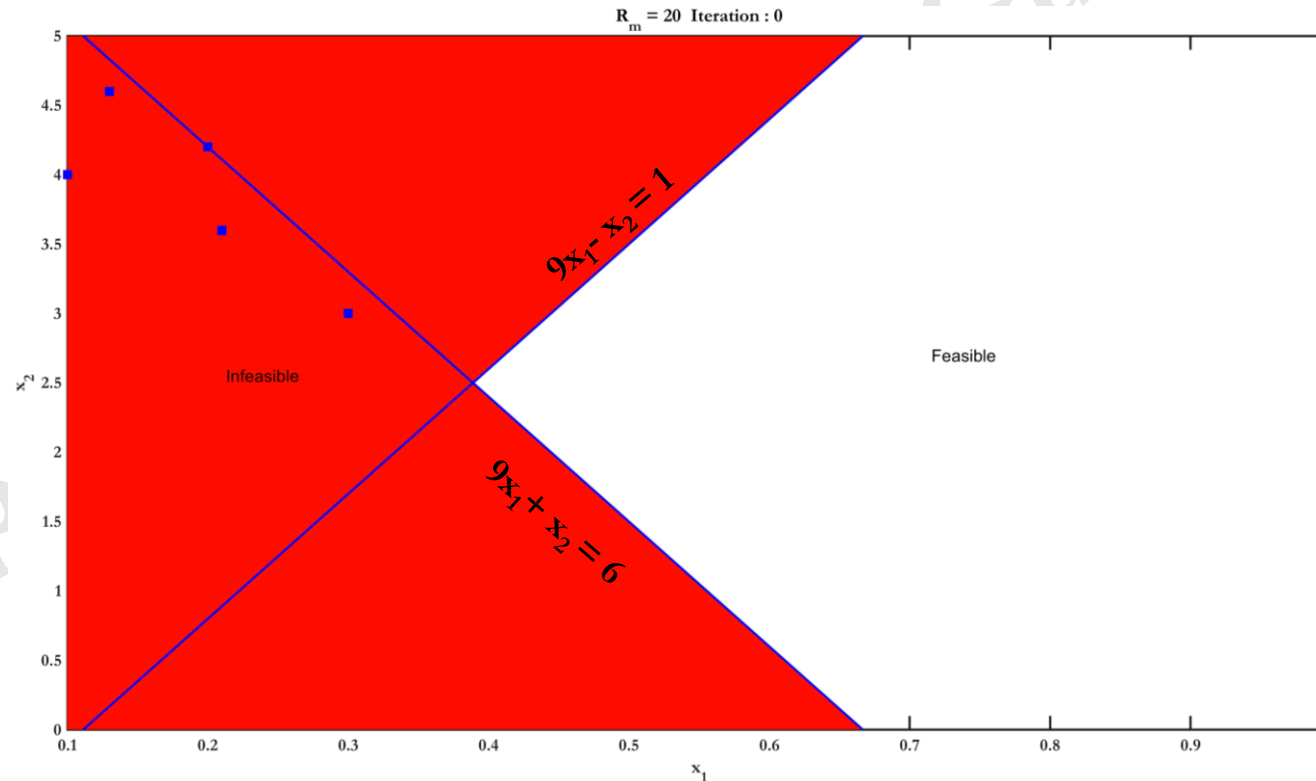
$$\begin{aligned} &\text{Minimize} && f(x) = \frac{1+x_2}{x_1} \\ &s.t. && g_1(x) = x_2 + 9x_1 \geq 6 \\ & && g_2(x) = -x_2 + 9x_1 \geq 1 \\ & && 0.1 \leq x_1 \leq 1, \quad 0 \leq x_2 \leq 5 \end{aligned}$$

Normalized Constraints

$$\underline{g}_1(x) = \frac{9x_1 + x_2}{6} - 1 \geq 0$$

$$\underline{g}_2(x) = 9x_1 - x_2 - 1 \geq 0$$

$$R_m = 20$$



# Constraint Handling

➤ Using hard penalty

$$F(x) = f(x) + \sum_{j=1}^J M_j$$

$$M_j = \begin{cases} \lambda_j, & \text{if } g_j(x) < 0; \\ 0, & \text{otherwise} \end{cases}$$

$$\begin{array}{l} \text{Min } f(x) \\ \text{s.t. } \left. \begin{array}{l} g_j(x) \geq 0 \\ h_k(x) = 0 \end{array} \right\} \forall j = 1, 2, \dots, J \end{array}$$

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## ➤ Using penalty function

$$F(x) = f(x) + \lambda \left[ \sum_{j=1}^J \langle g_j(x) \rangle + \sum_{k=1}^K |h_k(x)| \right]$$

$$\langle g_j(x) \rangle = \begin{cases} |g_j(x)|, & \text{if } g_j(x) < 0; \\ 0, & \text{otherwise} \end{cases}$$

$$\begin{array}{l} g(x) = -2 \\ \langle g(x) \rangle = |-2| = 2 \end{array}$$

$$\begin{array}{l} g(x) = 2 \\ \langle g(x) \rangle = 0 \end{array}$$

# Constraint Handling

➤ Using dynamic penalty function

$$F(x) = f(x) + (C \cdot t)^\alpha \left[ \sum_{j=1}^J \langle g_j(x) \rangle^\beta + \sum_{k=1}^K |h_k(x)|^\beta \right]$$

$C, \alpha, \beta$ : user defined parameters

$t$ : iteration counter

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# Constraint Handling

- Using dynamic penalty function

$$F(x) = f(x) + (C \cdot t)^\alpha \left[ \sum_{j=1}^J \langle g_j(x) \rangle^\beta + \sum_{k=1}^K |h_k(x)|^\beta \right]$$

$C, \alpha, \beta$ : user defined parameters

$t$ : iteration counter

- Method based on feasible over infeasible solutions.

$$F(x) = f(x) + \lambda \left[ \sum_{j=1}^J \langle g_j(x) \rangle + \sum_{k=1}^K |h_k(x)| \right] + R(t, x)$$

$R(t, x)$ : difference between the **best static penalized function value among all infeasible solutions** and **the worst feasible solution**

# Constraint Handling

- Using dynamic penalty function

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$R(t, x)$ : difference between the **best static penalized function value among all infeasible solutions** and **the worst feasible solution**

$$F(x) = \begin{cases} f(x), & \text{if } x \text{ is feasible} \\ f_{\max} + \sum_{j=1}^J \langle g_j(x) \rangle + \sum_{k=1}^K |h_k(x)| & \text{otherwise} \end{cases}$$

$f_{\max}$ : objective function value of worst feasible solution

**Thank You !!!**