

# **Modeling and Optimization with OPL**

## 5 Problems with multiple objective functions

Andreas Popp



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## 5.1 Soft constraints

## 5.2 Maximizing vs. minimizing

## 5.3 Multiple objective functions and Pareto optimality

## 5.4 Multicriteria optimization

## 5.5 Bottleneck objectives

## Maximin and minimax problems

## Maximin and minimax problems

## Explicit modeling of maxima and minima

5.1 Soft  
constraints

5.2 Maximizing vs.  
minimizing

5.3 Multiple  
objective functions  
and Pareto  
optimality

5.4 Multicriteria  
optimization

5.5 Bottleneck  
objectives

Maximin and minimax  
problems

Explicit modeling of  
maxima and minima

# 5.1 Soft constraints

# Example: production problem

5 Problems with  
multiple objective  
functions

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5.1 Soft  
constraints

5.2 Maximizing vs.  
minimizing

5.3 Multiple  
objective functions  
and Pareto  
optimality

5.4 Multicriteria  
optimization

5.5 Bottleneck  
objectives

Maximin and minimax  
problems

Explicit modeling of  
maxima and minima

$$\begin{aligned} \max \quad & \sum_{i \in I} p_i \cdot x_i \\ s.t. \quad & \sum_{i \in I} v_{ri} \cdot x_i \leq c_r \quad \forall r \in R \quad (\text{I}) \\ & x_i \geq 0 \quad \forall i \in I \end{aligned}$$

Constraint (I) is a “hard” constraint and must be fulfilled completely.

5.1 Soft  
constraints5.2 Maximizing vs.  
minimizing5.3 Multiple  
objective functions  
and Pareto  
optimality5.4 Multicriteria  
optimization5.5 Bottleneck  
objectivesMaximin and minimax  
problemsExplicit modeling of  
maxima and minima

$$\begin{aligned} \max \quad & \sum_{i \in I} p_i \cdot x_i \\ s.t. \quad & \sum_{i \in I} v_{ri} \cdot x_i \leq c_r + o_r \quad \forall r \in R \quad (I) \\ & x_i, o_r \geq 0 \quad \forall i \in I, r \in R \end{aligned}$$

Problem: no optimal solution, because the solution space is unbound in the direction of optimization.

5.1 Soft  
constraints5.2 Maximizing vs.  
minimizing5.3 Multiple  
objective functions  
and Pareto  
optimality5.4 Multicriteria  
optimization5.5 Bottleneck  
objectivesMaximin and minimax  
problemsExplicit modeling of  
maxima and minima

$$\begin{aligned} \max \quad & \sum_{i \in I} p_i \cdot x_i - \sum_{r \in R} k_r \cdot o_r \\ s.t. \quad & \sum_{i \in I} v_{ri} \cdot x_i \leq c_r + o_r \quad \forall r \in R \quad (\text{I}) \\ & o_r \leq m_r \quad \forall r \in R \quad (\text{II}) \\ & x_i, o_r \geq 0 \quad \forall i \in I, r \in R \end{aligned}$$

## Example: production problem with complete utilisation

## 5.1 Soft constraints

## Maximin and minimax problems

## Explicit modeling of maxima and minima

$$\begin{aligned} \max \quad & \sum_{i \in I} p_i \cdot x_i \\ s.t. \quad & \sum_{i \in I} v_{ri} \cdot x_i = c_r \quad \forall r \in R \quad (\text{I}) \\ & x_i \geq 0 \quad \forall i \in I \end{aligned}$$

Constraint (I) is a “hard” constraint and must be fulfilled completely.

# Soft equality constraints

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multiple objective  
functions

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5.1 Soft  
constraints

5.2 Maximizing vs.  
minimizing

5.3 Multiple  
objective functions  
and Pareto  
optimality

5.4 Multicriteria  
optimization

5.5 Bottleneck  
objectives

Maximin and minimax  
problems

Explicit modeling of  
maxima and minima

$$\begin{aligned} \max \quad & \sum_{i \in I} p_i \cdot x_i - \sum_{r \in R} k_r \cdot |o_r| \\ s.t. \quad & \sum_{i \in I} v_{ri} \cdot x_i = c_i + o_r \quad \forall r \in R \quad (I) \\ & x_i \geq 0, \quad o_r \leqslant 0 \quad \forall i \in I, r \in R \end{aligned}$$

Problem: The absolute value is not a linear function.

# Soft equality constraints

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Solution: Substitute  $o_r = o_r^+ - o_r^-$

## 5.1 Soft constraints

## Maximin and minimax problems

## Explicit modeling of maxima and minima

$$\begin{aligned}
 & \max \quad \sum_{i \in I} p_i \cdot x_i - \sum_{r \in R} k_r \cdot (o_r^+ + o_r^-) \\
 & s.t. \quad \sum_{i \in I} v_{ri} \cdot x_i = c_i + o_r^+ - o_r^- \quad \forall r \in R \\
 & \quad x_i, o_r^+, o_r^- \geq 0 \quad \forall i \in I, r \in R
 \end{aligned} \tag{I}$$

Be aware when eliminating absolute values

The decomposition of a variable in two summands is not unique. We have to make sure that one of the summands is zero.

5.1 Soft  
constraints

5.2 Maximizing vs.  
minimizing

5.3 Multiple  
objective functions  
and Pareto  
optimality

5.4 Multicriteria  
optimization

5.5 Bottleneck  
objectives

Maximin and minimax  
problems

Explicit modeling of  
maxima and minima

## 5.2 Maximizing vs. minimizing

# Maximizing vs. minimizing

5 Problems with  
multiple objective  
functions

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5.1 Soft  
constraints

5.2 Maximizing vs.  
minimizing

5.3 Multiple  
objective functions  
and Pareto  
optimality

5.4 Multicriteria  
optimization

5.5 Bottleneck  
objectives

Maximin and minimax  
problems

Explicit modeling of  
maxima and minima

Minimizing and maximizing are identical procedures.  
It holds:

$$\max_{x \in X} f(x) = - \min_{x \in X} -f(x)$$

Only the sign of the optimal value changes.

5.1 Soft  
constraints

5.2 Maximizing vs.  
minimizing

5.3 Multiple  
objective functions  
and Pareto  
optimality

5.4 Multicriteria  
optimization

5.5 Bottleneck  
objectives

Maximin and minimax  
problems

Explicit modeling of  
maxima and minima

## 5.3 Multiple objective functions and Pareto optimality

# Example: Lewbrandt GmbH

Total capacity: 120 h

Job	1	2	3	4	5
Gross margin	150 k€	100 k€	150 k€	50 k€	70 k€
Revenue	340 k€	190 k€	220 k€	85 k€	215 k€
Waste water	6.2 t	3.5 t	5.8 t	2.4 t	4.8 t
Capacity consumption	65 h	35 h	65 h	15 h	25 h

Which jobs should be accepted?

→ knapsack problem

## Problem

There are three objective functions, so there is no unique optimal solution.

5.1 Soft  
constraints

5.2 Maximizing vs.  
minimizing

5.3 Multiple  
objective functions  
and Pareto  
optimality

5.4 Multicriteria  
optimization

5.5 Bottleneck  
objectives

Maximin and minimax  
problems

Explicit modeling of  
maxima and minima

# Pareto optimality

## Definition: Pareto optimality

A solution is called Pareto optimal, if there is no other solution, which is better in one objective and at least as good in all other objectives.

## Selected solutions of the example „Lewbrandt GmbH“

$x_1$	$x_2$	$x_3$	$x_4$	$x_5$	profit	revenue	waste water	p. o.
0	1	0	1	0	150	275	5.9	yes
0	1	0	1	1	220	490	10.7	no
1	1	0	0	0	250	530	9.7	yes
1	1	0	1	0	300	615	12.1	yes

5.1 Soft  
constraints5.2 Maximizing vs.  
minimizing5.3 Multiple  
objective functions  
and Pareto  
optimality5.4 Multicriteria  
optimization5.5 Bottleneck  
objectivesMaximin and minimax  
problemsExplicit modeling of  
maxima and minima

5.1 Soft  
constraints

5.2 Maximizing vs.  
minimizing

5.3 Multiple  
objective functions  
and Pareto  
optimality

5.4 Multicriteria  
optimization

5.5 Bottleneck  
objectives

Maximin and minimax  
problems

Explicit modeling of  
maxima and minima

## 5.4 Multicriteria optimization

# Example objective function

Objective function as in example “Lewbrandt GmbH”:

- ▶ Profit:

$$\max f_G(\bar{x}) = 150 \cdot x_1 + 100 \cdot x_2 + 150 \cdot x_3 + 50 \cdot x_4 + 70 \cdot x_5$$

- ▶ Revenue:

$$\max f_U(\bar{x}) = 340 \cdot x_1 + 190 \cdot x_2 + 220 \cdot x_3 + 85 \cdot x_4 + 215 \cdot x_5$$

- ▶ Waste water:

$$\max f_A(\bar{x}) = -6,2 \cdot x_1 - 3,5 \cdot x_2 - 5,8 \cdot x_3 - 2,4 \cdot x_4 - 4,8 \cdot x_5$$

## Weighted objectives

Compose **one** comprehensive objective function by weighing the objectives and adding them together.

## Weighted objectives in example “Lewbrandt GmbH”

weights:  $a_g = 5$ ,  $a_U = 1$ ,  $a_A = 50$

new objective function:

$$\begin{aligned}\max f(\bar{\mathbf{x}}) &= a_g \cdot f_G(\bar{\mathbf{x}}) + a_U \cdot f_U(\bar{\mathbf{x}}) + a_A \cdot f_A(\bar{\mathbf{x}}) \\ &= 5 \cdot f_G(\bar{\mathbf{x}}) + 1 \cdot f_U(\bar{\mathbf{x}}) + 50 \cdot f_A(\bar{\mathbf{x}})\end{aligned}$$

Model: Multicriteria knapsack problem (weighted objectives)

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## Index sets:

/ set of items

$O$  set of objectives

#### Parameters:

$w_i$  weight of item  $i \in I$

$u_{oi}$  value of item  $i \in I$  w.r.t. objective  $o \in O$

$c$  knapsack's capacity

$a_o$  weight of objective  $o \in O$

## Decision variables:

$x_i$  binary decision variable; represents item  $i \in I$  being packed

## **Model description:**

$$\begin{aligned} \max \quad & \sum_{o \in O} a_o \sum_{i \in I} u_{oi} \cdot x_i \\ s.t. \quad & \sum_{i \in I} w_i \cdot x_i \leq c \\ & x_i \in \{0,1\} \quad \forall i \in I \end{aligned} \quad (\text{I})$$

## 5.4 Multicriteria optimization

## Maximin and minimax problems

## Explicit modeling of maxima and minima

## Main objective & aspiration levels

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Choose **one** objective as main objective. Define aspiration levels for the other objectives, which will be asserted by constraints.

Main objective & aspiration levels in example “Lewbrandt GmbH”

Let the waster water emission be the main objective. We want to achieve at least 225 k€ of profit and 480 k€ of revenue:

$$\begin{aligned} \max \quad & f_A(\bar{\mathbf{x}}) \\ s.t. \quad & f_A(\bar{\mathbf{x}}) \geq 225 \\ & f_U(\bar{\mathbf{x}}) \geq 480 \end{aligned}$$

## Model: Multicriteria knapsack problem (main objective)

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## Index sets:

/ set of items

$O$  set of objectives

## Parameters:

$w_i$  weight of item  $i \in I$

$u_{oi}$  value of item  $i \in I$  w.r.t. objective  $o \in O$

c knapsack's capacity

$h$  main objective  $h \in O$

$a_o$  aspiration level of objective  $o \in O \setminus \{h\}$

## Decision variables:

$x_i$  binary decision variable; represents item  $i \in I$  being packed

## Model description:

$$\max \quad \sum_{i \in I} u_{hi} \cdot x_i$$

$$s.t. \quad \sum_{i \in I} w_i \cdot x_i \leq c \quad (I)$$

$$\sum_{i \in I} u_{oi} \cdot x_i \geq a_o \quad \forall o \in O \setminus \{h\} \quad (\text{II})$$

$$x_i \in \{0,1\} \quad \forall i \in I$$

## 5.4 Multicriteria optimization

## Maximin and minimax problems

Explicit modeling of  
maxima and minima

# Goal programming (classic)

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Choose a goal value for all objective functions and penalize deviation from those target values.

Goal programmimg in example “Lewbrandt GmbH”

Goal values:  $a_G = 220$ ,  $a_U = 480$ ,  $a_A = -11$

$$\min |z_G| + |z_U| + |z_A|$$

$$s.t. \quad f_G(\bar{\mathbf{x}}) = 220 + z_G$$

$$f_U(\bar{\mathbf{x}}) = 480 + z_U$$

$$f_A(\bar{\mathbf{x}}) = -11 + z_A$$

5.1 Soft  
constraints

5.2 Maximizing vs.  
minimizing

5.3 Multiple  
objective functions  
and Pareto  
optimality

5.4 Multicriteria  
optimization

5.5 Bottleneck  
objectives

Maximin and minimax  
problems

Explicit modeling of  
maxima and minima

# Model: Multicriteria knapsack problem (GP1)

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## Index sets:

$I$  set of items

$O$  set of objectives

## Parameters:

$w_i$  weight of item  $i \in I$

$u_{oi}$  value of item  $i \in I$  w.r.t. objective  $o \in O$

$c$  knapsack's capacity

$a_o$  goal value for objective  $o \in O$

## Decision variables:

$x_i$  binary decision variable; represents item  $i \in I$  being packed

$z_o$  deviation from goal value of objective  $o \in O$

## Model description:

$$\min \sum_{o \in O} |z_o|$$
$$s.t. \quad \sum_{i \in I} w_i \cdot x_i \leq c \quad (I)$$

$$\sum_{i \in I} u_{oi} \cdot x_i = a_o + z_o \quad \forall o \in O \quad (II)$$
$$x_i \in \{0,1\}, z_o \leq 0 \quad \forall i \in I, o \in O$$

5.1 Soft  
constraints

5.2 Maximizing vs.  
minimizing

5.3 Multiple  
objective functions  
and Pareto  
optimality

5.4 Multicriteria  
optimization

5.5 Bottleneck  
objectives

Maximin and minimax  
problems

Explicit modeling of  
maxima and minima

# Goal Programming (extended version)

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Penalize only unwanted deviation and use weights for deviations.

Goal programming in example “Lewbrandt GmbH”

$$\min \quad w_G \cdot z_G + w_U \cdot z_U + w_A \cdot z_A$$

$$s.t. \quad f_G(\bar{x}) \geq 220 - z_G$$

$$f_U(\bar{x}) \geq 480 - z_U$$

$$f_A(\bar{x}) \geq -11 - z_A$$

5.1 Soft  
constraints

5.2 Maximizing vs.  
minimizing

5.3 Multiple  
objective functions  
and Pareto  
optimality

5.4 Multicriteria  
optimization

5.5 Bottleneck  
objectives

Maximin and minimax  
problems

Explicit modeling of  
maxima and minima

# Modell: Multicriteria knapsack problem (GP2)

## Index sets:

$I$  set of items

$O$  set of objectives

## Parameters:

$w_i$  weight of item  $i \in I$

$u_{oi}$  value of item  $i \in I$  w.r.t. objective  $o \in O$

$c$  knapsack's capacity

$a_o$  goal value of objective  $o \in O$

$b_o$  Abweichungskosten für Ziel  $o \in O$

## Decision variables:

$x_i$  binary decision variable; represents item  $i \in I$  being packed

$z_o$  deviation from goal value of objective  $o \in O$

## Model description:

$$\min \sum_{o \in O} b_o \cdot z_o$$

$$s.t. \quad \sum_{i \in I} w_i \cdot x_i \leq c \quad (I)$$

$$\sum_{i \in I} u_{oi} \cdot x_i \geq a_o - z_o \quad \forall o \in O \quad (II)$$

$$x_i \in \{0,1\}, z_o \geq 0 \quad \forall i \in I, o \in O$$

# Lexicographical ordering of solutions

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With a strict objective hierarchy it is possible to achieve a lexicographical ordering of the solutions.

Selected lexicographically ordered solutions of the example “Lewbrandt GmbH”

Let the objective hierarchy be: profit > revenue > waste water

$x_1$	$x_2$	$x_3$	$x_4$	$x_5$	profit	revenue	waste water
1	1	0	1	0	300	615	12,1
0	1	1	1	0	300	495	11,7
1	0	0	1	1	270	640	13,4
1	1	0	0	0	250	530	9,7
0	1	1	0	0	250	410	9,3

5.1 Soft  
constraints

5.2 Maximizing vs.  
minimizing

5.3 Multiple  
objective functions  
and Pareto  
optimality

5.4 Multicriteria  
optimization

5.5 Bottleneck  
objectives

Maximin and minimax  
problems

Explicit modeling of  
maxima and minima

## Algorithm: Preemptive Goal Programming

1. Let  $i = 1$
2. Solve the problem with the objective function  $f_i$  of objective  $i$ . Get the optimal solution  $\mathbf{x}^*$  with the optimal value  $f_i^*$ .
3. if  $i = n$ :  $\mathbf{x}^*$  is the lexicographically optimal solution.  
Stop.
4. Add the following constraint to the model:  
$$f_i(\mathbf{x}) = f_i^*$$
5. Let  $i = i + 1$  and go to step 2.

5.1 Soft  
constraints

5.2 Maximizing vs.  
minimizing

5.3 Multiple  
objective functions  
and Pareto  
optimality

5.4 Multicriteria  
optimization

5.5 Bottleneck  
objectives

Maximin and minimax  
problems

Explicit modeling of  
maxima and minima

5.1 Soft  
constraints

5.2 Maximizing vs.  
minimizing

5.3 Multiple  
objective functions  
and Pareto  
optimality

5.4 Multicriteria  
optimization

5.5 Bottleneck  
objectives

Maximin and minimax  
problems

Explicit modeling of  
maxima and minima

## 5.5 Bottleneck objectives

## Example: Arabasta County

town	concert hall	water park	museum
Alubarna	1,45 M\$	1,25 M\$	1,10 M\$
Nanohana	1,00 M\$	0,95 M\$	0,90 M\$
Erumalu	0,32 M\$	0,28 M\$	0,24 M\$

Each facility can only be built once. Which facility should be built in which town?

## Maximin problems

Multiple equally scaled single objective functions  $f_1, \dots, f_N$ .

The main objective function is:

$$\max \min_{n \in \{1, \dots, N\}} f_n(\bar{\mathbf{x}})$$

## Linearising of maximin problems

Let  $z_{\min} \leq 0$  be an auxiliary variable.

$$\max z_{\min}$$

$$s.t. \quad f_n(\bar{\mathbf{x}}) \geq z_{\min} \quad \forall n \in \{1, \dots, N\}$$

# Minimax problems

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Multiple equally scaled single objective functions  $f_1, \dots, f_N$ .

The main objective function is:

$$\min \max_{n \in \{1, \dots, N\}} f_n(\bar{x})$$

## Linearising of maximin problems

Let  $z_{\max} \leq 0$  be an auxiliary variable.

$$\min z_{\max}$$

$$s.t. \quad f_n(\bar{x}) \leq z_{\max} \quad \forall n \in \{1, \dots, N\}$$

5.1 Soft constraints

5.2 Maximizing vs. minimizing

5.3 Multiple objective functions and Pareto optimality

5.4 Multicriteria optimization

5.5 Bottleneck objectives

Maximin and minimax problems

Explicit modeling of maxima and minima

# Model: maximin assignment problem (Alternative 1)

## Index sets:

$R$  set of resources

$T$  set of tasks

## Parameters:

$p_{tr}$  profit if Task  $t$  is fulfilled by resource  $r$

## Decision variables:

$x_{tr}$  binary variable representing if task  $t$  is fulfilled by Ressource  $r$

$p_{\min}$  auxiliary variable for minimal profit

## Model description:

$$\max p_{\min}$$

$$\text{s.t. } \sum_{r \in R} x_{tr} = 1 \quad \forall t \in T \quad (\text{I})$$

$$\sum_{t \in T} x_{tr} \leq 1 \quad \forall r \in R \quad (\text{II})$$

$$p_{\min} \leq \sum_{r \in R} x_{tr} \cdot p_{tr} \quad \forall t \in T \quad (\text{III})$$

$$x_{rt} \in \{0, 1\}, p_{\min} \leq 0 \quad \forall r \in R, t \in T$$

5.1 Soft  
constraints

5.2 Maximizing vs.  
minimizing

5.3 Multiple  
objective functions  
and Pareto  
optimality

5.4 Multicriteria  
optimization

5.5 Bottleneck  
objectives

Maximin and minimax  
problems

Explicit modeling of  
maxima and minima

# Explicit modeling of maxima and minima

## Explicit modeling of maxima

$$f_n(\bar{\mathbf{x}}) \leq z_{\max} \quad \forall n \in \{1, \dots, N\}$$

$$z_{\max} - f_n(\bar{\mathbf{x}}) \leq M \cdot (1 - y_n) \quad \forall n \in \{1, \dots, N\}$$

$$\sum_{n=1}^N y_n = 1$$

## Explicit modeling of minima

$$f_n(\bar{\mathbf{x}}) \geq z_{\min} \quad \forall n \in \{1, \dots, N\}$$

$$f_n(\bar{\mathbf{x}}) - z_{\min} \leq M \cdot (1 - y_n) \quad \forall n \in \{1, \dots, N\}$$

$$\sum_{n=1}^N y_n = 1$$

5.1 Soft  
constraints5.2 Maximizing vs.  
minimizing5.3 Multiple  
objective functions  
and Pareto  
optimality5.4 Multicriteria  
optimization5.5 Bottleneck  
objectivesMaximin and minimax  
problemsExplicit modeling of  
maxima and minima

5.1 Soft  
constraints5.2 Maximizing vs.  
minimizing5.3 Multiple  
objective functions  
and Pareto  
optimality5.4 Multicriteria  
optimization5.5 Bottleneck  
objectivesMaximin and minimax  
problemsExplicit modeling of  
maxima and minima

# Model: maximin assignment problem (Alternative 2)

## Index sets:

$R$  set of resources

$T$  set of tasks

## Parameters:

$p_{tr}$  profit if Task  $t$  is fulfilled by resource  $r$

$M$  a sufficiently big number

## Decision variables:

$x_{tr}$  binary variable representing if task  $t$  is fulfilled by Resource  $r$

$p_{\min}$  auxiliary variable for minimal profit

$y_t$  binary selection variable for minimum for task  $t$

## Model description:

$$\max p_{\min}$$

$$\text{s.t. } \sum_{r \in R} x_{tr} = 1 \quad \forall t \in T \quad (\text{I})$$

$$\sum_{t \in T} x_{tr} \leq 1 \quad \forall r \in R \quad (\text{II})$$

$$p_{\min} \leq \sum_{r \in R} x_{tr} \cdot p_{tr} \quad \forall t \in T \quad (\text{III})$$

$$\sum_{r \in R} (x_{tr} \cdot p_{tr}) - p_{\min} \leq M \cdot (1 - y_t) \quad \forall t \in T \quad (\text{IV})$$

$$\sum_{t \in T} y_t = 1 \quad (\text{V})$$

$$x_{rt} \in \{0, 1\}, p_{\min} \leq 0 \quad \forall r \in R, t \in T$$