

# **A small tour of optimization models**

## *Theory of games and markets with examples*

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# OPTIMIZATION AND APPLICATIONS

Consulting:

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Web-sites:

*http : //soften.ktu.lt/~mockus*

*http : //pilis.if.ktu.lt/~jmockus*

*http : //eta.ktl.mii.lt/ mockus*

*http : //mockus.us/optimum* (short)

# Optimality

## Objective function

$$f(x), \quad x = (x_1, \dots, x_m). \quad (1)$$

## Global minimum

$$f(x_A) \leq f(x) \quad x \in A, \quad (2)$$

or

$$x_A = \arg \min_A f(x). \quad (3)$$

Here  $A$  is a feasible region.

## Local minimum

$$f(x_\epsilon) \leq f(x) \quad x \in \epsilon. \quad (4)$$

Here  $\epsilon$  is a vicinity of  $x_\epsilon$ .

# Discrete and convex optimization

## Discrete optimization

if variables  $x_i$  are discrete.

## Linear programming

if

$$f(x) = \sum_{i=1}^m c_i x_i, \quad (5)$$

$$A : \sum_{i=1}^m a_{ij} x_i \geq b_j, \quad j = 1, \dots, m, \quad x_i \geq 0. \quad (6)$$

## Convex programming

if both objective  $f(x)$  and feasible region  $A$  are convex.

# Diet problem

$$\min_x \sum_{i=1}^m (c_i - s_i)x_i + g \left( \sum_{i=1}^m a_{i1}x_i - b_1 \right), \quad (7)$$

$$\sum_{i=1}^m a_{ij}x_i \geq b_j, \quad j = 1, \dots, n \quad (8)$$

$$x_i \geq 0, \quad i = 1, \dots, m \quad (9)$$

For example,

$c_1$  is a price of bread,

$a_{11}$  are calories of bread,

$b_1$  are necessary calories.

$s_i$  is the taste (expressed in money units) ,  $g$  beauty (expressed in money units)

Solving by standard simplex or interior point methods.

# Diverse diet problem

$$\min_x \sum_{i=1}^m (c_i - s_i)x_i + g\left(\sum_{i=1}^m a_{i1}x_i - b_1\right) - d\left(\sum_{i=1}^m d_i x_i\right), \quad (10)$$

$$\sum_{i=1}^m a_{ij}x_i \geq b_j, \quad j = 1, \dots, n \quad (11)$$

Here

$d$  is the taste of food diversity (expressed in money units),

$d_i$  is the dish indicator:  $d_i = 1$ , if  $x_i$  is a dish,  $d_i = 0$ ,

otherwise,

If  $x_i$  is a dish then  $x_i = \text{int}$  and  $0 \leq x_i \leq 1$ .

# Longer term diet problem

$$\min_x \sum_{i=1}^m (c_i - s_i)x_i + g\left(\sum_{i=1}^m a_{i1}x_i - Tb_1\right) - d\left(\sum_{i=1}^m d_i x_i\right), \quad (12)$$

$$\sum_{i=1}^m a_{ij}x_i \geq Tb_j, \quad j = 1, \dots, n \quad (13)$$

$$\sum_{i=1}^m d_i x_i \leq T, \quad \sum_{i=1}^m d_i \geq T. \quad (14)$$

Here  $d$  is the taste of food diversity,  $T$  is the time period,  $d_i$  is the dish indicator:  $d_i = 1$ , if  $x_i$  is a dish, otherwise  $d_i = 0$ . If  $x_i$  is a dish then  $x_i = int$  and  $0 \leq y_i \leq 1$ .

# Combinatorial diet problem

$$\min_x \sum_{i=1}^m (c_i - s_i)x_i + g\left(\sum_{i=1}^m a_{i1}x_i - Tb_1\right) - d(x), \quad (15)$$

$$\sum_{i=1}^m a_{ij}x_i \geq Tb_j, \quad j = 1, \dots, n \quad (16)$$

Here  $x = (x_i, i = 1, \dots, m)$ ,  $T$  is the time period,  $x_i$  is a number of dishes (meals)  $i$  during time  $T$ , thus  $x_i = 1, 2, \dots, T$ ,  $d(x)$  is a function defining the taste of food diversity (for example,  $d(x)$  could be a number of different non-zero components  $x_i$  in the diet  $x$ ), Bayesian Heuristic Approach (BHA) is recommended method for improving a user defined initial diet by optimizing parameters of Simulated Annealing (SA).

# Solving combinatorial diet problem

Step 1. Set an initial diet  $x = x^1$  and evaluate its quality by calculating the sum:

$$D(x^1) = \sum_{i=1}^m (c_i - s_i)x_i^1 + g\left(\sum_{i=1}^m a_{i1}x_i^1 - Tb_1\right) - d(x^1), \quad (17)$$

Step 2. Generate next diet  $x^2$  by changing randomly some dishes and evaluate the quality  $D(x^2)$ .

Step 3. If  $h_2 = D(x^2) - D(x^1) \leq 0$  go to  $x^2$ .

If  $h_2 > 0$  go to  $x^2$  with probability  $r_2$ ;

keep  $x^1$  and return to step 1 with probability  $1 - r_2$ .

Probability  $r_2$  is generated by SA formula ?? defined in the slide "Simulated Annealing"

Parameter  $x$  of SA is optimized using GMJ system.

# Defining SA parameters

Step 1. Generate probability  $r_2$  by SA formula 173.

Step 2. Optimize parameter  $x$  of SA by GMJ.

The standart reference to GMJ is this:

```
public Domain domain ()
```

```
return domain;
```

```
public double f (Point pt)
```

```
.....
```

```
return f
```

Here 'domain' defines constraints of SA variables,

'pt' is vector of SA variables,

'f' is the 'goodness' function of the best diet

after 'IT' iterations using fixed SA variables.

# Simplex algorithm

Simple example:

$$\min_x z \quad (18)$$

$$z = x_1 - x_3 \quad (19)$$

$$x_1 + x_2 + x_3 = 1, x_i \geq 0 \quad (20)$$

Here  $x_2$  base variables,  
 $x_1, x_3$  free variables.  $x_1 = 1$  base solution obtained  
when free variables are equal to zero

$x_1 = x_3 = 0$ , then  $z = 0$ .

This base solution is improved by increasing  $x_3$ ,  
because  $c_3 = -1 < 0$ .

New base solution  $x_3 = 1, x_1 = x_2 = 0$ , where  $z = -1$ , can't  
be improved,

since both free variables are non-negative  $c_1 = 1, c_2 = 0$ .