



fast based on LP + enumeration + advanced features

declarative create the model, apply a solver

generic & specific algorithms

MILP perks

optimality primal-dual bounds

EXPICESIVE logic, nonlinear, discrete many decision problems

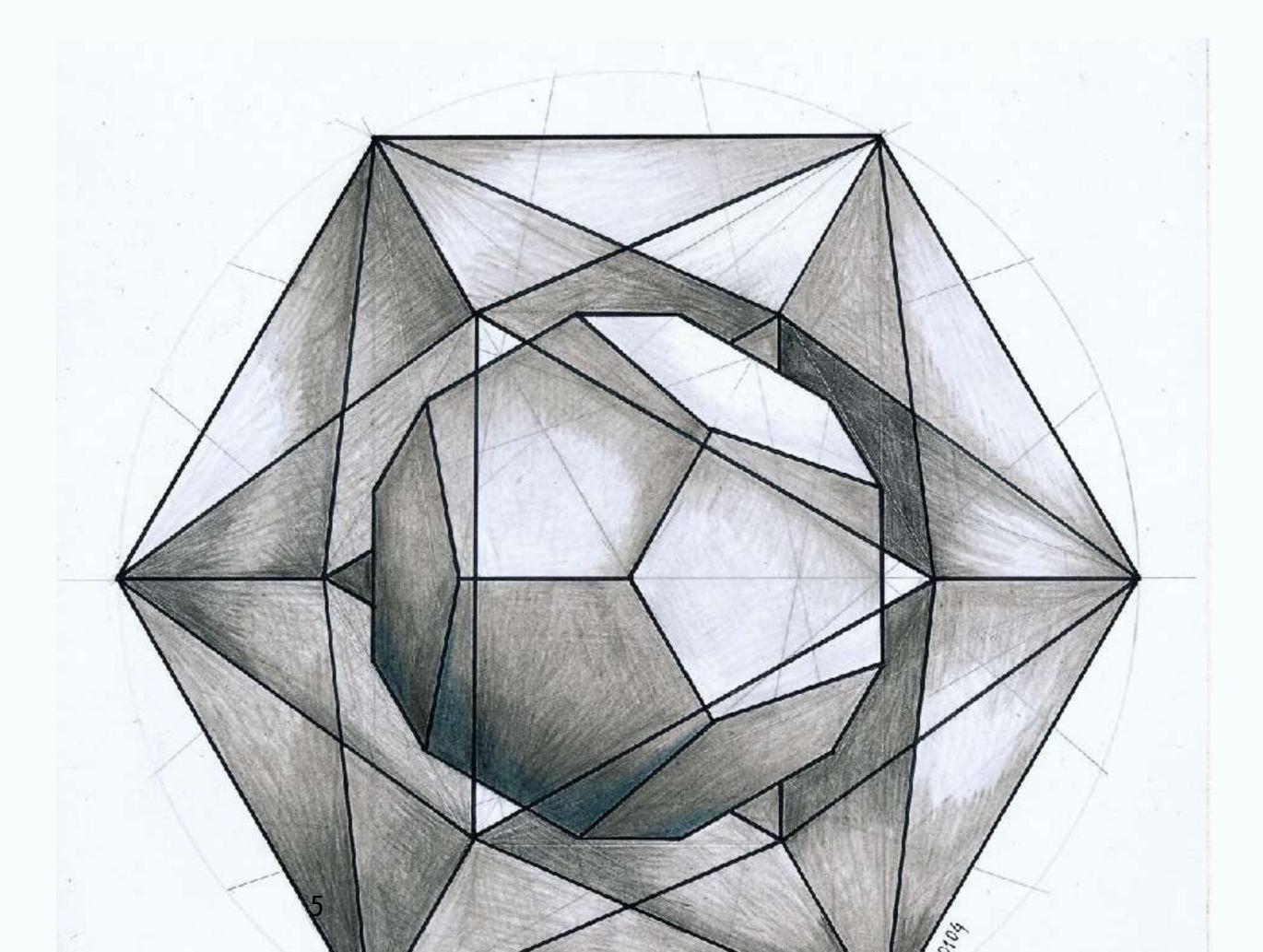
flexible change the model, not the solver

how to model? techniques & applications

how difficult? complexity & distance to LP

how to solve?
main techniques & modern solvers
decomposition methods

how to model?



Mixed Integer Linear Program

$$\min cx$$

$$Ax \ge b$$

$$x \in \mathbb{Z}^p \times \mathbb{R}^{n-p}$$

$$c \in \mathbb{R}^n, A \in \mathbb{R}^{m \times n}, b \in \mathbb{R}^m$$

 $\begin{array}{ll} \text{objective} & cx \\ \text{linear constraints} & Ax \geq b \\ \text{integrity constraints} & x_j \in \mathbb{Z} \\ \text{right hand side} & b \\ \text{cost vector} & c \\ \text{solution space} & \mathbb{R}^n \\ \text{feasible set} & \{x \in \mathbb{Z}^p \times \mathbb{R}^{n-p} \, | \, Ax \geq b \} \\ \end{array}$



true or false

- select item j
- associate item j to resource i
- variable $y \ge 0$ greater than constant a?
- select at most n items

$$x_j = 1, \quad x_j \in \{0,1\}$$

$$x_{ij} = 1, \quad x_{ij} \in \{0,1\}$$

$$y \ge ax, x \in \{0,1\}$$

$$x_1, ..., x_n \in \{0, 1\}$$



$$\max \sum_{j=1}^{n} c_j x_j$$
s.t.
$$\sum_{j=1}^{n} w_j x_j \le K$$

$$x_j \in \{0,1\} \qquad j = 1..n$$

Integer Knapsack Problem

Input n items, value c_j and
weight w_j for each item j,
capacity K.

Output a maximum value subset of items whose total weight does not exceed K.

logic with binaries

x,y binary variables; f continuous variable; a, k, n constants

- either x or y
- if x then y
- if x then f ≤ a
- at most 1 out of n
- at least k out of n

$$x + y = 1$$

$$y \ge x$$

$$ax + M(1 - x)$$

 $y \ge x$ "big M" $f \le ax + M(1-x)$ big enough but keep it tight!

logic with binaries

x,y binary variables; f continuous variable; a, k, n constants

- either x or y
- if x then y
- if x then f ≤ a
- at most 1 out of n
- at least k out of n

$$x + y = 1$$

$$y \ge x$$

$$f \le ax + M(1-x)$$

$$x_1 + \cdots + x_n \le 1$$

$$x_1 + \cdots + x_n \ge k$$

$$\min \sum_{j=1}^{n} c_j x_j + \sum_{j=1}^{n} \sum_{i=1}^{m} d_{ij} y_{ij}$$

s.t.
$$\sum_{j=1}^{n} y_{ij} = 1$$

$$y_{ij} \leq x_j$$

$$x_j \in \{0, 1\}$$

$$y_{ij} \in \{0, 1\}$$

$$i = 1..m$$

$$j = 1..n, i = 1..m$$

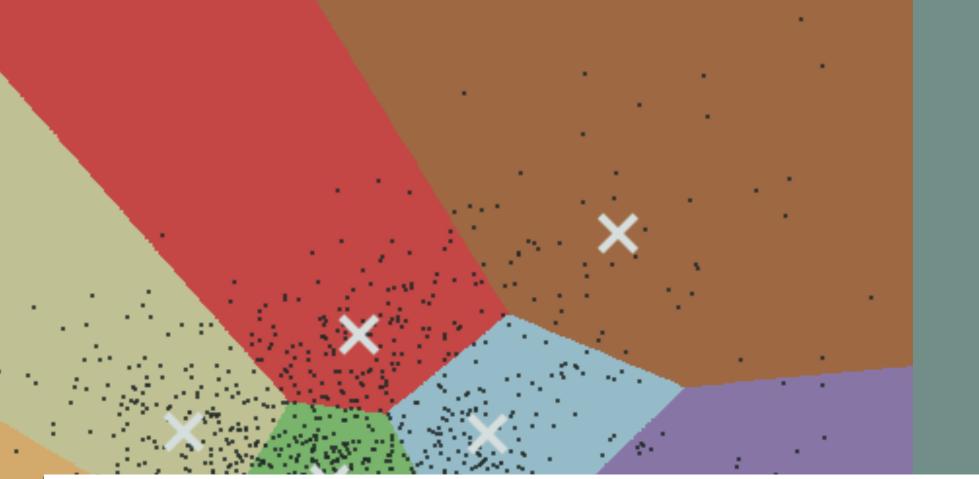
$$j = 1...n$$

$$j = 1..n, i = 1..m$$

Uncapacitated Facility Location Problem

Input m facility locations, m customers, cost cj to open facility j, cost dij to serve customer i from facility j

Output a mimimum (opening and service) cost assignment of customers to facilities.



K-median clustering

$$\min \sum_{i=1}^{n} \sum_{j=1}^{n} d_{ij} y_{ij}$$
s.t.
$$\sum_{j=1}^{n} y_{ij} = 1$$
 $i = 1..n$

$$y_{ij} \le x_{j}$$
 $i, j = 1..n$

$$\sum_{j=1}^{n} x_{j} = k$$

$$y_{ij} \in \{0, 1\}, x_{j} \in \{0, 1\}$$
 $i, j = 1..n$

Input n data points, distance
dij between each two points
i,j, number k of clusters.
Output k centers minimizing
the sum of distances between
each point and its nearest
center.



Input n data points $m_j \in \mathbb{R}^p$, a number K of clusters. Euclidean distance.

K-median clustering

Output define K points as centers so as to minimize the sum of the distances between each point and its nearest center.

K-mean clustering

Output partition the points into K sets so as to minimize the sum of the distances between each point and the mean of points in its cluster.

K-mean clustering

cannot precompute the distance to the centers anymore: modeled with nonlinear constraints

 x_{jk} is j assigned to cluster k? y_k coordinates of the center of k? d_{jk} distance from j to the center of k?

$$\min \sum_{k=1}^{K} \sum_{j=1}^{n} x_{jk} d_{jk}$$

$$s.t \begin{cases} d_{jk} = \sum_{i=1}^{p} (m_j^i - y_k^i)^2 & \forall j, k \end{cases}$$

$$\sum_{k=1}^{K} x_{jk} = 1 \quad \forall j$$

$$x_{jk} \in \{0,1\}, y_k^i \in \mathbb{R}, d_{jk} \ge 0$$

CONVEX

K-mean clustering

```
x_{jk} is j assigned to cluster k? y_k coordinates of the center of k? d_{jk} distance from j to the center of its cluster k?
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$$\min \sum_{k=1}^{K} \sum_{j=1}^{n} d_{jk}$$

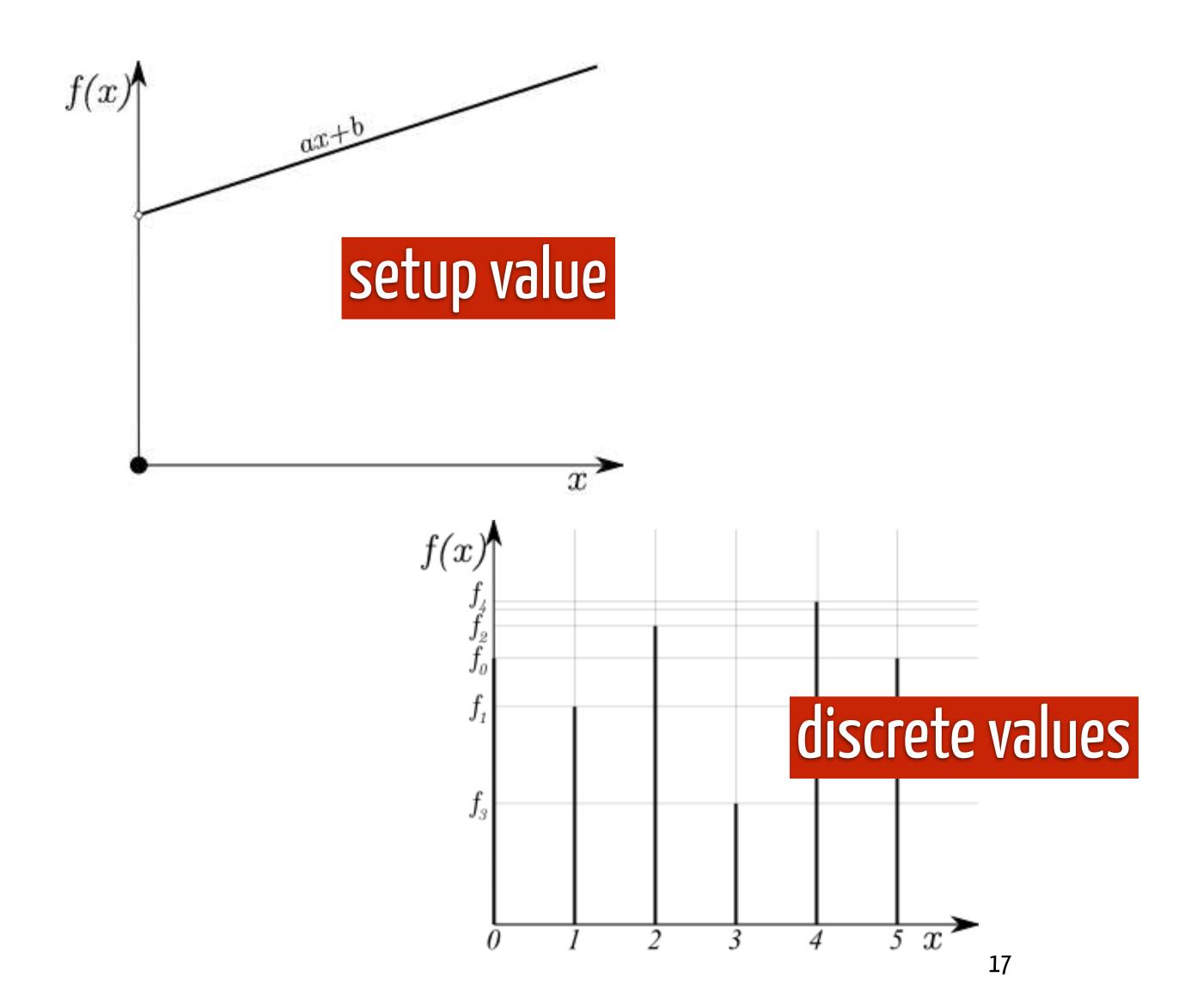
$$s.t \left[d_{jk} \ge \sum_{i=1}^{p} (m_{j}^{i} - y_{k}^{i})^{2} - \overline{d}_{jk} (1 - x_{jk}) \quad \forall j, k \right]$$

$$\sum_{k=1}^{K} x_{jk} = 1 \quad \forall j$$

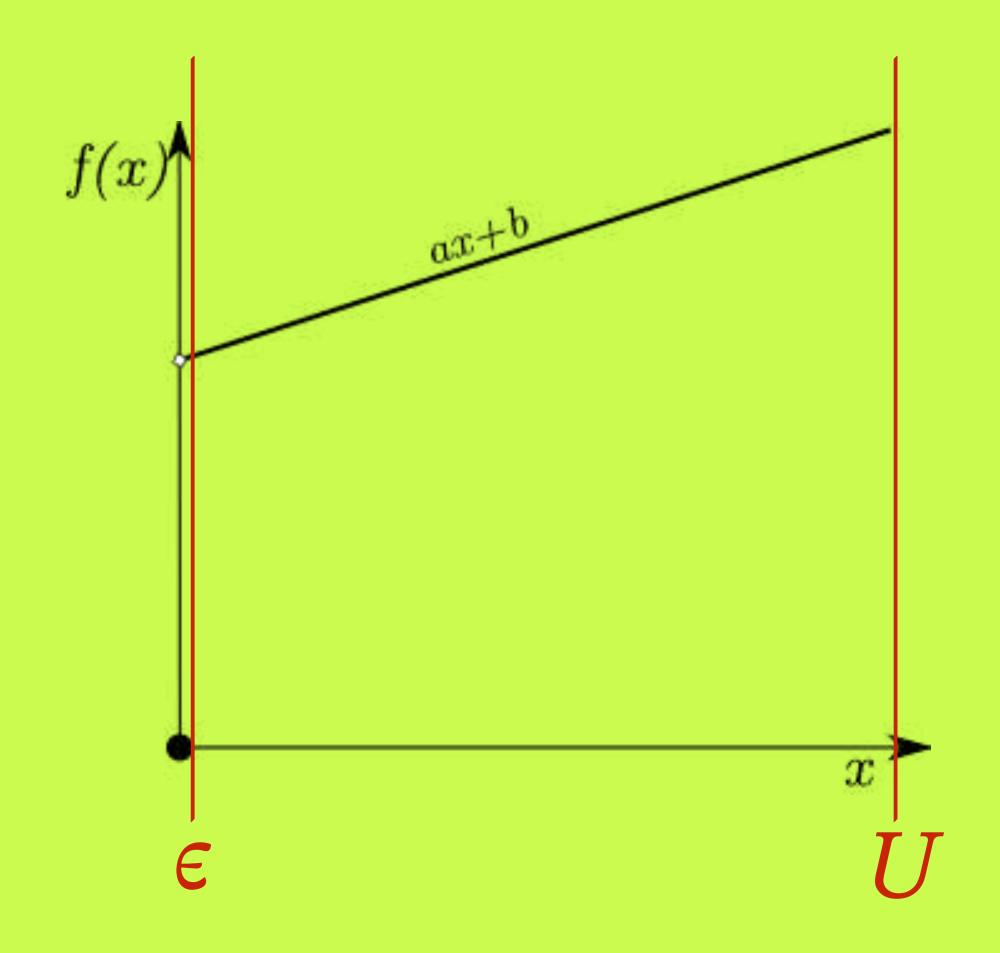
$$x_{jk} \in \{0,1\}, y_{k}^{i} \in \mathbb{R}, d_{jk} \ge 0$$

convexify the nonlinear constraints using big-M (optimization is still nonconvex because of integrality)

non-linear functions







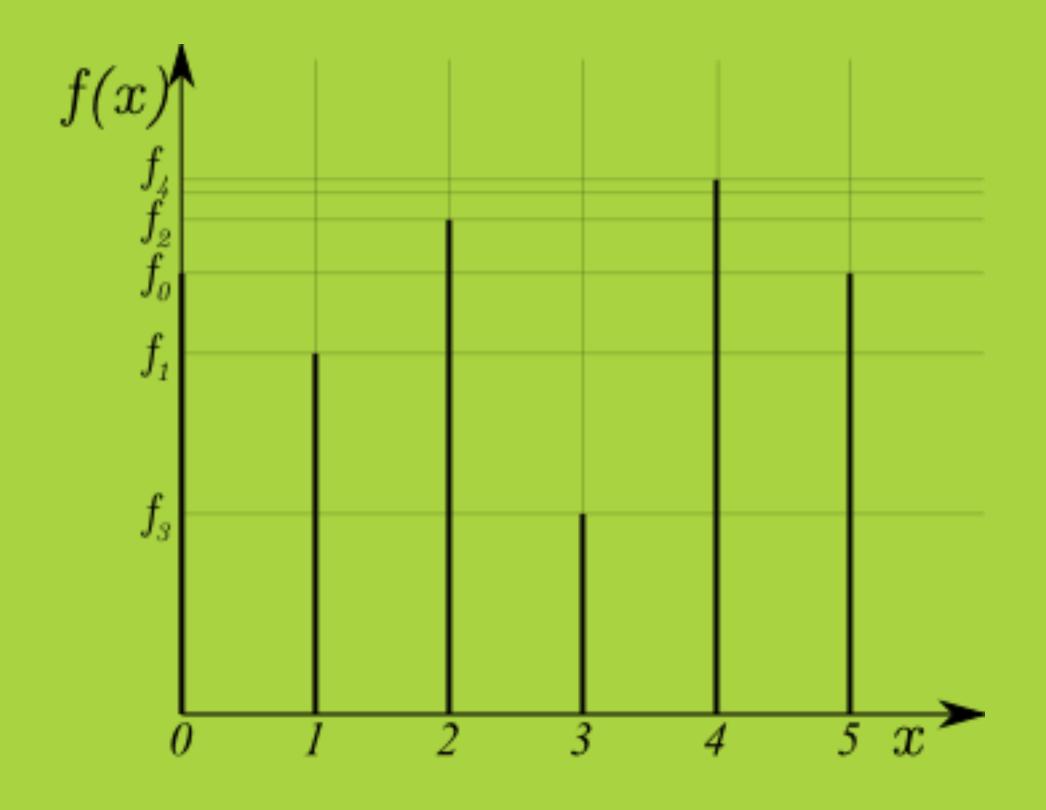
setup value

$$f(x) = ax + b\delta$$

$$\epsilon \delta \le x \le U\delta$$

$$\delta \in \{0, 1\}$$

δ is x positive ?



discrete values

$$f(x) = \sum_{i} \delta_{i} f_{i}$$

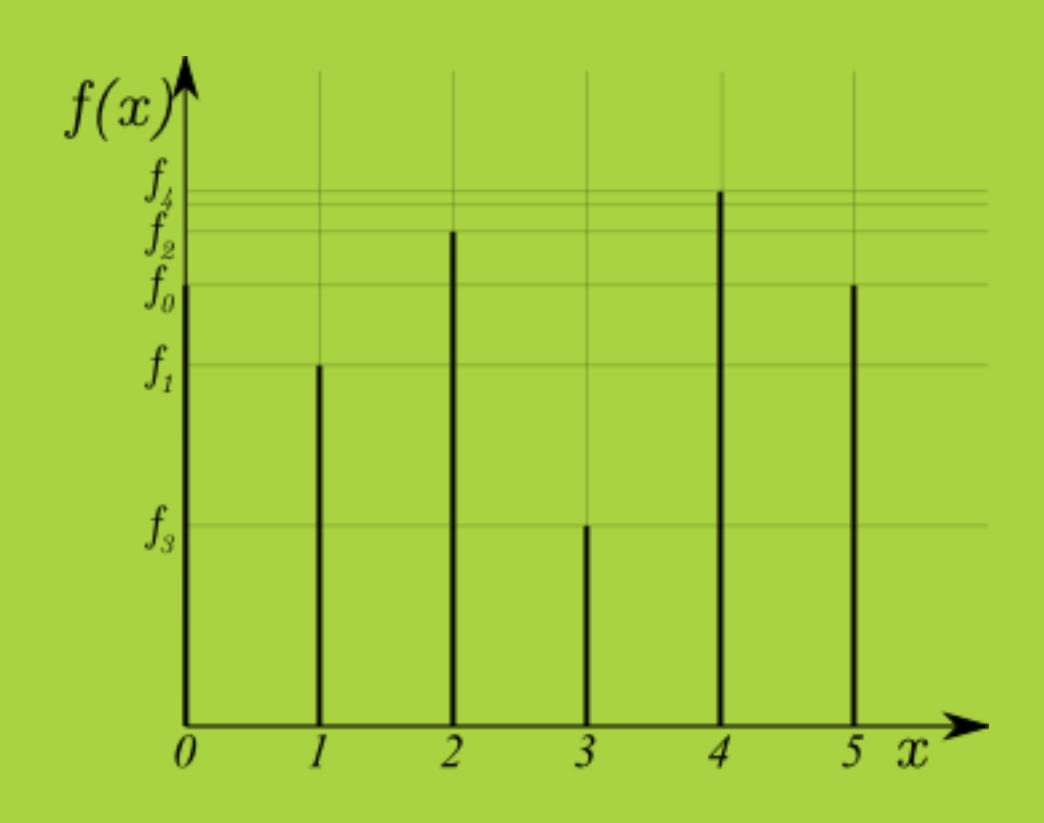
$$\sum_{i} i \delta_{i} = x$$

$$\sum_{i} \delta_{i} = 1$$

$$\delta_{i} \in \{0, 1\} \ i = 0..n$$

 δ_i is x=i (and $f(x)=f_i$)?

Special Ordered Set of type 1: ordered set of variables, all zero except at most one



discrete values

$$f(x) = \sum_{i} \delta_{i} f_{i}$$

$$\sum_{i} i \delta_{i} = x$$

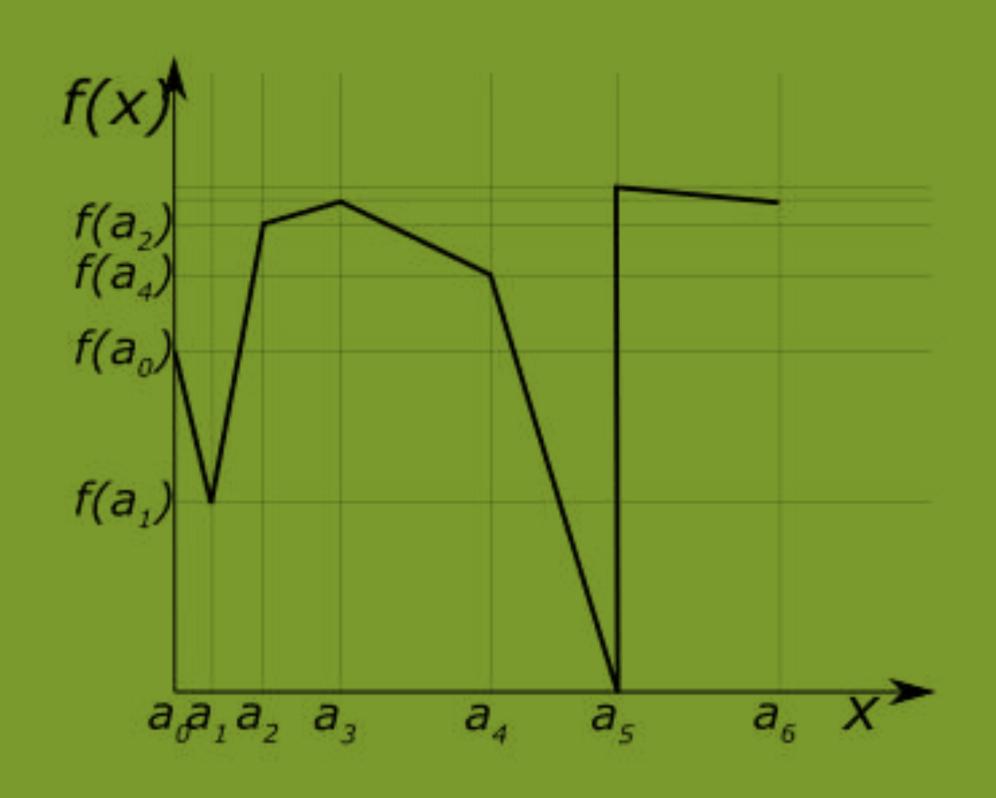
$$\sum_{i} \delta_{i} \ge 1$$

$$\delta_{i} \in \{0, 1\} \ i = 0..n$$

$$SOSI(\delta)$$

$$\delta_i$$
 is $x=i$ (and $f(x)=f_i$)?

Special Ordered Set of type 2: ordered set of variables, all zero except at most two consecutive



piecewise linear

$$f(x) = \sum_{i} \lambda_{i} f(a_{i})$$

$$\sum_{i} a_{i} \lambda_{i} = x$$

$$\sum_{i} \lambda_{i} = 1$$

$$\lambda_{i} \in [0, 1] \ i = 0..n$$

$$SOS2(\lambda)$$

$$\lambda_i$$
 is $x=a_i$? (then $\lambda_i a_i + \lambda_{i+1} a_{i+1}$ in $[a_i, a_{i+1}]$ if $\lambda_i + \lambda_{i+1} = 1$)



modeling with Z

$$\chi_i = 5$$

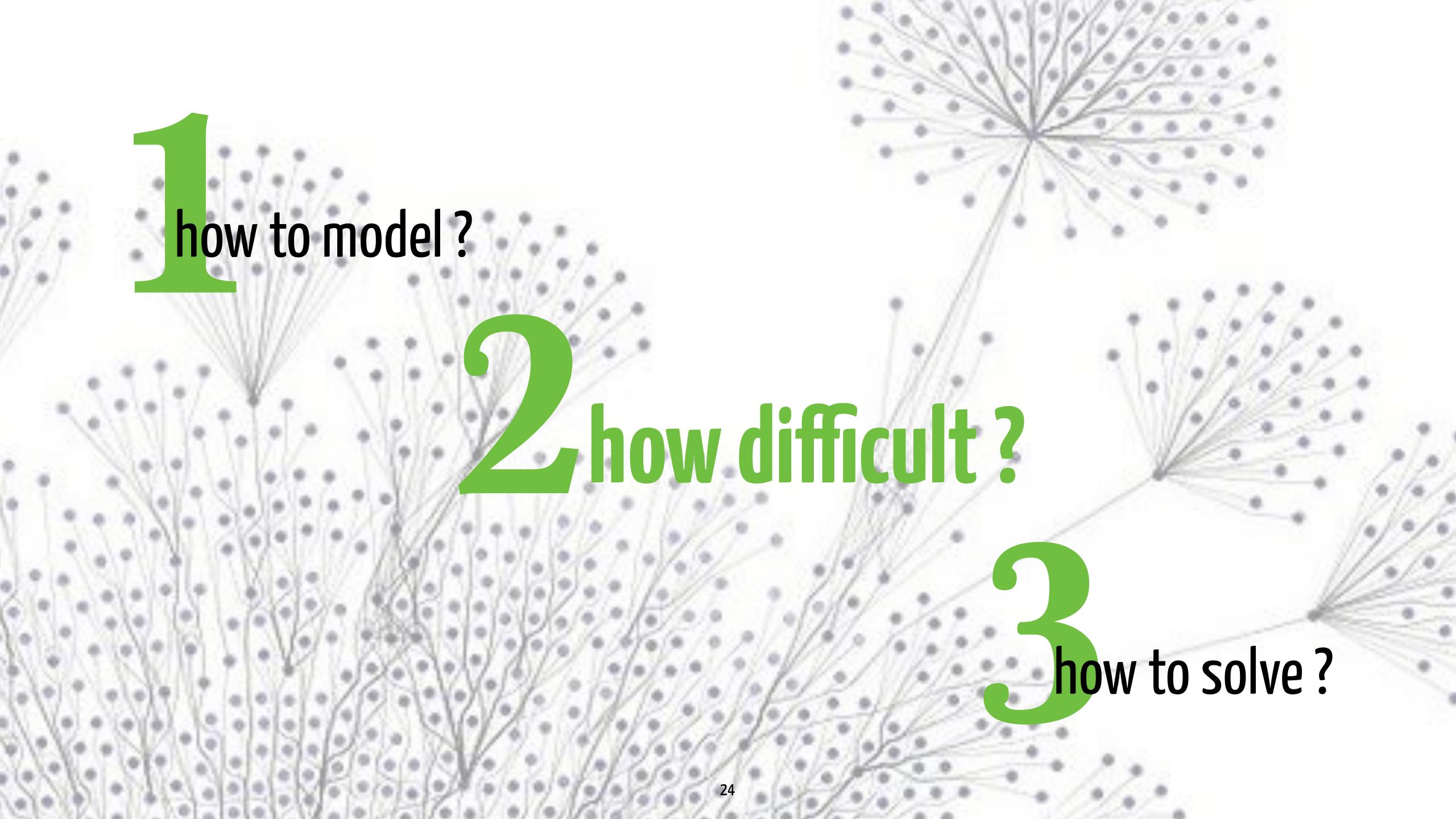
to order i is the 5th item
to count 5 items are selected
to measure time task i starts at time 5
to measure space item i is located on floor 5

$$\simeq \delta_{i5} = 1$$

Binary Integer Linear Program (BIP) {0,1}n

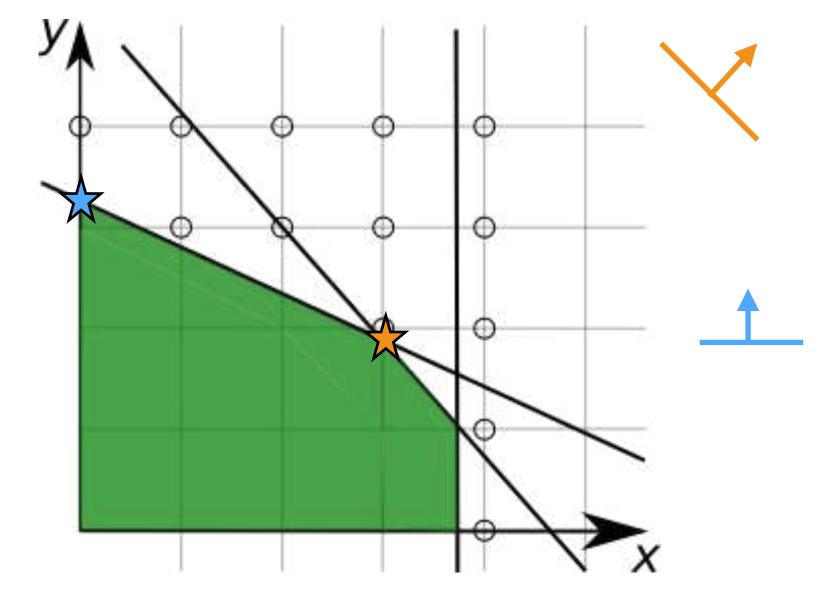
Integer Linear Program (IP) Zn

Mixed Integer Linear Program (MIP) Zn U Qn



Linear Programming cheat sheet

- MILP without integrality = LP relaxation
- LP feasible set = polyhedron
- convex optimization
- if LP is feasible and bounded, at least one vertex is optimal
- primal simplex algorithm: visit adjacent vertices as cost decreases
- strong duality: $\min\{cx \mid Ax \ge b, x \ge 0\} = \max\{ub \mid uA \le c, u \ge 0\}$
- interior point method runs in polynomial time (simplex can be better in practice)



50%

$$\min \sum_{j=1}^{m} s_{j}^{+} + s_{j}^{-}$$
s.t.
$$\sum_{i=1}^{n} a_{ij} x_{i} + s_{j}^{+} - s_{j}^{-} = \frac{d_{j}}{2} \qquad j = 1..m$$

$$x_{i} \in \{0, 1\} \qquad i = 1..n$$

$$s_{j}^{+} \ge 0, s_{j}^{-} \ge 0 \qquad j = 1..m$$

Market Split Problem

Input 1 company, 2 divisions, m
products with availabilities d_j,
n retailers with demands a_{ij} in
each product j.

Output an assignment of the retailers to the divisions approaching a 50/50 production split.

x_i is retailer i assigned to division 1 ? s_j gap to the 50% split goal for product j

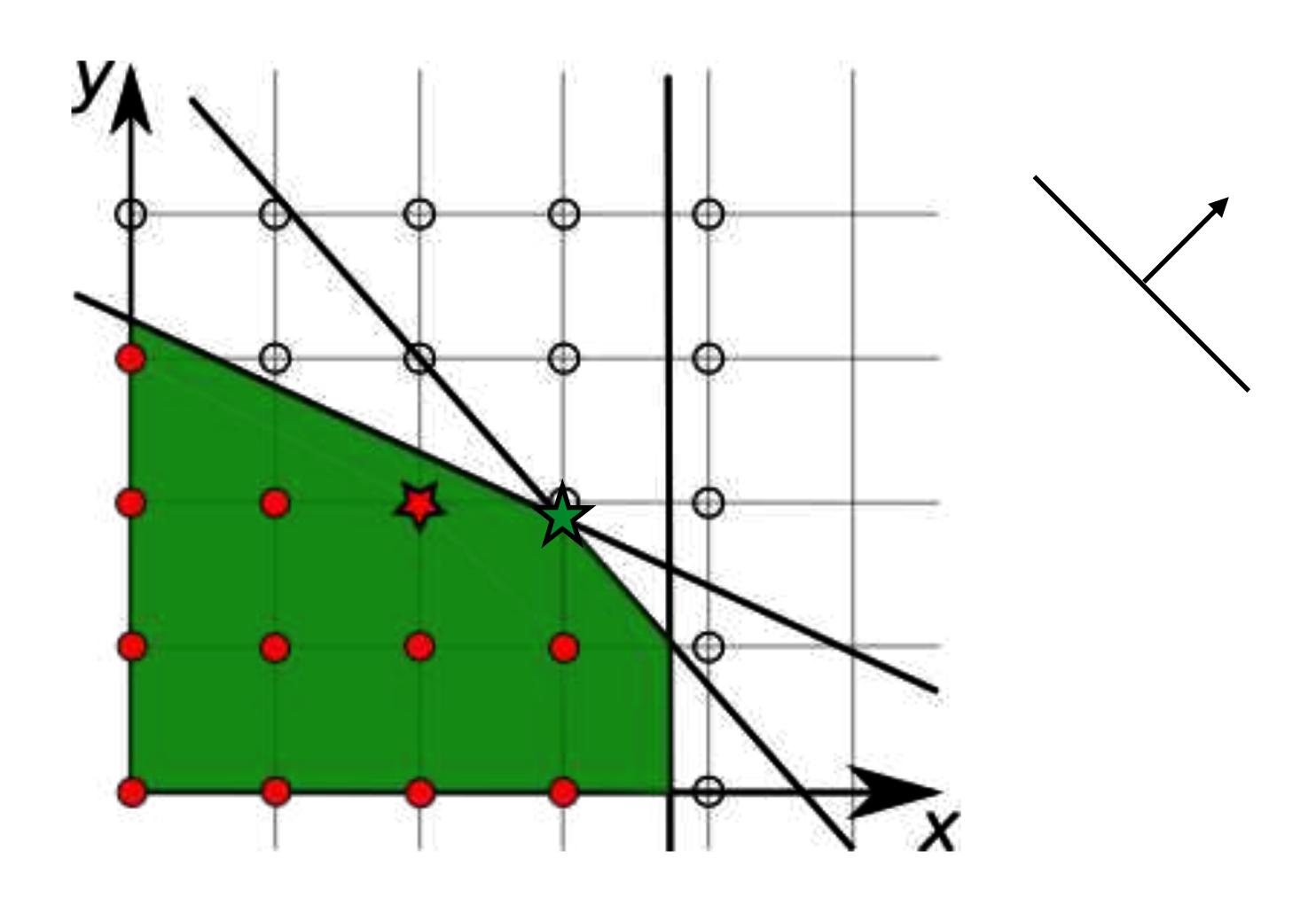
50 %

MIPLIB markshare_5_0

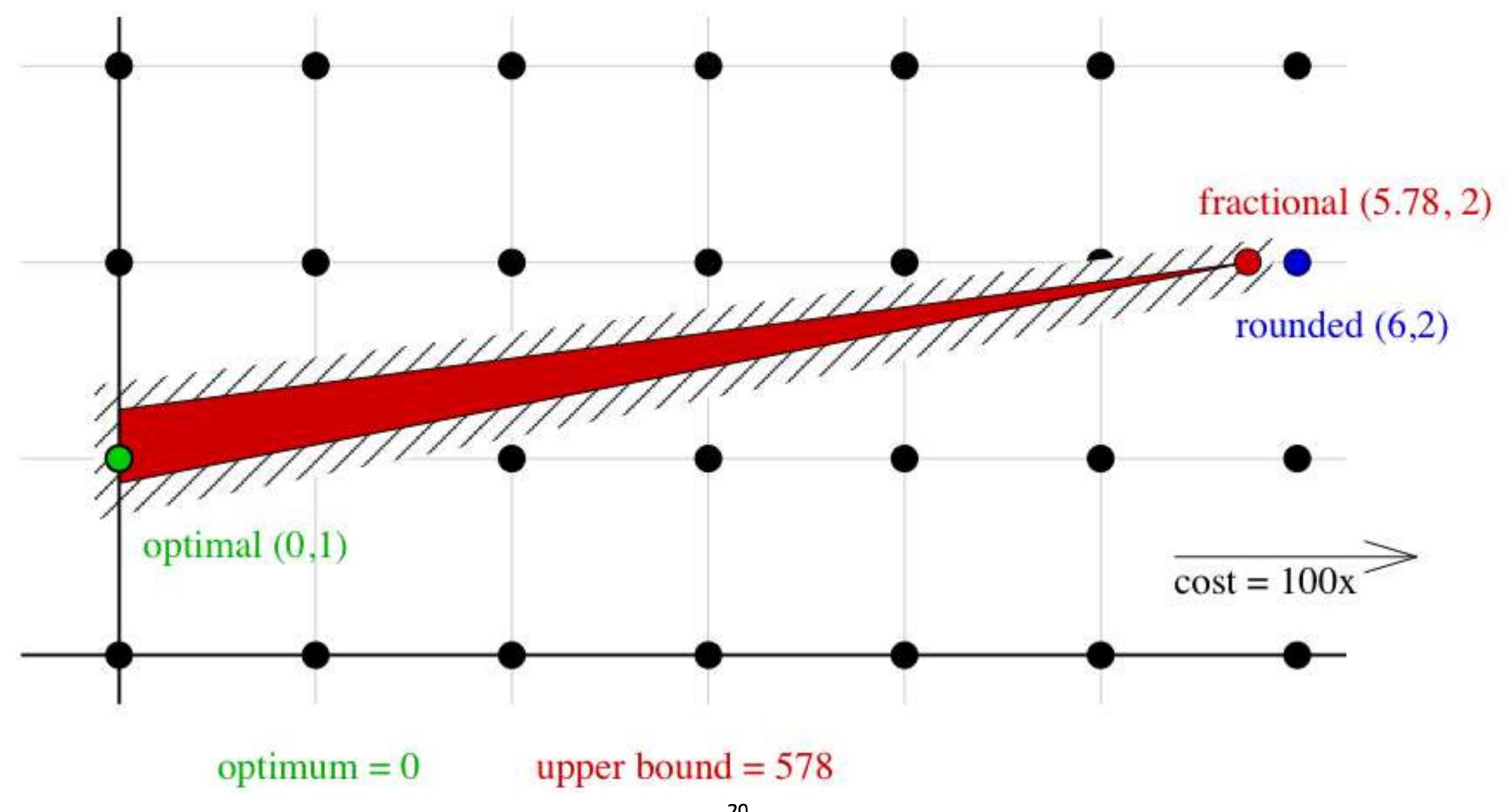
MIPLIB markshare_5_0

```
50 %
                        50 %
         [sofdem:~/Documents/Code/gurobi]$ gurobi.sh mymip.py markshare_5_0.mps.gz
         Changed value of parameter Presolve to 0
            Prev: -1 Min: -1 Max: 2 Default: -1
         Optimize a model with 5 rows, 45 columns and 203 nonzeros
         Found heuristic solution: objective 5335
         Variable types: 5 continuous, 40 integer (40 binary)
         Root relaxation: objective 0.0000 0e+00, 15 iterations 0.00 econds
                          Current Node
             Nodes
                                                Objective Bounds
                                                                            Work
                                                                        It/Node Time
                        Obj Depth IntInf | Incumbent
                                                         BestBd
          Expl Unexpl |
                                                                  Gap |
                         0.00000
                                         5 5335.00000
                                                        0.00000
                                                                  100%
                                                                                0s
                                                                            2.1 1241s
        *62706364 28044
                                     38
                                              1.0000000
                                                           0.00000
                                                                     100%
         Explored 233848403 nodes (460515864 simplex iterations) in 3883.56 seconds
         Thread count was 4 (of 4 available processors)
        Optimal solution found (tolerance 1.00e-04)
         Best objective 1.00000 0000000e+00, best bound 1.000000000000e+00, gap 0.0%
         Optimal objective 1
```

ILP ≠ LP relaxation



ILP ≠ round LP relaxation



general ILP is NP-hard

small problems are easy some specific problems are easy



min
$$s_{n+1}$$
 = p₁+...+p_n
s.t. $s_{n+1} \ge s_j + p_j$

$$s_j - s_i \ge Mx_{ij} + (p_i - M)$$

$$x_{ij} + x_{ji} = 1$$

$$s_j \in \mathbb{Z}_+ \ge 0$$

$$x_{ij} \in \{0, 1\}$$

$$j = 1..n$$

 $i, j = 1..n$
 $i, j = 1..n; i < j$
 $j = 1..n + 1$
 $i, j = 1..n$

1 | Cmax Scheduling Problem

Input n tasks, duration pi
for each task i, 1 machine
Output a minimal makespan
schedule of the tasks on the
machine without overlap



$\min \sum_{(i,j)\in A} c_{ij} x_{ij}$

s.t.
$$\sum_{j \in \delta^+(i)} x_{ij} - \sum_{j \in \delta^-(i)} x_{ij} = b_i \qquad i \in V$$

$$x_{ij} \le h_{ij} \tag{i,j} \in A$$

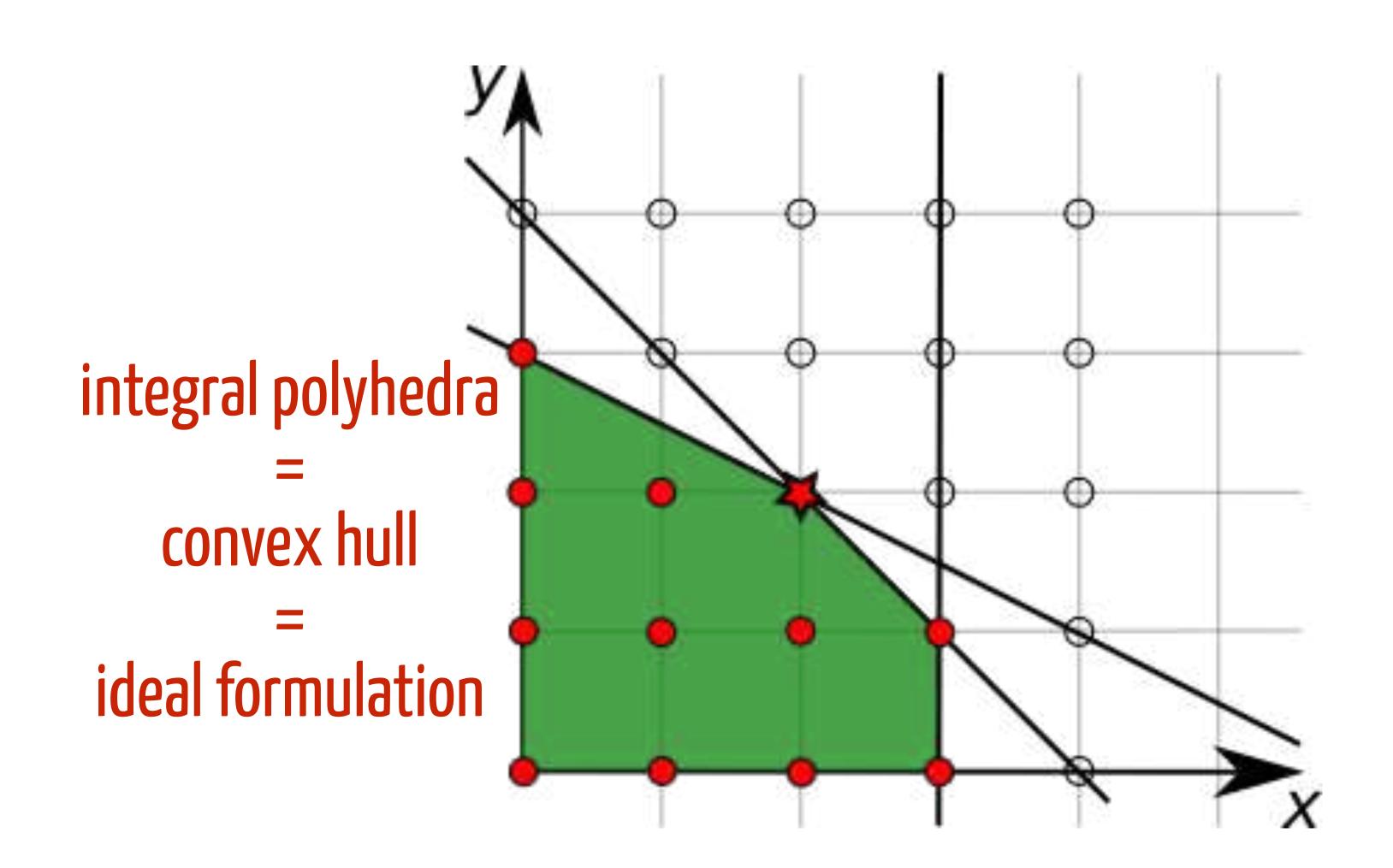
$$x_{ij} \in \mathbb{Z}_+ \ge 0 \tag{i,j} \in A$$

Capacitated Transhipment Problem

Input digraph (V,A), demand
or supply bi at each node i,
capacity hij and unit flow
cost cij for each arc (i,j)
Output a mimimum cost integer
flow to satisfy the demand

Xij flow on arc (i,j)

LP = ILP sometimes



totally unimodular matrix (theory)

$$(P) = \max\{ cx \mid Ax \le b, x \in \mathbb{Z}_+^n \}$$

- basic feasible solutions of the LP relaxation (\bar{P}) take the form: $\bar{x} = (\bar{x}_B, \bar{x}_N) = (B^{-1}b, 0)$ where B is a square submatrix of (A, I_m)
- Cramer's rule: $B^{-1} = B^*/det(B)$ where B^* is the adjoint matrix (made of products of terms of B)
- Proposition: if (P) has integral data (A,b) and if $det(B)=\pm 1$ then \bar{x} is integral

Definition

A matrix A is totally unimodular (TU) if every square submatrix has determinant +1, -1 or 0.

Proposition

If A is TU and b is integral then any optimal solution of (\bar{P}) is integral.

totally unimodular matrix (practice)

How to recognize TU?

Sufficient condition

A matrix A is TU if

- \blacksquare all the coefficients are +1, -1 or 0
- each column contains at most 2 non-zero coefficient
- there exists a partition (M_1, M_2) of the set M of rows such that each column j containing two non zero coefficients satisfies $\sum_{i \in M_1} a_{ij} \sum_{i \in M_2} a_{ij} = 0.$

Proposition

A is TU \iff A^t is TU \iff (A,I_m) is TU where A^t is the transpose matrix, I_m the identity matrix

Interlude

Show that the Transhipment ILP is ideal Show that the Scheduling ILP is NOT ideal