
Solving strategies for the optimisation of energy systems

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Degrees of freedom of a system

DOF : Degree of freedom of the system

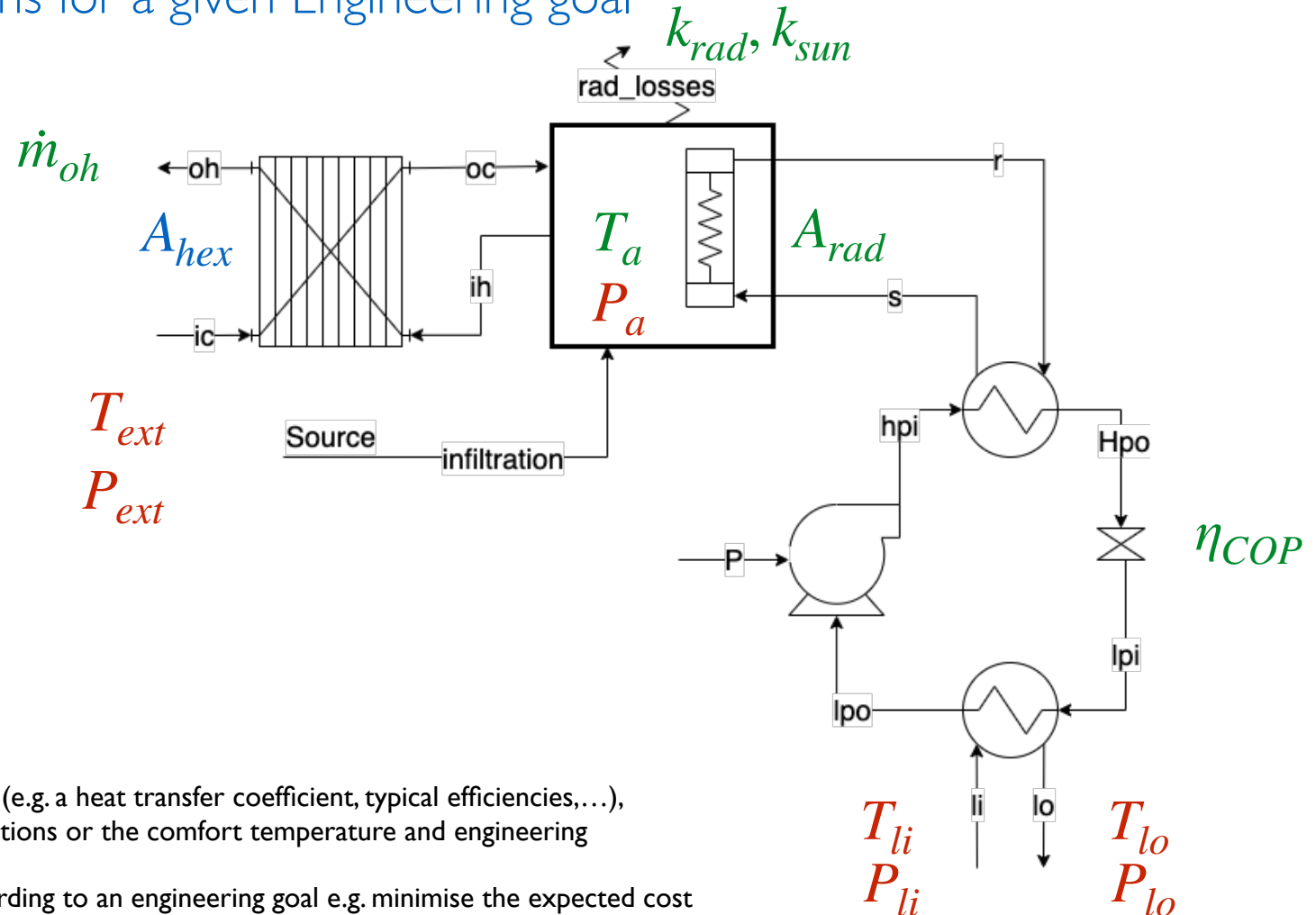
$$\text{DOF of the system} = N_{\{\text{state variables}\}} - N_{\{\text{Model equations}\}}$$

- **State variables** needed to characterise the state
- **Model** define the thermodynamic knowledge
- **DOF** are fixed by specifications: $X - X^* = 0$ or $X = X^*$:
 - Knowledge
 - e.g. parameters of a heat transfer film coefficient correlation
 - Context Specifications
 - e.g. the ambient or the comfort temperature set point
 - Engineering Decisions for a given Engineering goal
 - e.g. minimum expected cost

Heat recovery by a double flux heat exchanger

- Specifications

- Knowledge
- Context Specifications
- Engineering Decisions for a given Engineering goal



Degrees of freedom are fixed by knowledge (e.g. a heat transfer coefficient, typical efficiencies,...), context specifications like the ambient conditions or the comfort temperature and engineering decisions (like the heat exchanger area)-
 Engineering decisions need to be taken according to an engineering goal e.g. minimise the expected cost

Thermo-economic Model

System model	$F(X_{state}, \pi) = 0$	Model equations
	$S(X_{state}, \pi) = 0$	Context specifications
<i>with</i>	π	Model parameters
	X_{state}	State variables

DOF : Degree of freedom of the system

DOF of the system = $N_{\{\text{state variables}\}} - N_{\{\text{Model equations}\}}$

$$N_{\{\text{Decision variables}\}} = \text{DOF} - N_{\{\text{Context}\}} - N_{\{\text{Knowledge}\}}$$

Decision variables :

- Degrees of freedom for which I do not have any rationale to fix its value
- I only know the possible range

$$X_{d,min} \leq X_d \leq X_{d,max}$$

Thermo-economic optimisation

Find the value of the decision variables X_d with a given objective

$$X_{d,min} \leq X_d \leq X_{d,max}$$

“I would like to fix the value of X_d so that it minimise an objective function”

$$\min_{X_{state}} \quad TotalCost(X_{state}, \pi) \quad \rightarrow \text{Objective function}$$

$$s.t. \quad F(X_{state}, \pi) = 0 \Rightarrow \text{equipment model}$$

$$S(X_{state}, \pi) = 0 \Rightarrow \text{Specification equations} \quad \rightarrow \text{Constraints}$$

$$G(X_{state}, \pi) \leq 0 \quad \text{Inequality constraints}$$

$$X_{state}^{min} \leq X_{state} \leq X_{state}^{max} \quad \text{Bounds}$$

where

$$X_{state} = \{x_{statevariables}, x_{UnitParameters}, y_{decision} \in \{0, 1\}\}$$

$$X_d \text{ is a subset of } X_{state}$$

Typical objective function formula

An objective function defines the relative importance of flows and installed capacity of equipments during the expected lifetime of the system

$$OBJ_o = \sum_s \sum_f^{states\ flows} (v_{f,s,o} \cdot t_{op,s} \cdot \dot{m}_{f,s}) + \sum_u^{units} v_{u,o} \cdot \frac{1}{\tau_u} \cdot O_u(S_u)$$

- with

- $\dot{m}_{f,s}$: the flow [kg/s or kW] of resource or product f in the state of operation s
 - $v_{f,s,o}$: the value (relative importance) of $\dot{m}_{f,s}$ for the objective function o
 - $v_{f,s,o}$ can be the expected cost of a flow (e.g. [CHF/kg or CHF/MJ] or the Life Cycle Impact of the flow
 - $t_{op,s}$: a time of operation of the system in the state s of flow \dot{m}_s in [s/year]
 - $v_{u,o}$: the value (relative importance) of the capacity function $O_u(S_u)$ of unit u with the size S_u for the objective function o
 - $\frac{1}{\tau_u}$: annualisation of the installed capacity of unit u over its expected lifetime
 - $O_u(S_u)$ can be the investment cost in [CHF], materials, Life Cycle Inventory e.g. in [tonCO_{2eq}], Land footprint, Life cycle impact associated to construction of unit u
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Example of objective functions : engineering goals

$$TotalCost[CHF/year] = OPEX + CAPEX + Tax$$

$$OPEX = \sum_{p=1}^{n_p} \left(\sum_{r=1}^{n_r} \dot{m}_{r,p}^+ c_{r,p}^+ + \dot{E}_p^+ c_{e,p}^+ - \dot{E}_p^- c_{e,p}^- + \sum_{u=1}^{n_u} f_{u,p} c_{m_u} \right) d_p$$

$$CAPEX = \sum_{u=1}^{n_u} \frac{1}{\tau(n_{y,u}, i)} (I1_u y_u + I2_u f_u^{max})$$

$$Tax = CO_2^+ \gamma^{CO_2^+}$$

$$CO_2^+ = \sum_{p=1}^{n_p} \left(\sum_{r=1}^{n_r} \dot{m}_{r,p}^+ \epsilon_r^{CO_2} + \dot{E}_p^+ \epsilon_{e,p}^{CO_2^+} - \dot{E}_p^- \epsilon_{e,p}^{CO_2^-} \right) d_p$$

$$Impact = \zeta^{CO_2^+} (CO_2^+ + \sum_{u=1}^{n_u} \frac{1}{n_{y,u}} (\xi_{c_u}^{CO_2} + \xi_{d_u}^{CO_2}) f_u^{max})$$

$$RES = \sum_{p=1}^{n_p} \left(\sum_{r_{res}=1}^{n_{r_{res}}} \dot{m}_{r_{res},p}^+ + \sum_{u=1}^{n_u} f_{u,p} e_{u,p}^{res+} \right) d_p$$

$$\dot{E}_p^+ + \dot{E}_p^- + \sum_{u=1}^{n_u} f_{u,p} (e_{u,p}^{res+} - e_{u,p}^-) = 0 \quad \forall p = 1..n_p$$

Inequality constraints

Constraints limit the search space for the values of X_d

- **Operating limits**
 - safety, equipment limits,
- **Regulation**
 - Emission : environment
- **Market limits => products quality**
- **Technology limits and heuristics**
 - Materials
- **Models limits**
 - Correlations
 - Structures and models
- **Numerical safety ; Two types**
 - **Hard**: numerical validity of the model: i.e. can not be violated otherwise the system state can not be calculated
 - E.g. flow inversion
 - try to place the hard constraints on the variables
 - **Soft**: may be violated during resolution. The state exists but is not expected to be valid as a solution

Different types of problems

- **Optimal operation (Equipment Size is fixed)**
 - Decision : Operation set point
 - Objective : Operating cost (CHF/s)
- **Optimal operation strategy (Equipment Size is fixed)**
 - Decision : Set point strategy + start/stop
 - Objective : Operating cost (CHF/period)
- **Scheduling (Equipment Size is fixed)**
 - Decision : Set points strategy, start/stop, when ?, in which equipment ?
 - Objective : Operating cost (CHF/period)

Different types of problems

- **Optimal sizing (equipment size is decided)**
 - Decision : operating set points, unit sizes, investment
 - Objective : Total cost (CHF, CHF/an)
- **Optimal design (choice and size of equipment is decided)**
 - Decision : operating set points (Operating strategy), unit sizes, investment, configuration
 - Objective : Total cost (CHF, CHF/an)
- **Retrofit (choice and size of some of the equipment is decided, reuse of existing equipment with fixed sizes / configuration)**
 - Decision : operating set points (Operating strategy), unit sizes, configuration, reuse, reconfiguration, investments
 - Objective : total cost

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- How do I use a flowsheet model to solve an optimization problem ?

$F(X_{state}) = 0 \Rightarrow$ equipment model

$L(X_{state}) = 0 \Rightarrow$ linking equations

$T(X_{state}) = 0 \Rightarrow$ constitutive equations

$S(X_{state}) = 0 \Rightarrow$ Specification equations

Black box method

Optimisation : $\min \text{OBJ}(X^* \text{ decision})$
Subject to $G_{\text{inequality}}(X^* \text{ decision}) \geq 0$

$X^* \text{ decision}$

$\text{OBJ}(X^* \text{ decision})$
 $G(X^* \text{ decision}) \text{ inequality}$
Status

Model : Solve

$F(X_{\text{dependent}}, X_{\text{specification}}, X_{\text{decision}}) = 0$

$S(X_{\text{dependent}}, X_{\text{specification}}, X_{\text{decision}}) = 0 \Rightarrow X(X^* \text{ decision})$

$X_{\text{decision}} - X^* \text{ decision} = 0$

then calculate $\text{OBJ}(X(X^* \text{ decision}))$

$G(X(X^* \text{ decision}))$

Black Box strategy

$$\min_{X_{decision}^*}$$

$$TotalCost(X_{decision}^*, X(X_{decision}^*), \pi)$$

$$s.t. \quad G(X_{decision}^*, X(X_{decision}^*), \pi) \leq 0 \quad \text{inequality constraints}$$

where

$$X_{decision}^* = \{x_{decision}, y_{decision} \in \{0, 1\}\}$$

$$X(X_{decision}^*)$$

Calculated by solving:

**NxN
problem**

$$\left\{ \begin{array}{l} F(X_{state}, \pi) = 0 \Rightarrow \text{model equations} \\ S(X_{state}, \pi) = 0 \Rightarrow \text{Specification equations} \\ X_{state} - X_{decision}^* = 0 \Rightarrow \text{Specification of the value of decision variables} \end{array} \right.$$

where

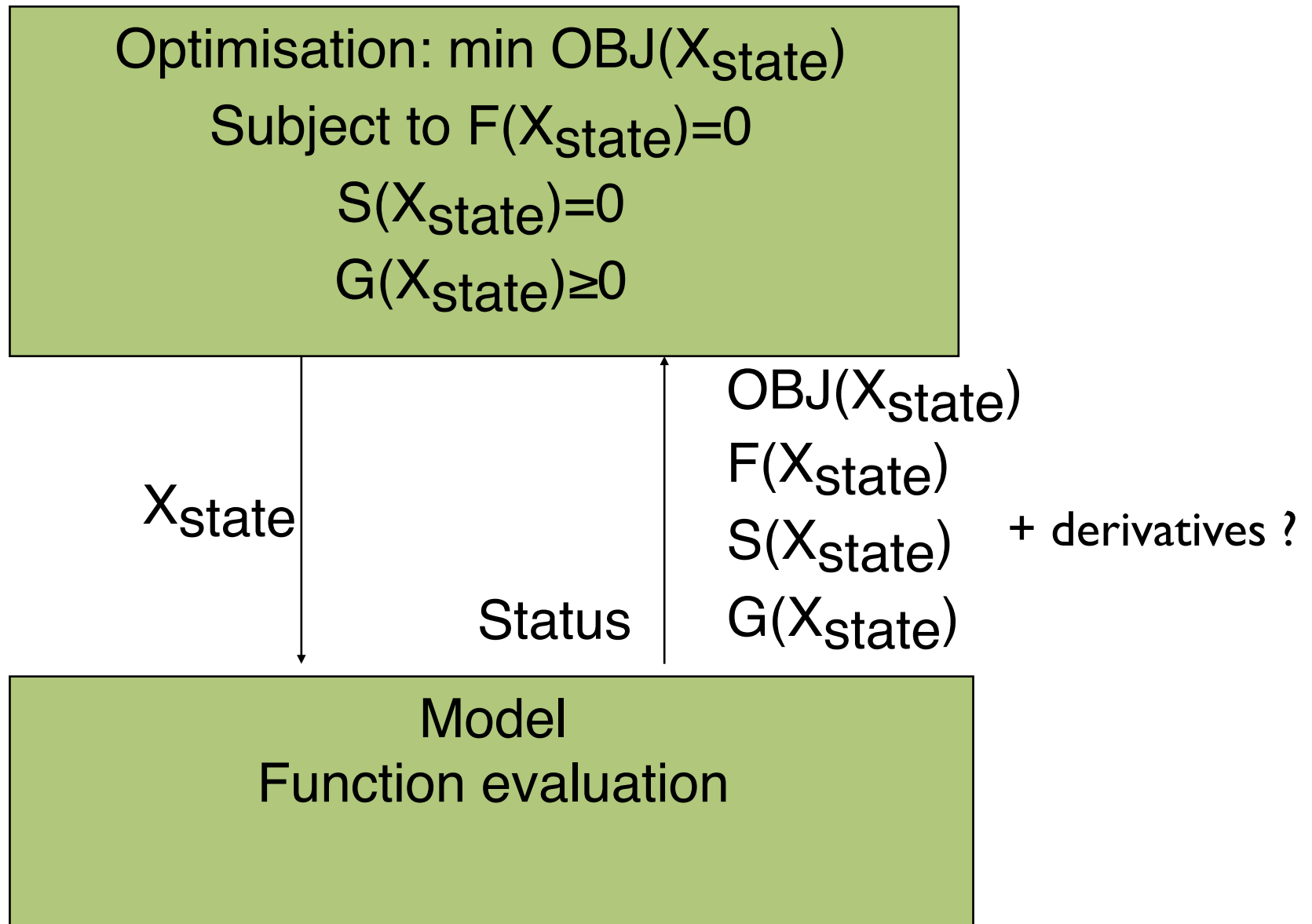
$$X_{state} = \{x_{StateVariables}, x_{UnitParameters}, y_{decision} \in \{0, 1\}\}$$

Black Box approach

- + Can be used in any kind of optimisation method
 - Heuristic, direct et indirect (derivatives)
- + Solving method to be defined => problem analysis
- + Selection of X^* _{decision} by the user
- + Robustness
- + Non convergence analysis
- + Gives a list of system states
- Flexibility
- Computation time
- Inequality might be a problem when not on decision variables
- Derivative calculation (especially when done in iterative loops)

$$\frac{\partial f}{\partial x_i} = \frac{f(X + \Delta x_i) - f(X)}{\Delta x_i}$$

Simultaneous method



Simultaneous strategy

$$\begin{aligned} \min_{X_{state}} \quad & TotalCost(X_{state}, \pi) \\ \text{s.t.} \quad & F(X_{state}, \pi) = 0 \Rightarrow \text{equipment model} \\ & S(X_{state}, \pi) = 0 \Rightarrow \text{Specification equations} \\ & G(X_{state}, \pi) \leq 0 \quad \text{Inequality constraints} \\ & X_{state}^{min} \leq X_{state} \leq X_{state}^{max} \quad \text{Bounds} \end{aligned}$$

where

$$X_{state} = \{x_{statevariables}, x_{UnitParameters}, y_{decision} \in \{0, 1\}\}$$

Requires : Non Linear Constrained Optimisation solving method

X_d is not defined (hidden in X_{state})

Simultaneous approach

- + Optimisation under constraints
 - => indirect method with 2nd derivatives if possible
- + Flexibility : different pbm with the same model (changing specification set and not the model)
- + Efficient and robust for on-line optimisation systems
 - Bounds definition !
- + Computing time
- + Automatic DOF analysis and sensitivity analysis
- + Hard and soft inequality constraints
- Initialisation !
- Derivatives calculation (during function evaluation, symbolic ?)
- No system state when no convergence
- Scaling !
- No explanations when the problem does not converge !
- Push button system?

Two levels approach

Optimisation:

$$\min \text{OBJ}(X^* \text{ decision}, X^* \text{ specification})$$

Subject to $H(X^* \text{ decision}, X^* \text{ specification})=0$

$$G(X^* \text{ decision}, X^* \text{ specification}) \geq 0$$

Requires a
constrained
optimisation solver

$X^* \text{ decision},$
 $X^* \text{ specification}$

Status

$\text{OBJ}(X^* \text{ decision}, X^* \text{ specification})$
 $H(X^* \text{ decision}, X^* \text{ specification})$
 $G(X^* \text{ decision}, X^* \text{ specification})$
+ derivatives ?

Modèle

$$X_{\text{dependent}} = \Phi(X^* \text{ decision}, X^* \text{ specification})$$

Two level strategy

$$\begin{aligned} \min_{X_{decision}^*, X_{Specs}^*} \quad & TotalCost(X_{decision}^*, X_{Specs}^*, X(X_{decision}^*, X_{Specs}^*), \pi) \\ \text{s.t.} \quad & H(X_{decision}^*, X_{Specs}^*, X(X_{decision}^*, X_{Specs}^*), \pi) = 0 \quad \text{some equality constraints} \\ & G(X_{decision}^*, X_{Specs}^*, X(X_{decision}^*, X_{Specs}^*), \pi) \leq 0 \quad \text{inequality constraints} \end{aligned}$$

where

$$X_{decision}^* = \{x_{decision}, y_{decision} \in \{0, 1\}\}$$

$(X_{decision}^*, X_{Specs}^*)$

Calculated by solving :

**(N+Ns)x(N+Ns)
problem**

$$\left\{ \begin{aligned} F(X_{state}, \pi) = 0 &\Rightarrow \text{system model} \\ S(X_{state}, \pi) = 0 &\Rightarrow \text{Specification equations} \\ X_{decision} - X_{decision}^* = 0 &\Rightarrow \text{Specification of the value of decision variables} \\ X_{Specs} - X_{Specs}^* = 0 &\Rightarrow \text{Specification of the value of specification variables} \end{aligned} \right.$$

where

$$X_{state} = \{x_{StateVariables}, x_{Unitparameters}, y_{decision} \in \{0, 1\}\}$$

Two levels strategy

- + Best of both world
- + Robustness of unit models (single calculation mode)
- + Code Maintenance
- + Derivative chaining is possible
 - Analytical calculation is possible
- + Soft inequality constraints
- + Sensitivity analysis
- Conditional simulation
- Computing time (solving the lower level iteratively)
- Hard bounds (except when they are at the decision variable level).
- Noise in derivatives evaluation
 - Internal iterative calculations