Solving strategies for the optimisation of energy systems

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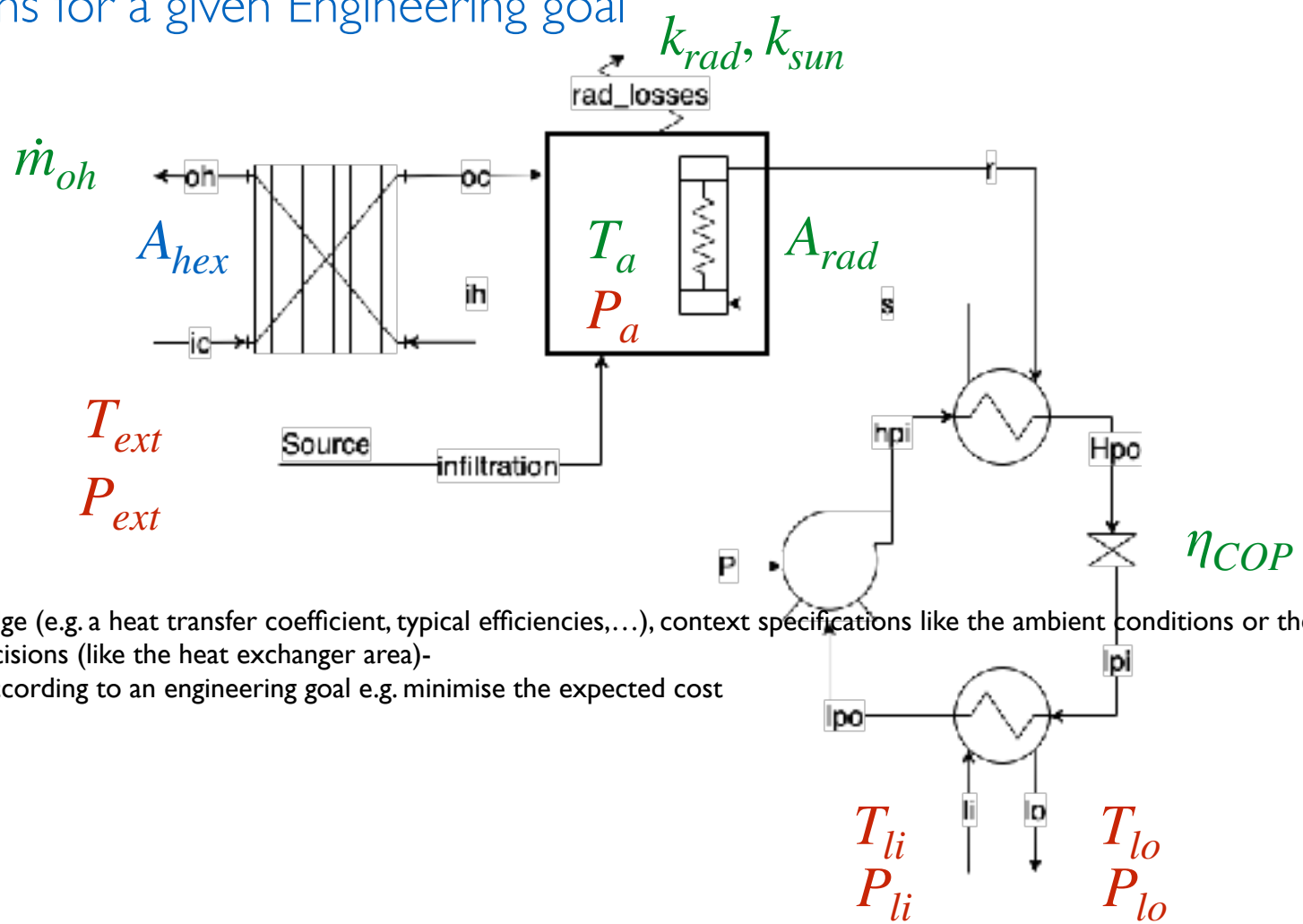
Heat recovery by a double flux heat exchanger

- Degrees of freedom ?
  - 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

$$F(X_{state}, \pi) = 0 \quad \text{Model equations}$$
$$S(X_{state}, \pi) = 0 \quad \text{Context specifications}$$
$$with \quad \pi \quad \text{Model parameters}$$
$$X_{state} \quad \text{State variables}$$

DOF : Degree of freedom of the system

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

Decision variables

N_{Decision variables} = DOF - N_{Context specifications}

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

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

Thermo-economic optimisation

Fixing the value of the decision variables

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

- Degree of freedom for which I do not have any rationale to fix the value
- I only know the possible range
- I would like to fix the value of X_d so that it minimise an objective function

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

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

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

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

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

$$where$$

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

X_d 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

$$OBJ_o = \sum_{s}^{states} (v_{s,o} \cdot t_{op,s} \cdot \dot{m}_s) + \sum_{u}^{units} v_{u,o} \cdot \frac{1}{\tau_u} \cdot O_u(S_u)$$

- with
  - $v_{s,o}$: the value (relative importance) of $\dot{m}_s$ for the objective function o
    - $v_{s,o}$ can be the expected cost of a flow or the Life Cycle Impact of the flow
  - $t_{op,s}$: a time of operation of the system in the state s of flow $\dot{m}_s$
  - $v_{u,o}$: 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 lifetime
    - $O_u(S_u)$ can be the investment cost, materials, Life Cycle Inventory, Land footprint, Life cycle impact associated to construction of unit u

Example of objective functions : engineering goals

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

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

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

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

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

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

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

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

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)

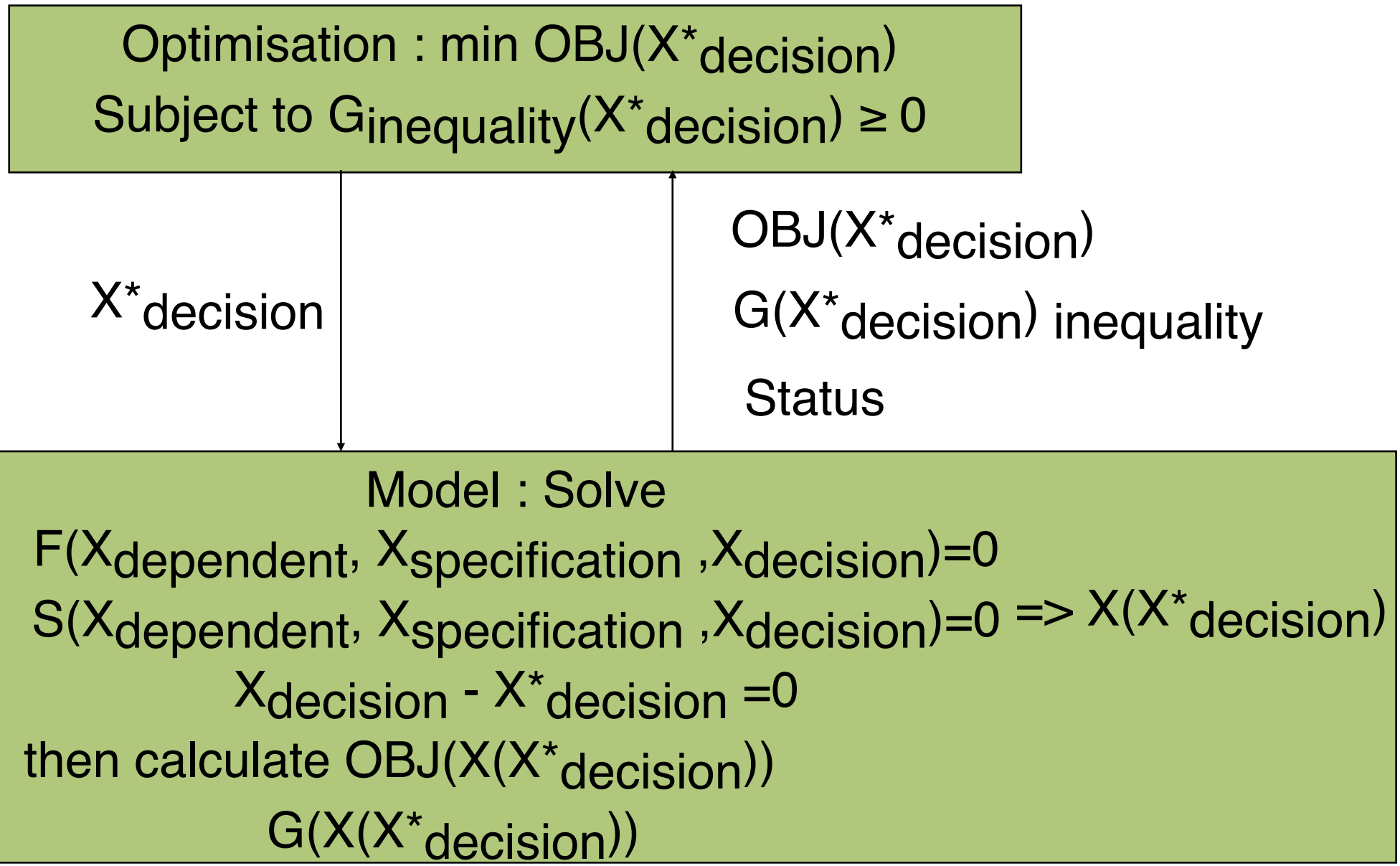
Different types of problems

- **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

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

$$F(X_{state}) = 0 \Rightarrow \text{equipment model}$$
$$L(X_{state}) = 0 \Rightarrow \text{linking equations}$$
$$T(X_{state}) = 0 \Rightarrow \text{constitutive equations}$$
$$S(X_{state}) = 0 \Rightarrow \text{Specification equations}$$

Black box method



Black Box strategy

$$\min_{X^*_{decision}} \quad TotalCost(X^*_{decision}, X(X^*_{decision}), \pi)$$

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

$$where$$

$$X^*_{decision} = \{x_{decision}, y_{decision} \epsilon \{0,1\}\}$$

$$X(X^*_{decision}) \quad \text{Calculated by solving:}$$

NxN problem

$$\begin{cases} F(X_{state}, \pi) = 0 \Rightarrow \text{model equations} \\ S(X_{state}, \pi) = 0 \Rightarrow \text{Specification equations} \\ X_{state} - X^*_{decision} = 0 \Rightarrow \text{Specification of the value of decision variab} \end{cases}$$

$$where$$

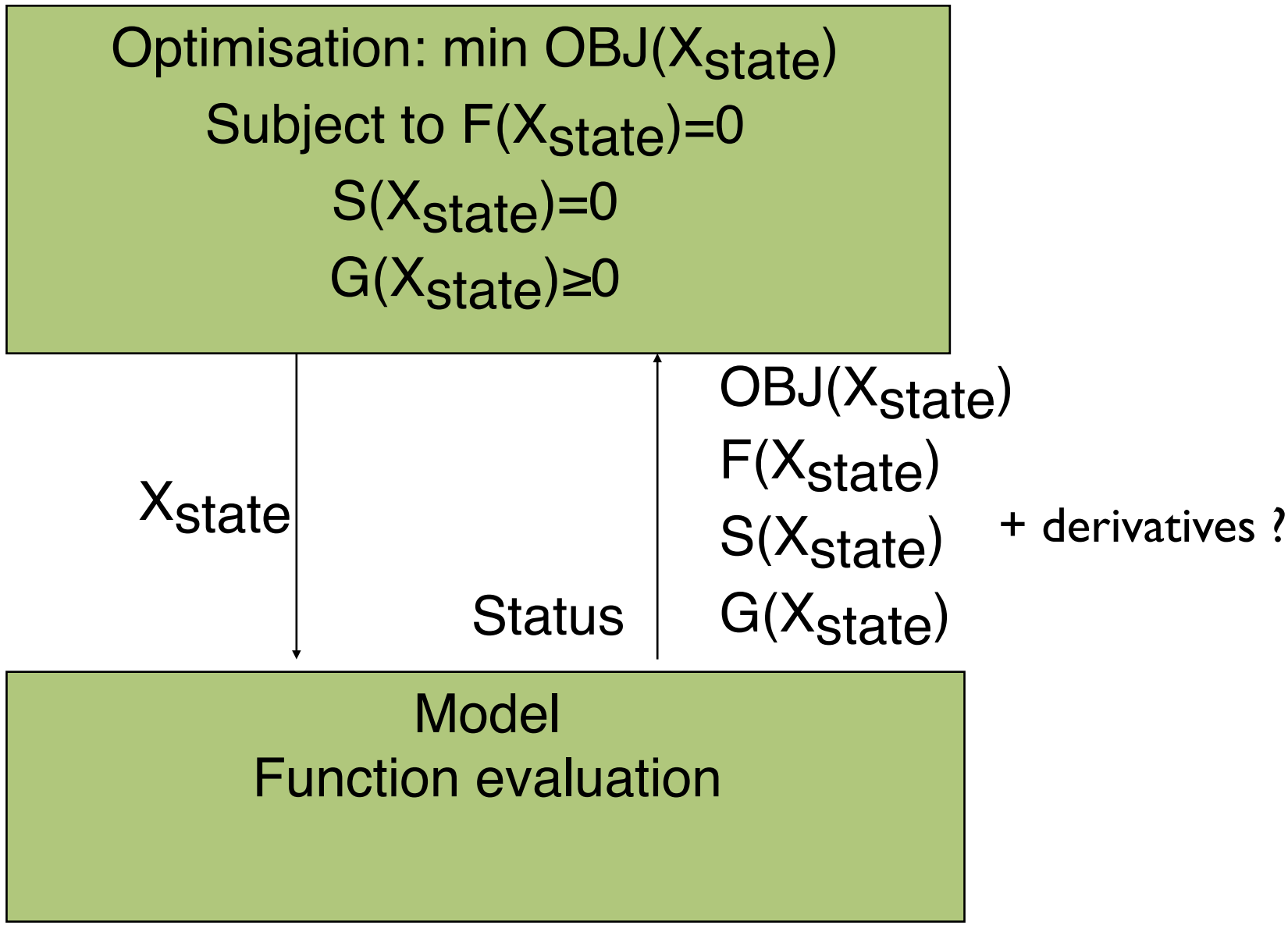
$$X_{state} = \{x_{StateVariables}, x_{UnitParameters}, y_{decision} \epsilon \{0,1\}\}$$

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

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

where

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

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 ?)

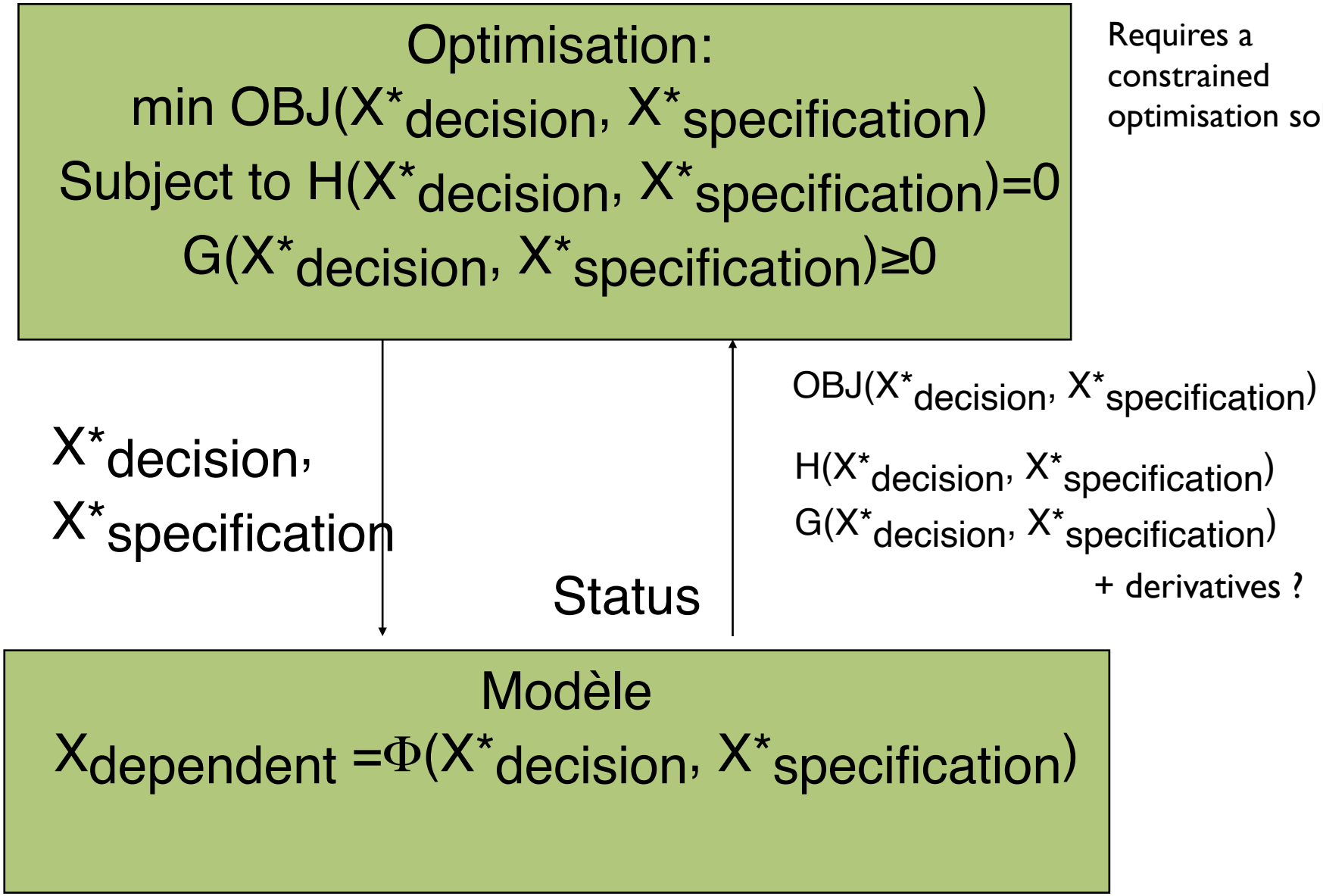
- No system state when no convergence

- Scaling !

- No explanations when the problem does not converge !

- Push button system?

Two levels approach



Two level strategy

$$\min_{X^*_{decision}, X^*_{Specs}} \quad TotalCost(X^*_{decision}, X^*_{Specs}, X(X^*_{decision}, X^*_{Specs}), \pi)$$

$$s.t. \quad H(X^*_{decision}, X^*_{Specs}, X(X^*_{decision}, X^*_{Specs}), \pi) = 0 \quad \text{some equality constraints}$$

$$G(X^*_{decision}, X^*_{Specs}, X(X^*_{decision}, X^*_{Specs}), \pi) \leq 0 \quad \text{inequality constraints}$$

where

$$X^*_{decision} = \{x_{decision}, y_{decision} \epsilon \{0,1\}\}$$

$(X^*_{decision}, X^*_{specs})$ Calculated by solving :

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

$$F(X_{state}, \pi) = 0 \Rightarrow \text{system model}$$

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

$$X_{decision} - X^*_{decision} = 0 \Rightarrow \text{Specification of the value of decision variables}$$

$$X_{specs} - X^*_{specs} = 0 \Rightarrow \text{Specification of the value of specification variables}$$

where

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

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