Graph-Based Optimization Modeling Language (GBOML)

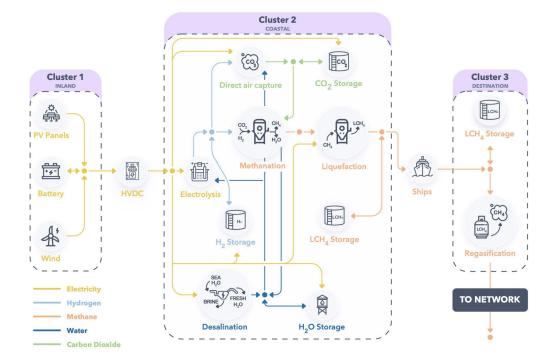
TotalEnergies

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Many energy system planning and control problems can be formulated as mathematical programs

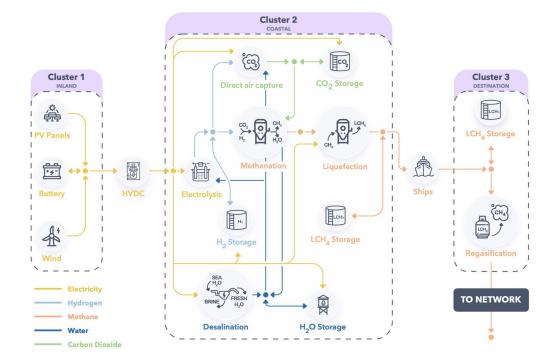




For example, we may want to *design* and *operate* a system producing synthetic fuel so as to minimise its cost

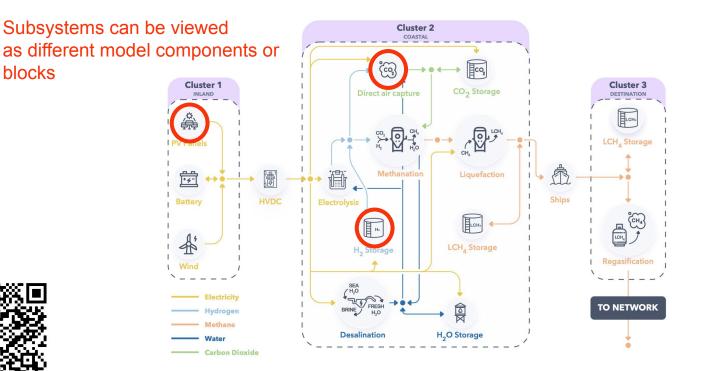
min [investment costs] + [operating costs]
s.t. [system design constraints]
[operating constraints of subsystems]
[coupling constraints between subsystems]
[regulatory and environmental constraints]

Mathematical programs found in energy system planning and control applications have special structure





Notably, the underlying system is often a collection of subsystems whose operation must be optimized over time





This structure can be represented via a hypergraph abstraction augmented with some concept of time-indexing

$$\begin{array}{ll} \min & \sum_{n \in \mathcal{N}} \left[f_0^n(X^n, Z^n) + \sum_{t \in \mathcal{T}} f^n(X^n, Z^n, t) \right] \\ \text{s.t.} & h_k^n(X^n, Z^n, t) = 0, \; \forall t \in \mathcal{T}_k^n, \; k = 1, \dots K^n, \; \forall n \in \mathcal{N} \\ & g_k^n(X^n, Z^n, t) \leq 0, \; \forall t \in \bar{\mathcal{T}}_k^n, \; k = 1, \dots \bar{K}^n, \; \forall n \in \mathcal{N} \\ & H^e(Z^e) = 0, \; \forall e \in \mathcal{E} \\ & G^e(Z^e) \leq 0, \; \forall e \in \mathcal{E} \\ & X^n \in \mathcal{X}^n, Z^n \in \mathcal{Z}^n, \; \forall n \in \mathcal{N}. \end{array}$$

Sets of *nodes*, *hyperedges* and *time periods* define the hypergraph and time-indexed structure

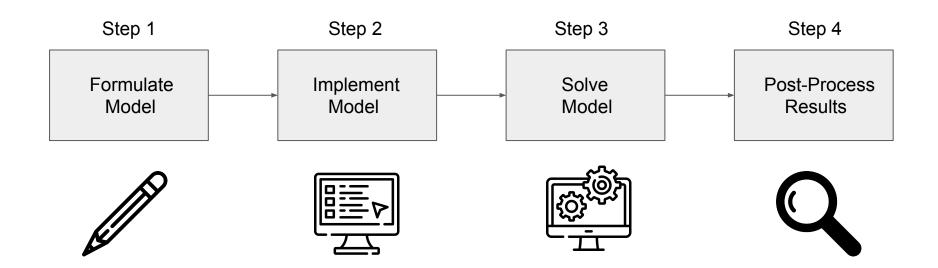
set of nodes set of time periods min $\sum_{n \in \mathcal{N}} \left[f_0^n(X^n, Z^n) + \sum_{t \in \mathcal{T}} f^n(X^n, Z^n, t) \right]$ s.t. $h_k^n(X^n, Z^n, t) = 0, \ \forall t \in \mathcal{T}_k^n, \ k = 1, \dots, K^n, \ \forall n \in \mathcal{N}$ $g_k^n(X^n, Z^n, t) \leq 0, \ \forall t \in \overline{\mathcal{T}}_k^n, \ k = 1, \dots, \overline{K}^n, \ \forall n \in \mathcal{N}$ $H^e(Z^e) = 0, \ \forall e \in \mathcal{E}$ - set of hyperedges $G^e(Z^e) \leq 0, \forall e \in \mathcal{E}$ $X^n \in \mathcal{X}^n, Z^n \in \mathcal{Z}^n, \forall n \in \mathcal{N}.$

We distinguish between *internal* variables belonging to nodes and *coupling* variables used to link nodes

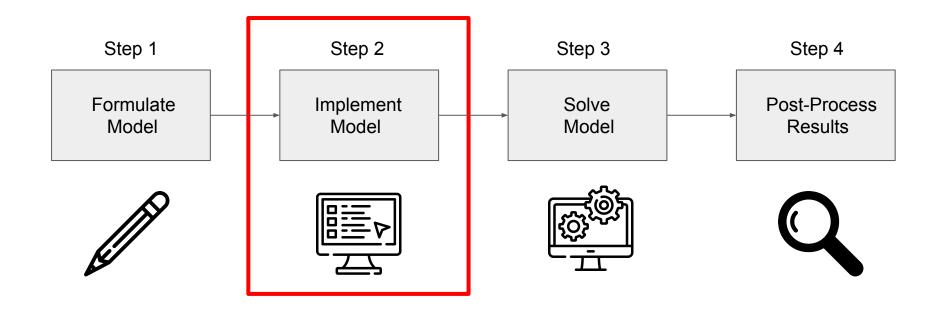
$$\begin{array}{ll} \text{min} & \sum_{n \in \mathcal{N}} \left[f_0^n(X^n, Z^n) + \sum_{t \in \mathcal{T}} f^n(X^n, Z^n, t) \right] \\ \text{s.t.} & h_k^n(X^n, Z^n, t) = 0, \ \forall t \in \mathcal{T}_k^n, \ k = 1, \dots, K^n, \ \forall n \in \mathcal{N} \\ & g_k^n(X^n, Z^n, t) \leq 0, \ \forall t \in \bar{\mathcal{T}}_k^n, \ k = 1, \dots, \bar{K}^n, \ \forall n \in \mathcal{N} \\ & H^e(Z^e) = 0, \ \forall e \in \mathcal{E} \\ & G^e(Z^e) \leq 0, \ \forall e \in \mathcal{Z}^n, \ \forall n \in \mathcal{N}. \end{array}$$

Each node has its own objective and constraints while hyperedges implement coupling constraints node objective min $\sum_{n \in \mathcal{N}} \left| f_0^n(X^n, Z^n) + \sum_{t \in \mathcal{T}} f^n(X^n, Z^n, t) \right|$ s.t. $h_k^n(X^n, Z^n, t) = 0, \ \forall t \in \mathcal{T}_k^n, \ k = 1, \dots, K^n, \ \forall n \in \mathcal{N}$ $g_k^n(X^n, Z^n, t) \le 0, \ \forall t \in \overline{\mathcal{T}}_k^n, \ k = 1, \dots, \overline{K}^n, \ \forall n \in \mathcal{N}$ $H^e(Z^e) = 0, \ \forall e \in \mathcal{E}$ node constraints $G^e(Z^e) \le 0, \ \forall e \in \mathcal{E}$ coupling constraints $X^n \in \mathcal{X}^n, Z^n \in \mathcal{Z}^n, \forall n \in \mathcal{N}.$

Working with optimization models involves at least four basic steps



We will focus on the second step: model encoding and implementation



Two classes of tools are available to implement models

- 1. Algebraic Modeling Languages (AMLs):
 - Formulation close to mathematical notation (e.g., index-based notation)
 - Very **expressive** (e.g., can represent any mixed-integer nonlinear program)
 - Often interface with multiple solvers
 - Sometimes open source
 - Examples :









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- 2. Object-Oriented Modeling Environments (OOMEs):
 - Focus on **one** particular **application** (e.g., generation expansion planning)
 - Usually make use of predefined components that can be "imported"
 - Typically have advanced **data processing** capabilities tailored to application
 - Often open source
 - Examples :

PLEXOS $\triangle PyPSA$ **Dispa-SET** Calliope Power system modelling

Each approach has drawbacks

AMLs typically fail to **expose** or **exploit** block **structure**, although this may be used to:

- simplify model encoding
- enable model re-use
- speed up model generation
- facilitate the use of structure-exploiting algorithms

OOMEs, for their part:

- Lack expressiveness
- Often cumbersome to add new components
- Usually **rely** on **AMLs** and inherit any **shortcomings**

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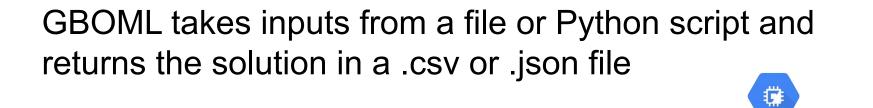
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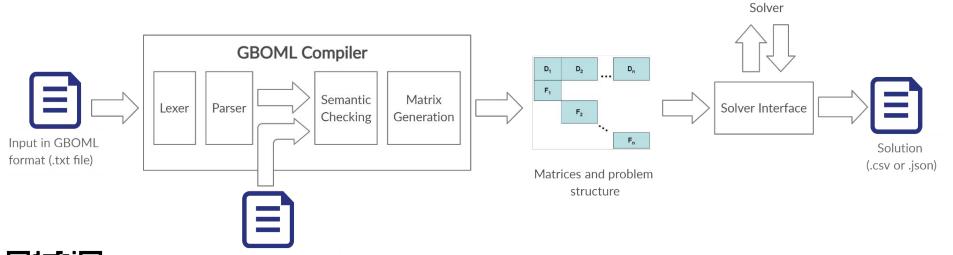
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GBOML is a modeling language with its own parser

- Software developed in Python:
 - Has very **few dependencies** (PLY, NumPy, SciPy)
 - Provides two methods to **encode** models (**file** and Python **API**)
 - Interfaces with **several solvers** (Gurobi, CPLEX, Xpress, Cbc/Clp, HiGHS and DSP)
 - Produces plain **.csv** or structured **.json** outputs
- Model structure is exploited on multiple levels:
 - Model **encoding** via dedicated language constructs
 - Model generation via parallelism and multiprocessing
 - Solving via structure-exploiting solvers such as DSP





Script-based GBOML*



A Jupyter notebook has been prepared to illustrate how GBOML can be used



References

[1] Mathias Berger et al., (2021). "Remote Renewable Hubs for Carbon-Neutral Synthetic Fuel Production", in Frontiers in Energy Research 9, p.200. DOI 10.3389/fenrg.2021.671279. https://www.frontiersin.org/article/10.3389/fenrg.2021.671279

[2] Miftari et al., (2022). GBOML: Graph-Based Optimization Modeling Language. Journal of Open Source Software, 7(72), 4158, <u>https://doi.org/10.21105/joss.04158</u>

[3] Miftari et al., (2022). GBOML repository: https://gitlab.uliege.be/smart_grids/public/gboml

[4] Freepik Icons (used on slides 10-11): https://www.flaticon.com/authors/freepik