

Linking Cluster-Level and National-Level Models to assess chemical industry transition pathways

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November 27th, 2025



Why?

- Chemical industry is a critical part of the transition to net-zero emissions
- Even greater climate impact when considering embedded carbon in chemical products

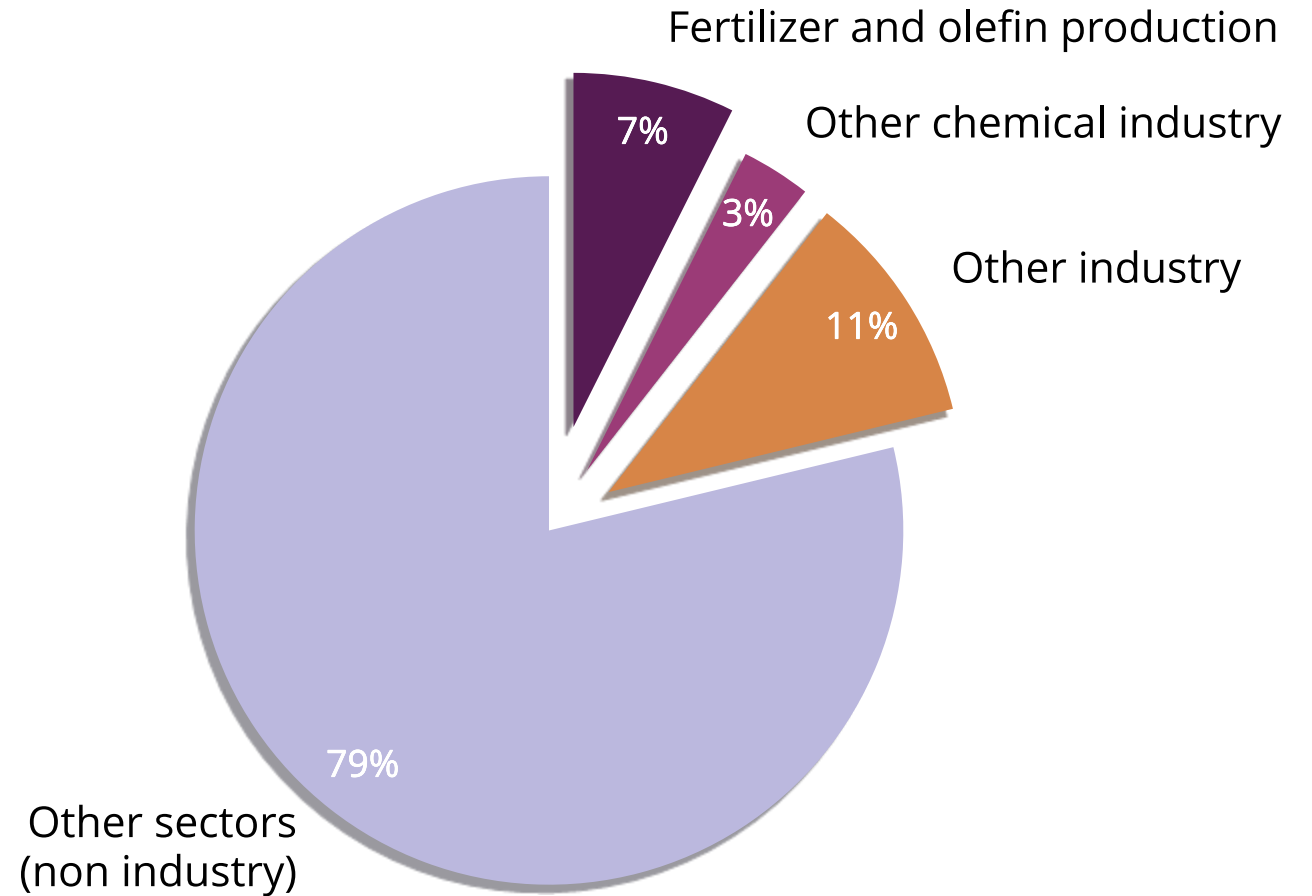


Figure 1: Global GHG emissions by sector in 2023 [1-3].

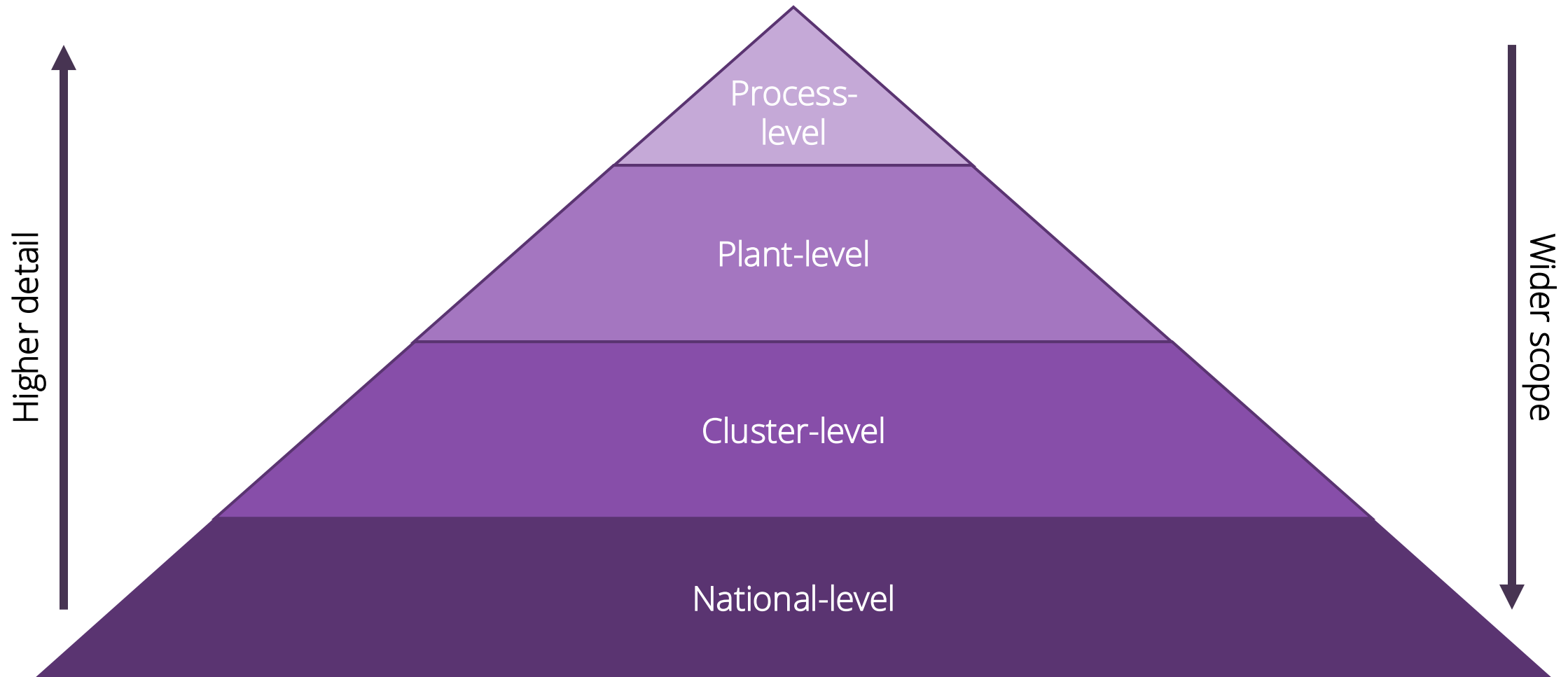
Challenges of chemical industry transformation

- Complex processes
- Plants co-located in industrial clusters
- Central role in global economy
- Need to redesign production routes at the Mt/year scale

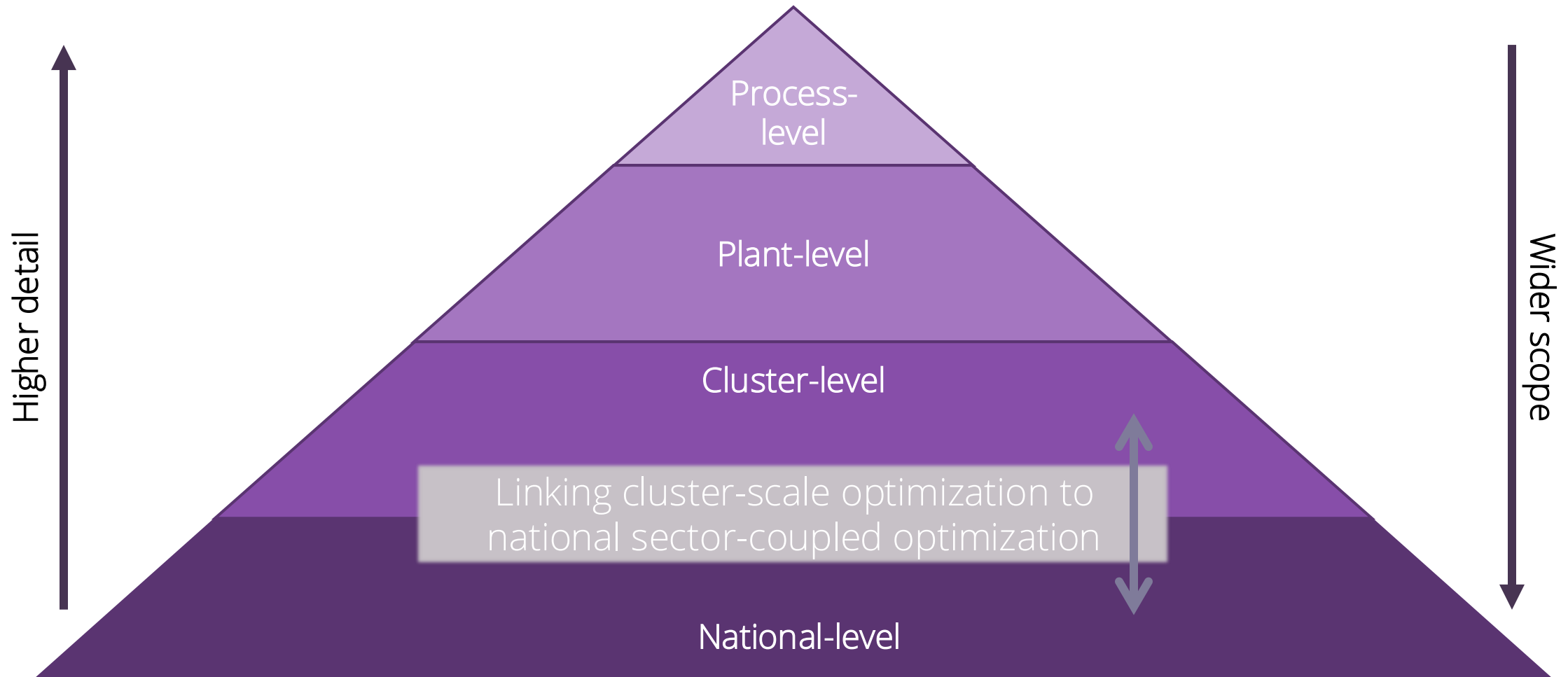
Aim:

To develop analytical tools that can support long-term decision-making by capturing the economic and technical complexity at different scales.

Modeling the transition at different scales



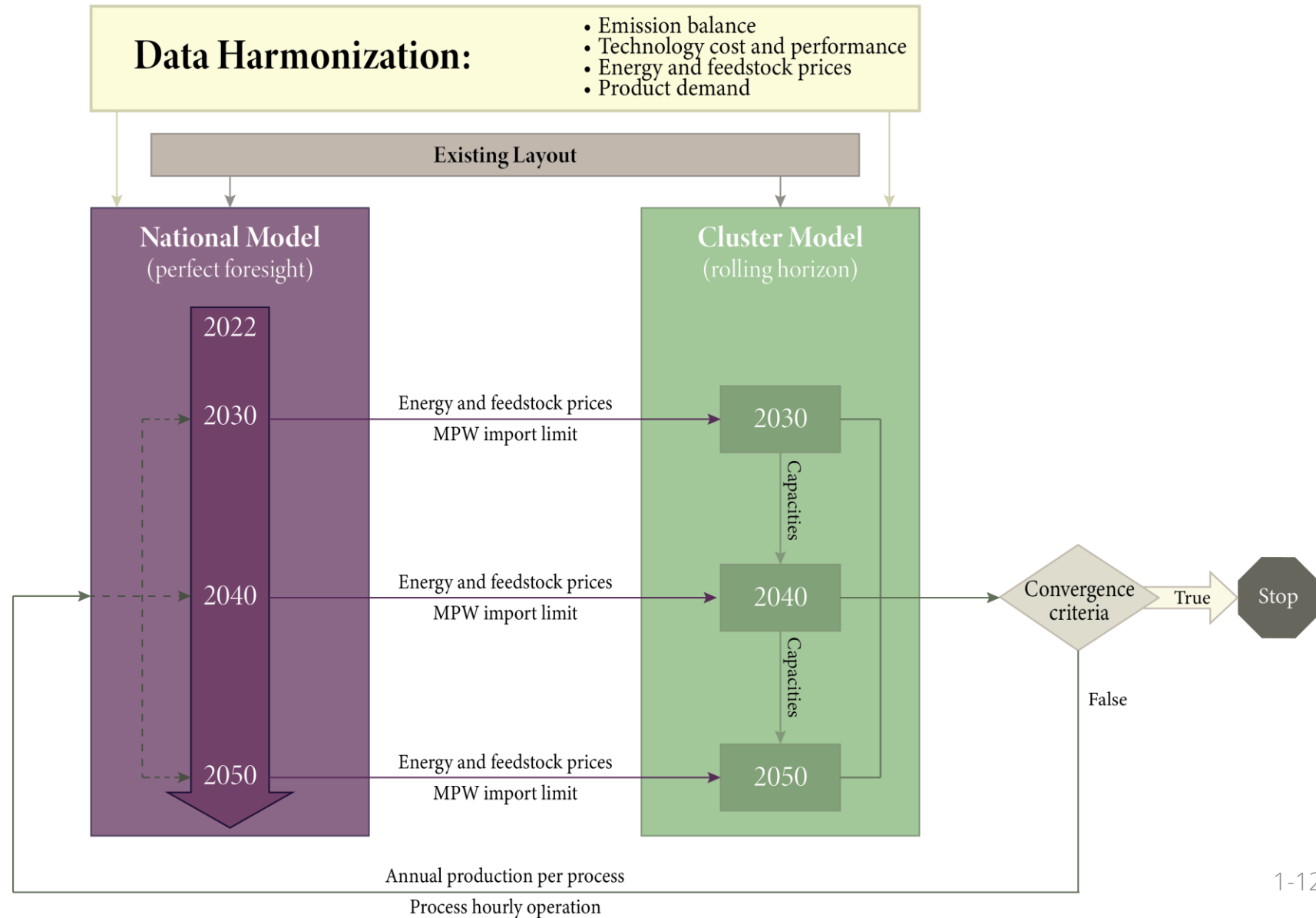
Modeling the transition at different scales



Agenda

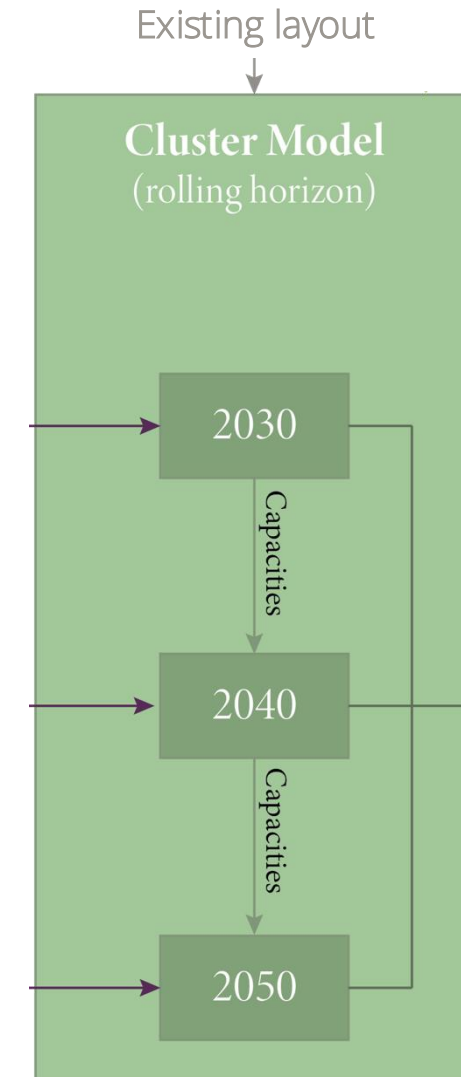
1. Models and case study
2. Standalone results
3. Hard-linking framework
4. Cluster ambitions within national strategies
5. Conclusion

Hard-linking framework



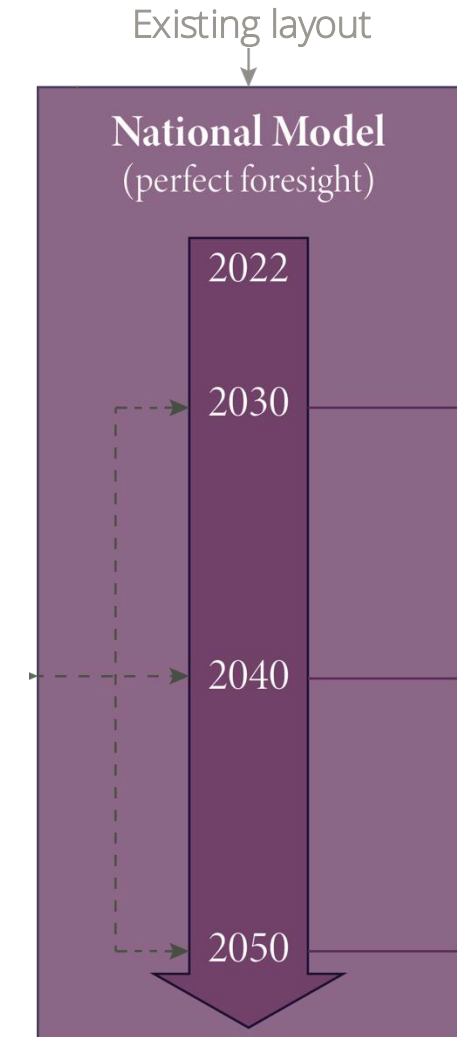
Cluster-level model

- MILP optimization framework
- Include both design and operational variables in optimization problem
- Full-year and hourly resolution (with design days)
- Sequential optimization of investment intervals

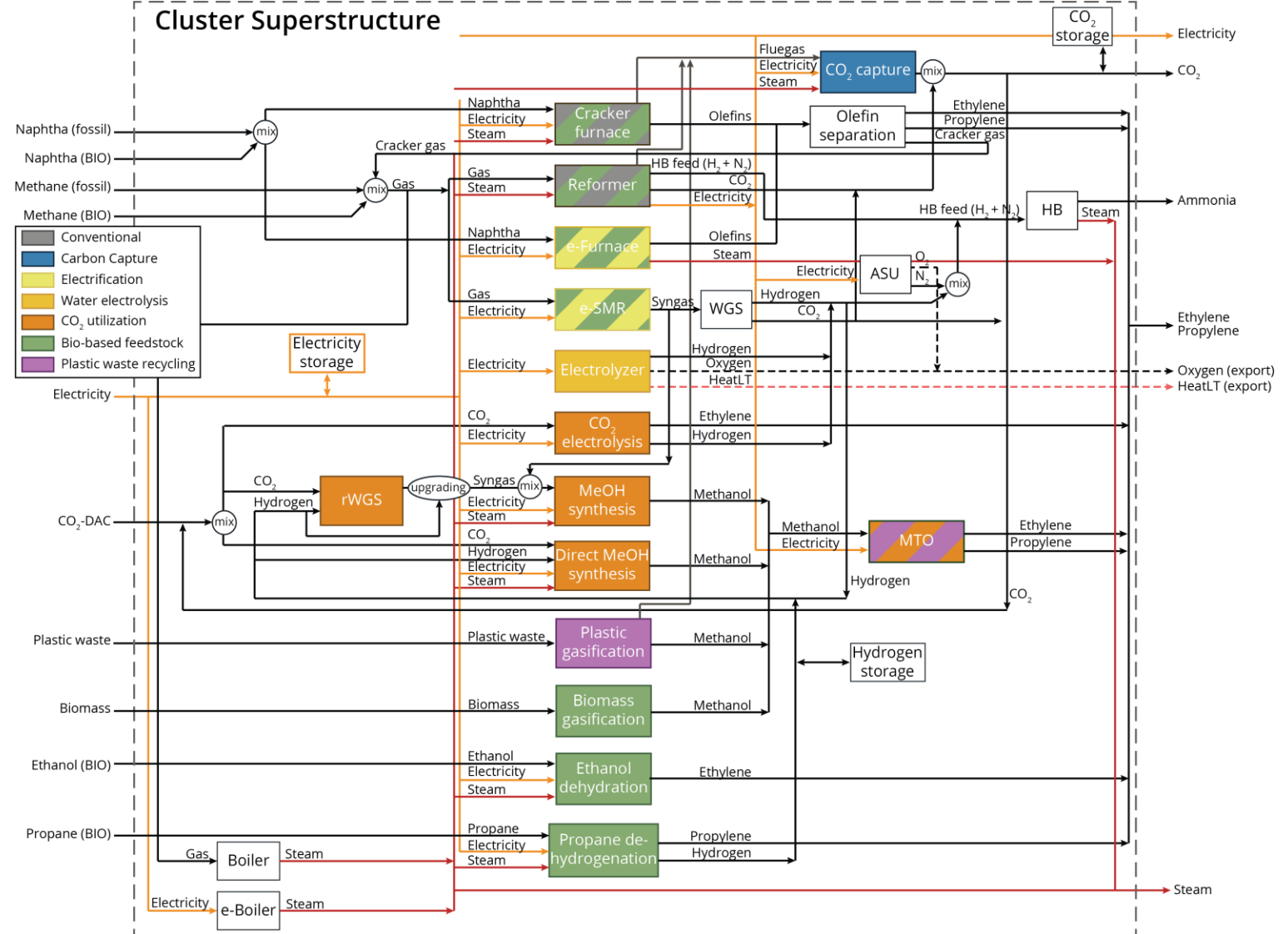
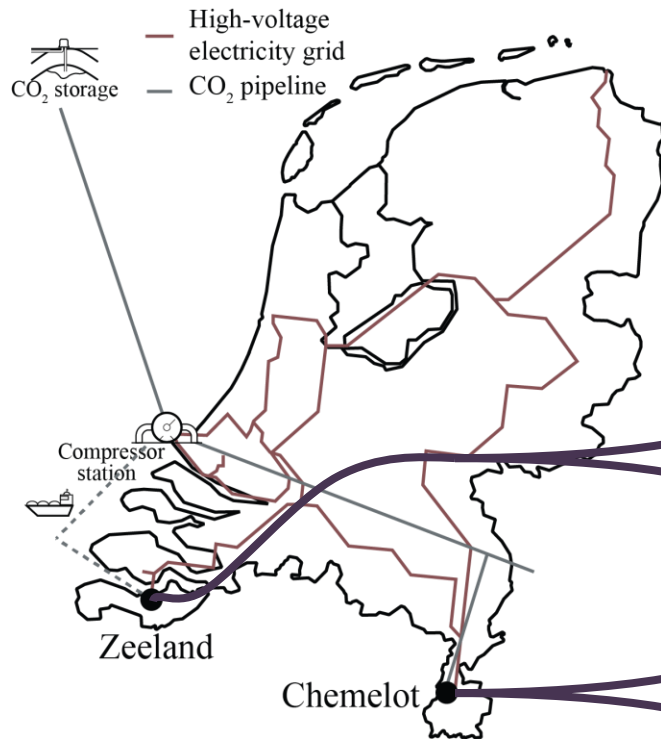


National model

- LP formulation in AIMMS
- Fully sector-coupled model of the Netherlands
- Hourly resolution with perfect foresight until 2050



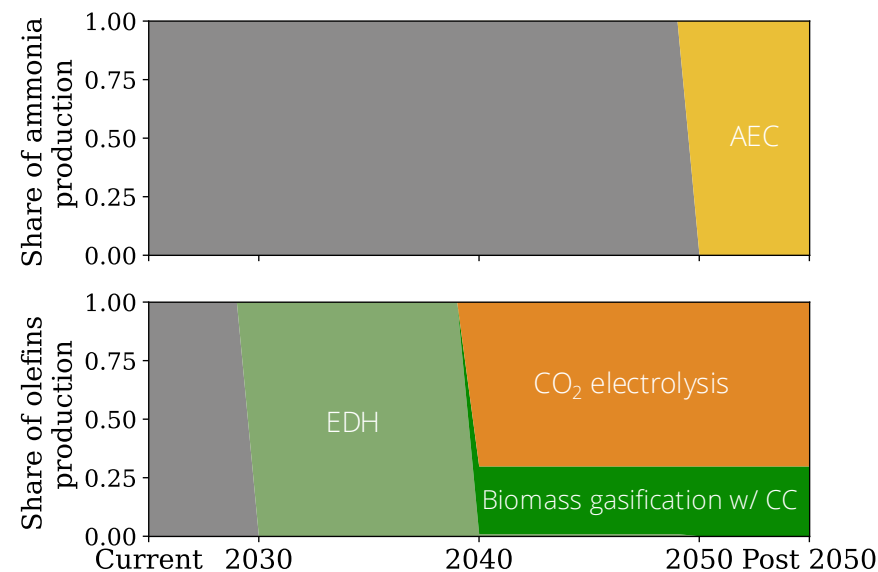
Case study at cluster-level



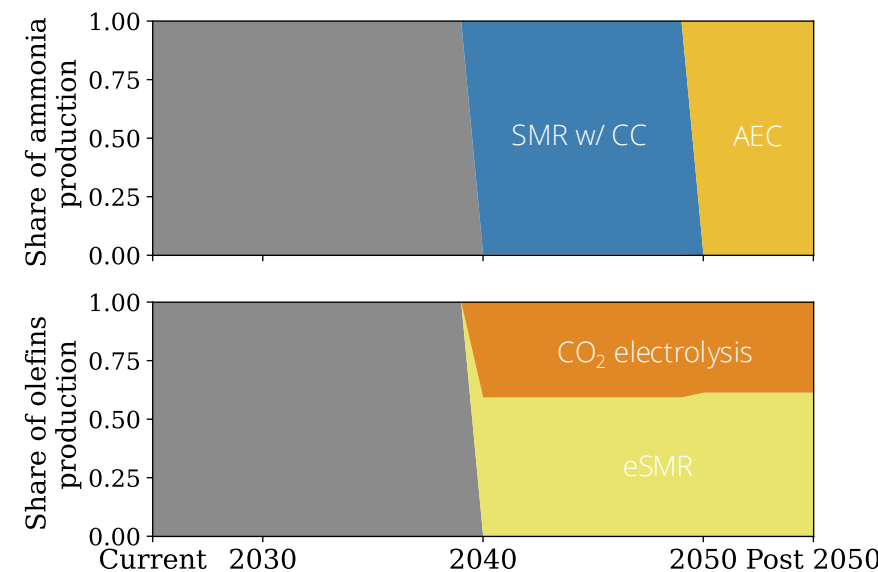
Cluster emission reductions

- Three ambition levels for the cluster to reduce emissions
 - High ambition: direct, energy-related and end-of-life emissions
 - Medium ambition: direct and energy-related emissions
 - Low ambition: cost-minimization
- All have a CO₂ price of 150 €/t from 2030 onwards

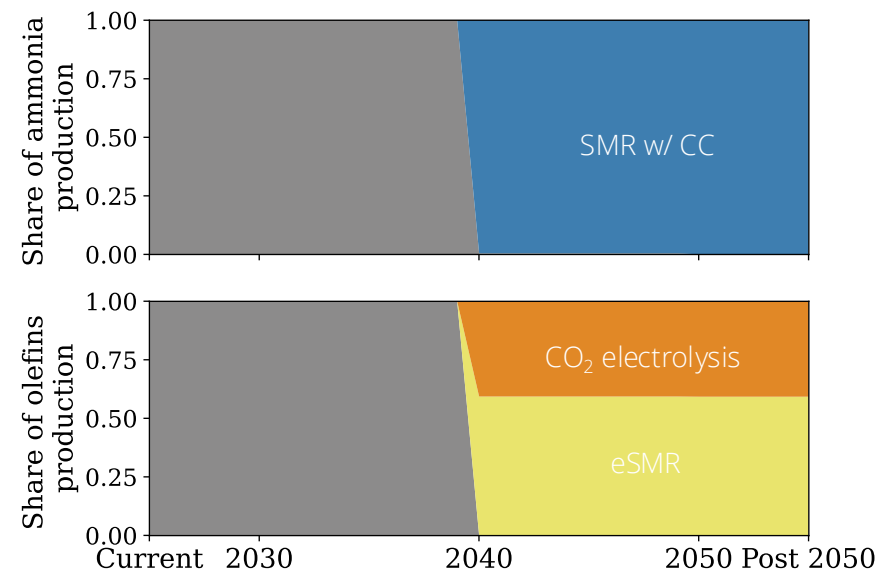
Cluster emission reduction pathways – standalone results



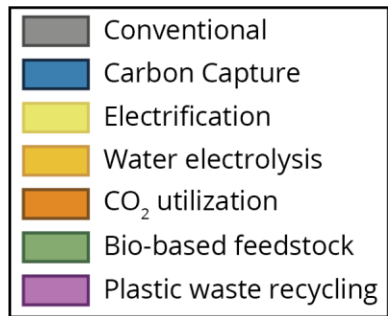
High ambition
(scope 1-3)



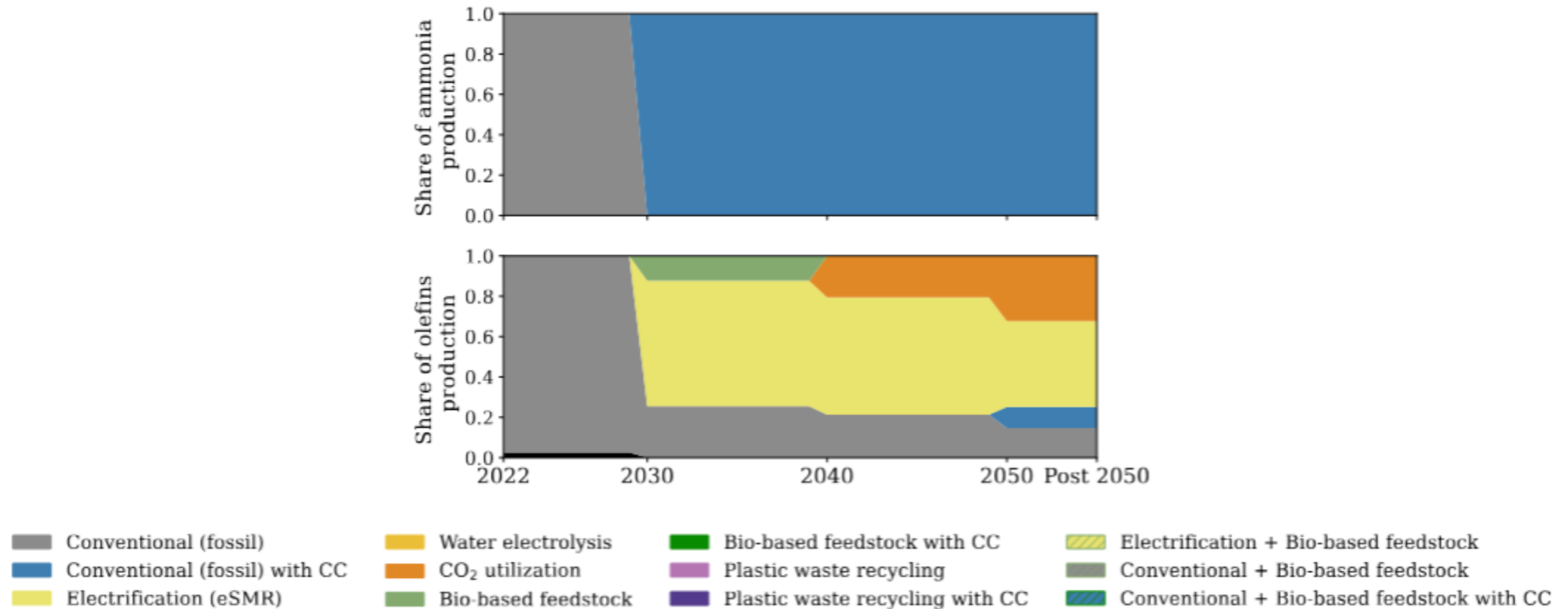
Medium ambition
(scope 1-2)



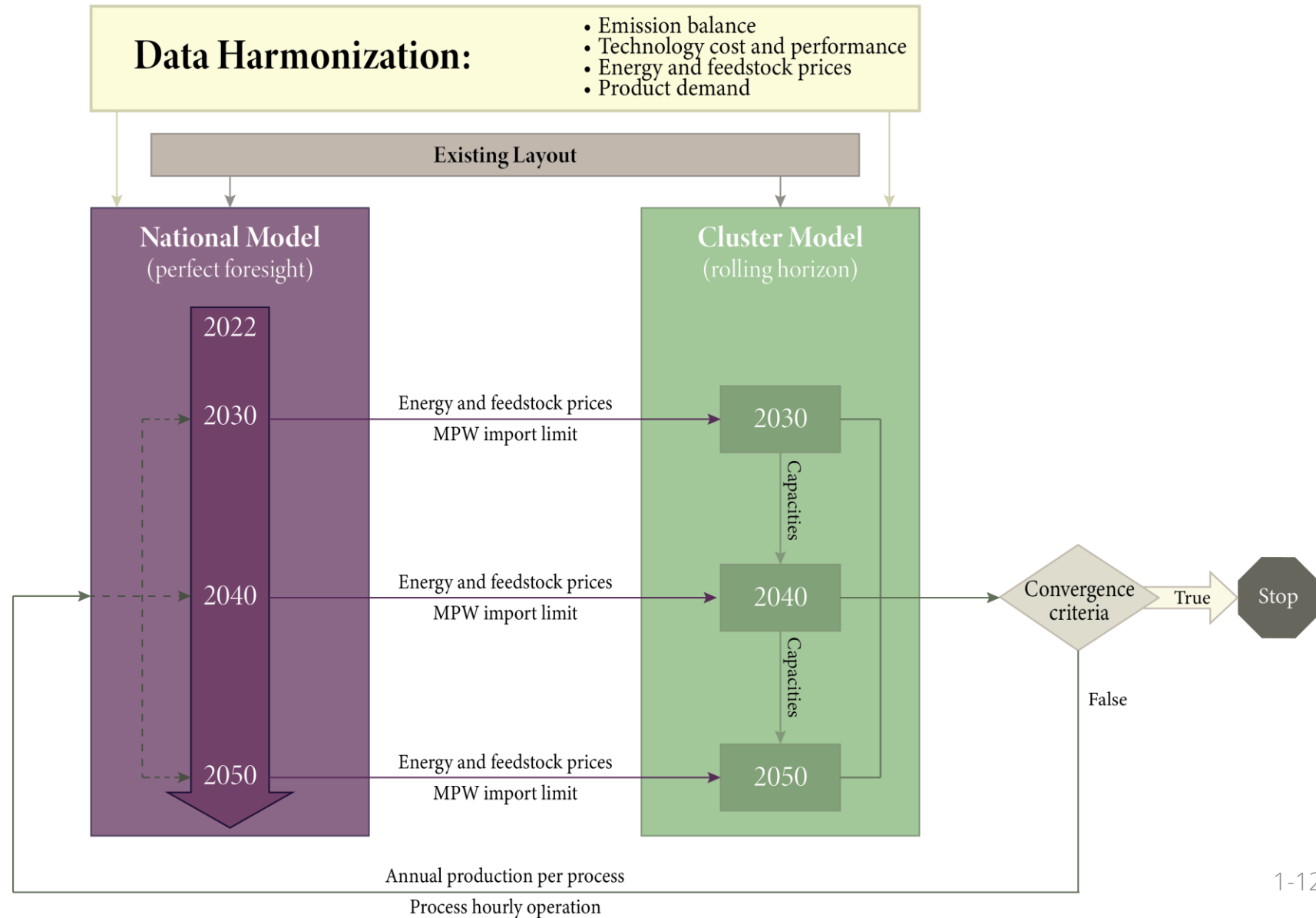
Low ambition (minimum costs)



National strategy – standalone results



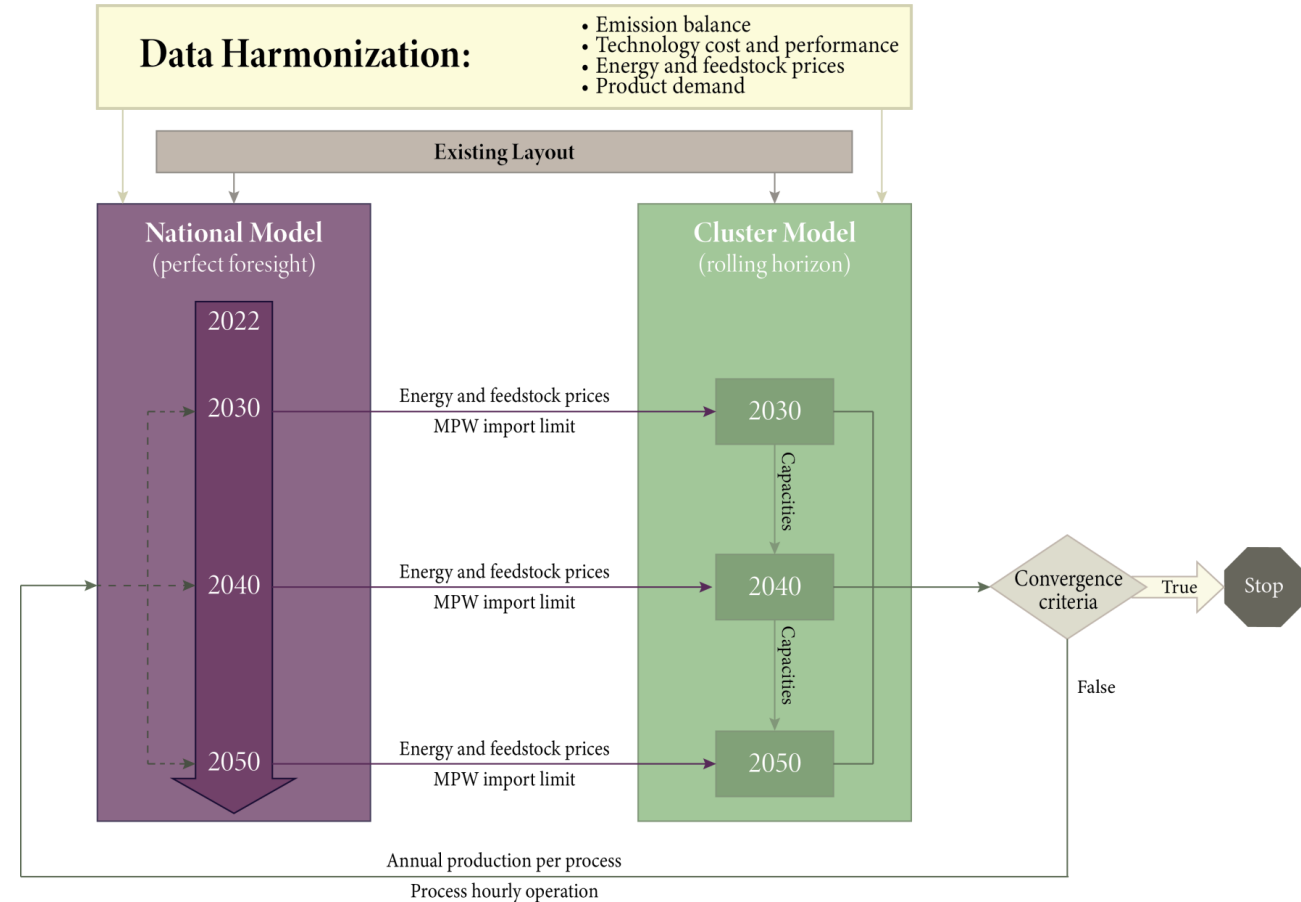
Hard-linking framework



Linking variables

National to cluster

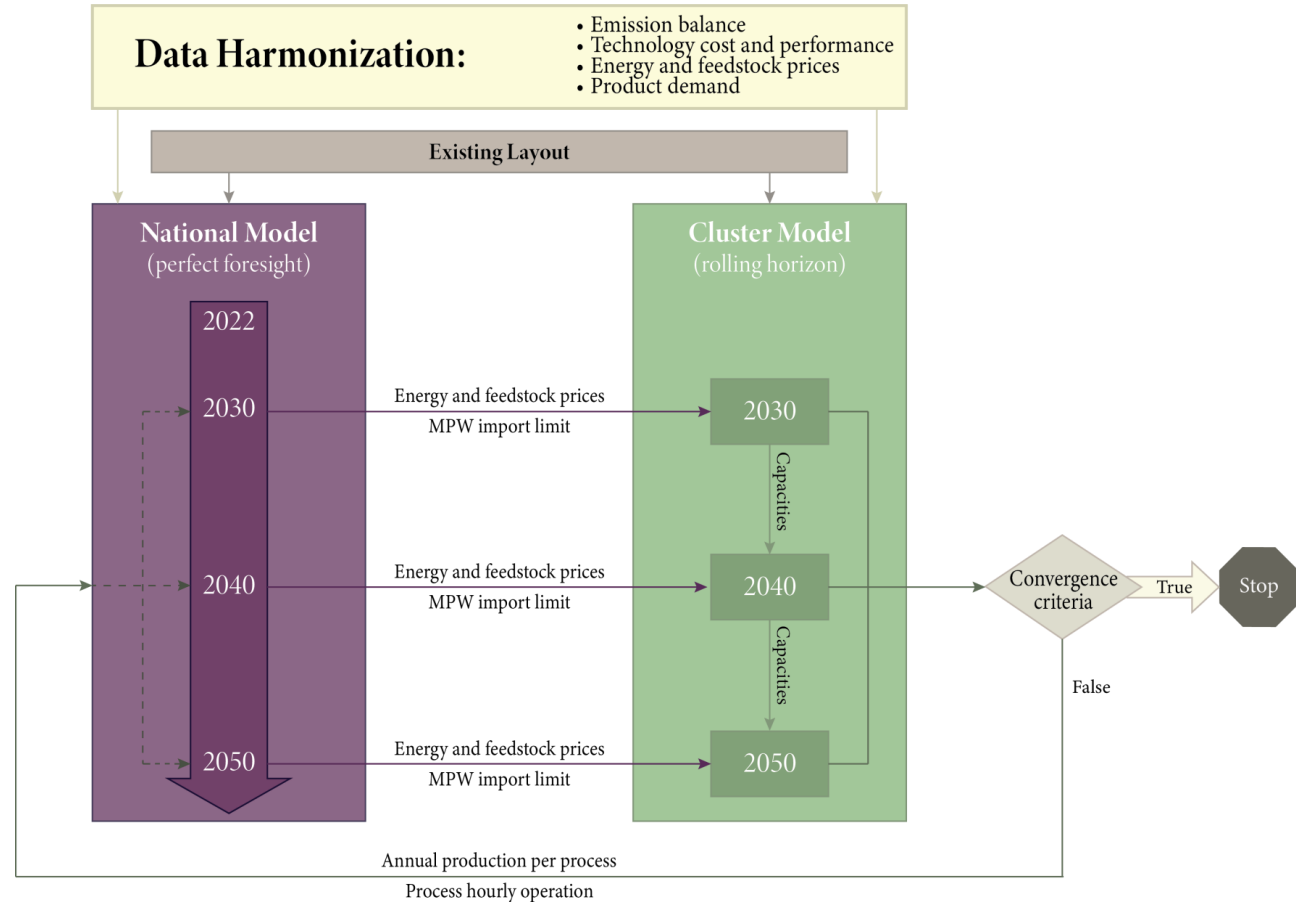
- Endogenous variables from the national model:
 1. Marginal costs of energy and feedstocks
 2. Availability of plastic waste for gasification



Linking variables

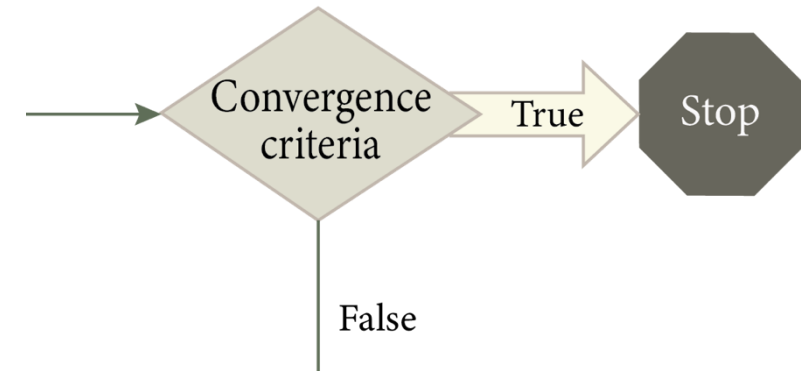
Cluster to national

- Endogenous variables from the cluster model:
 1. Annual production of installed technologies
 2. Hourly operating profiles of electrified technologies



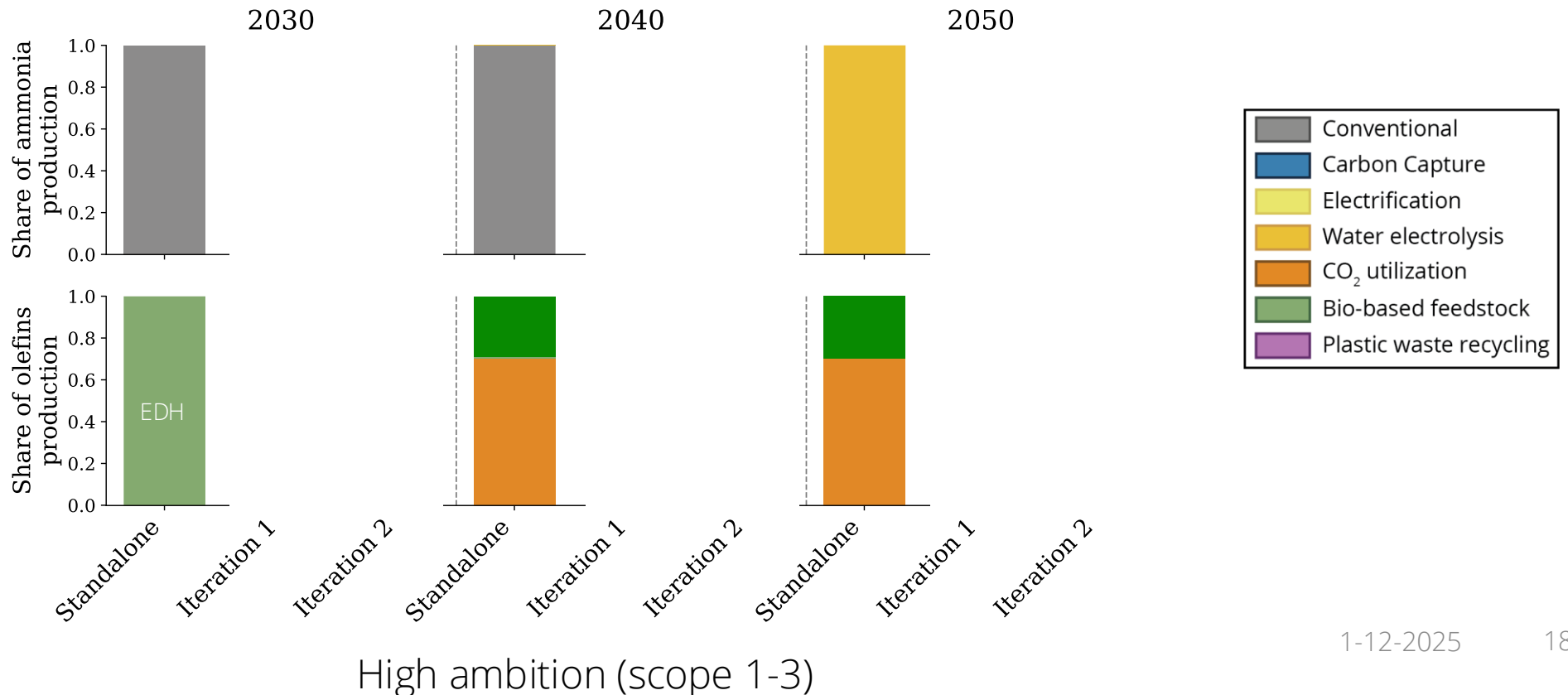
Convergence criteria

- Convergence assessment based on cluster variables
- Calculate the relative change in annual production between iterations for each process
- Required to meet both criteria:
 1. Average relative change of all processes: $<1\%$
 2. Max relative change of all processes: $<10\%$



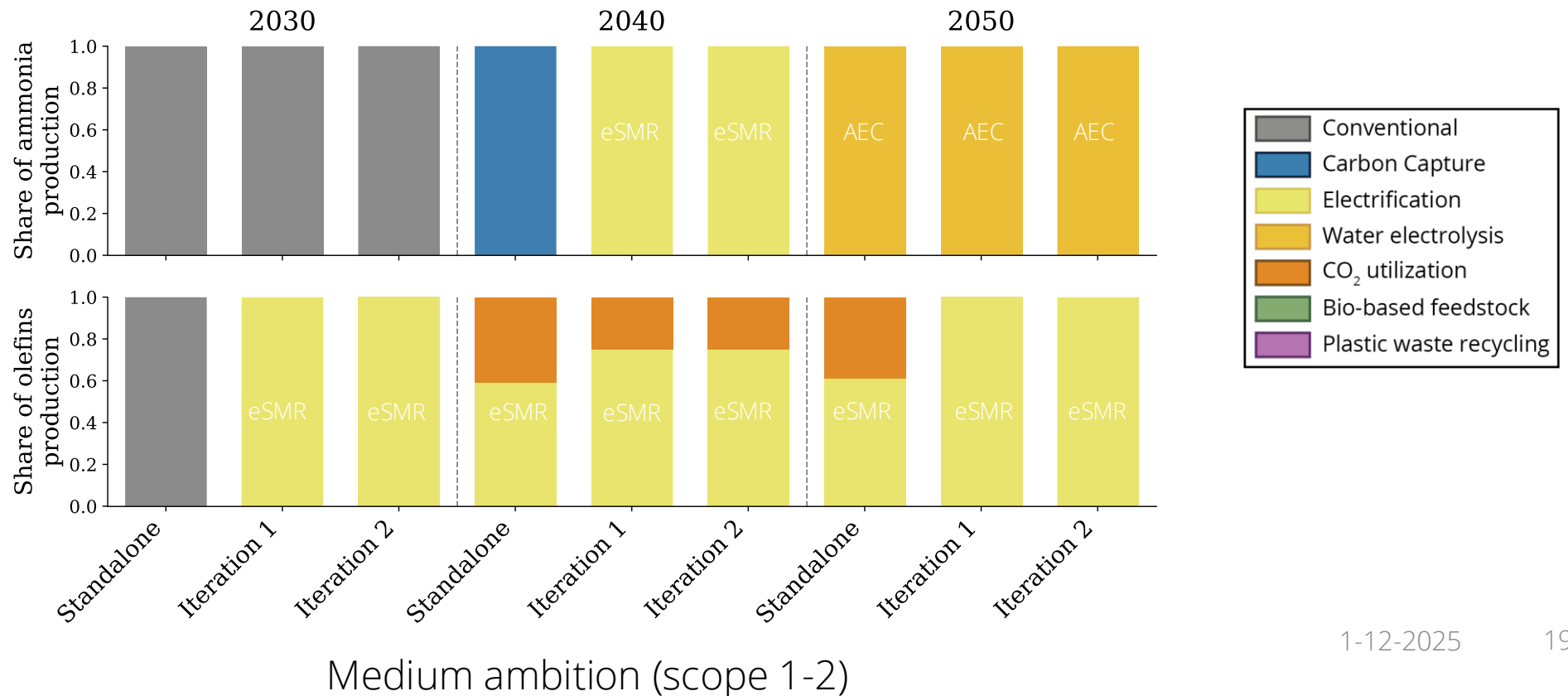
Cluster ambitions and national policy alignment

- Mismatch of supply and demand at national level leads to alternative least-cost olefin pathway in 2030
- Path-dependent consequences: eSMR is used for ammonia production in 2040

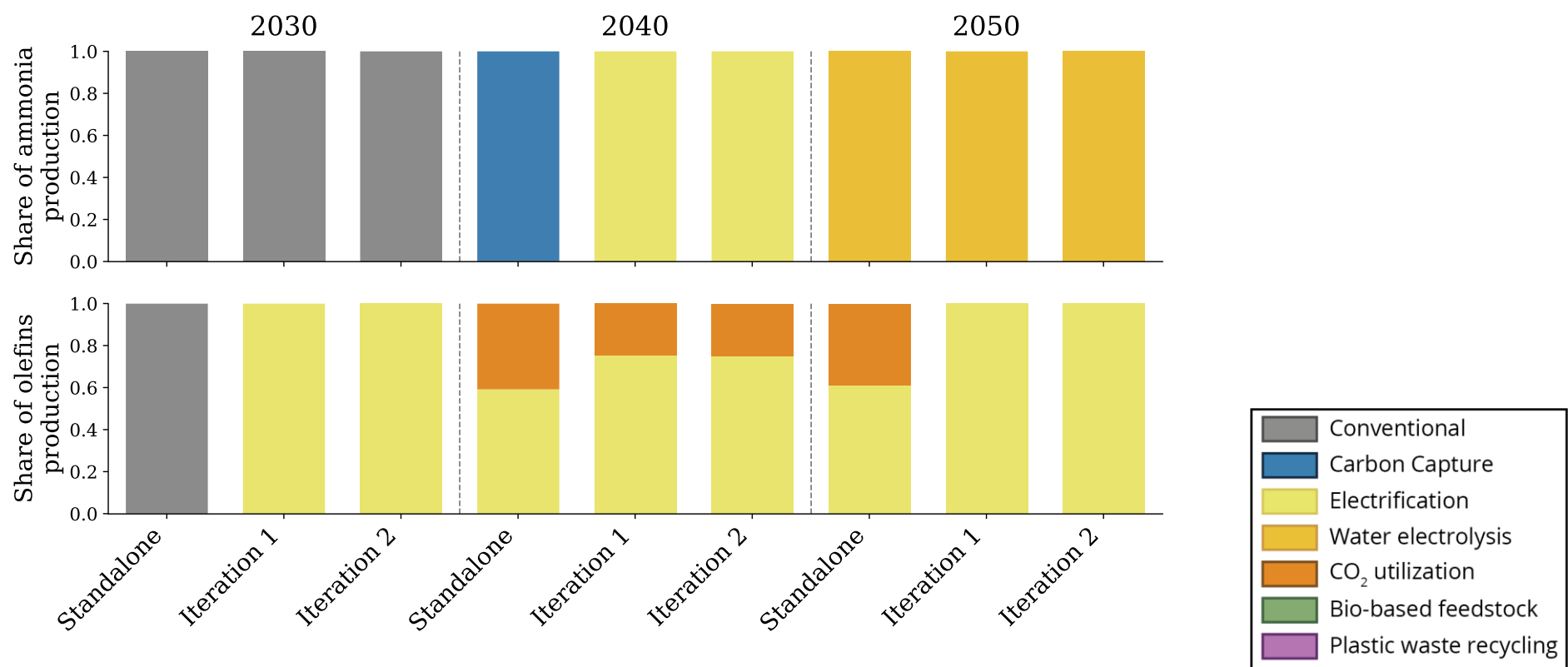


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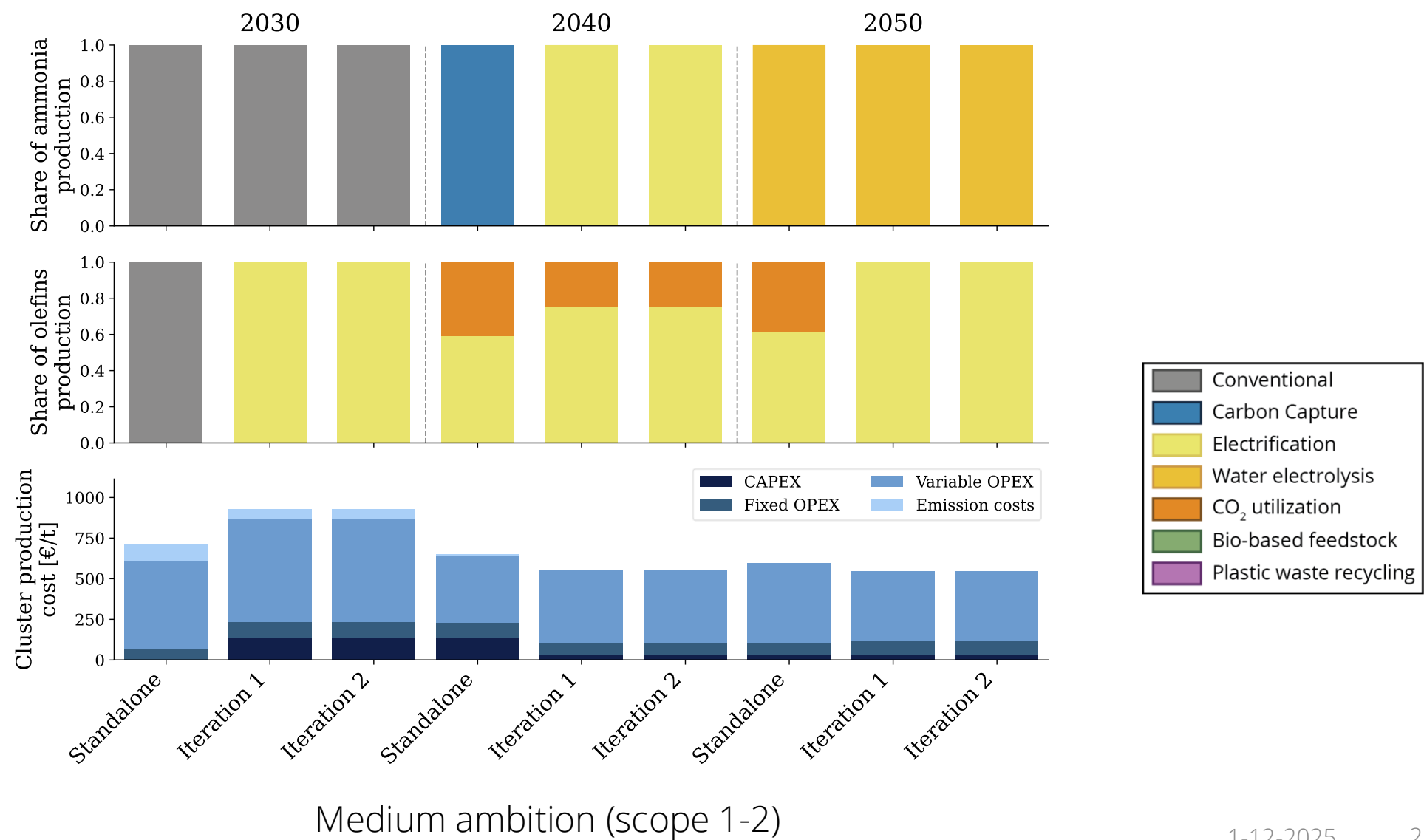


Costs of misaligned emission reduction ambitions

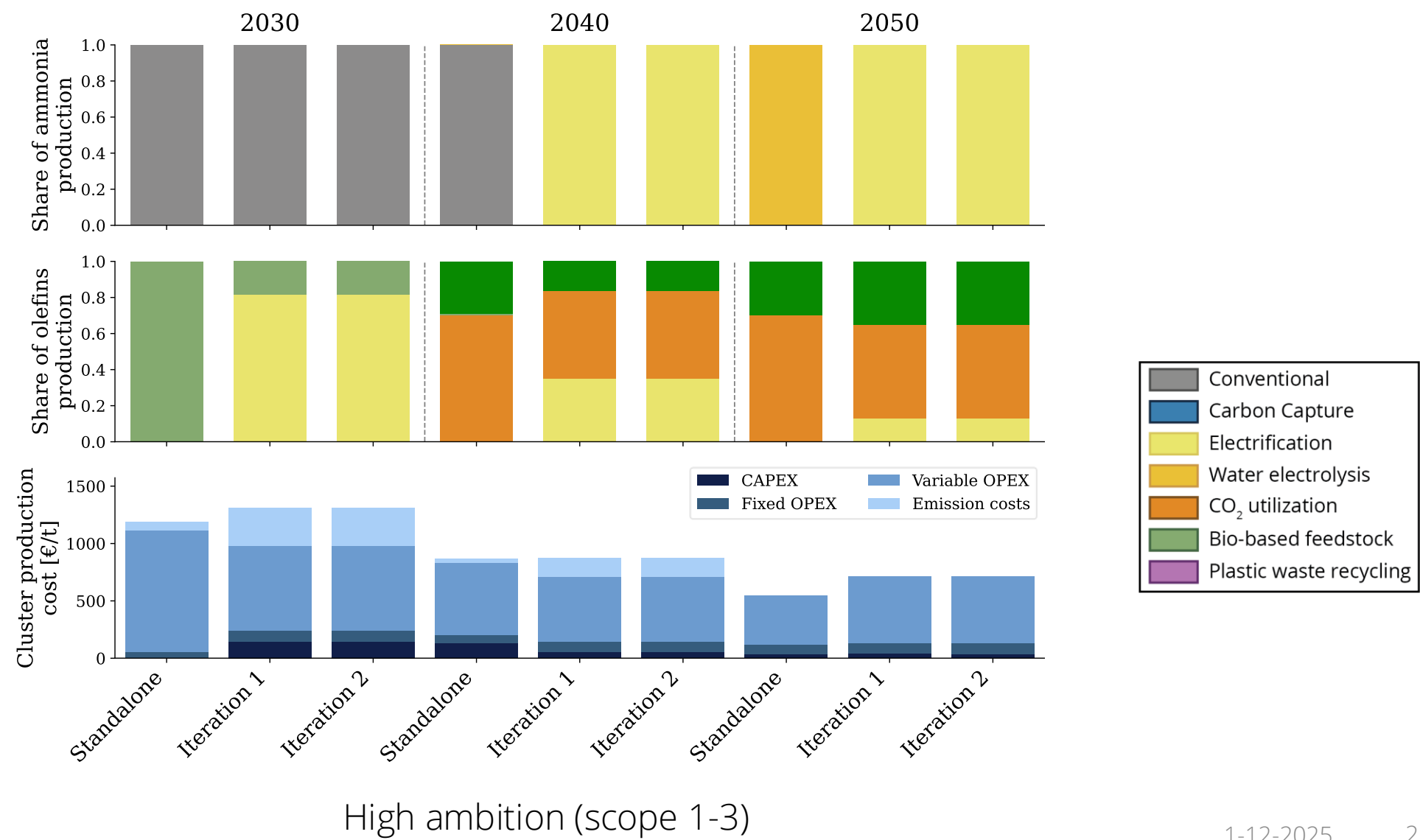


Medium ambition (scope 1-2)

Costs of misaligned emission reduction ambitions



Costs of misaligned emission reduction ambitions



Conclusions

1. Several near-term investments consistently emerge as “low-regret” technologies
2. Alternative feedstocks need end-of-life carbon pricing to compete with fossil
3. Data harmonization and model-linking can identify trade-offs between different scales of the transition
4. Misaligned ambitions between clusters and the national system can lead to alternative least-cost pathways and higher costs



**Utrecht
University**

References, data and code availability

Tiggeloven, J. L., Faaij, A. P. C., Kramer, G. J., & Gazzani, M. (2023). *Optimization of Electric Ethylene Production: Exploring the Role of Cracker Flexibility, Batteries, and Renewable Energy Integration*. *Industrial & Engineering Chemistry Research*, 62(40), 16360–16382.

Tiggeloven, J. L., Faaij, A. P. C., Kramer, G. J., & Gazzani, M. (2025). *Optimizing Emissions Reduction in Ammonia-Ethylene Chemical Clusters: Synergistic Integration of Electrification, Carbon Capture, and Hydrogen*. *Industrial & Engineering Chemistry Research*, 64, 4479–4497.

Tiggeloven, J. L., Faaij, A. P. C., Kramer, G. J., & Gazzani, M. (submitted). *Rethinking Chemical Clusters: Greenfield and Brownfield Transition to Net-Zero by 2050*.

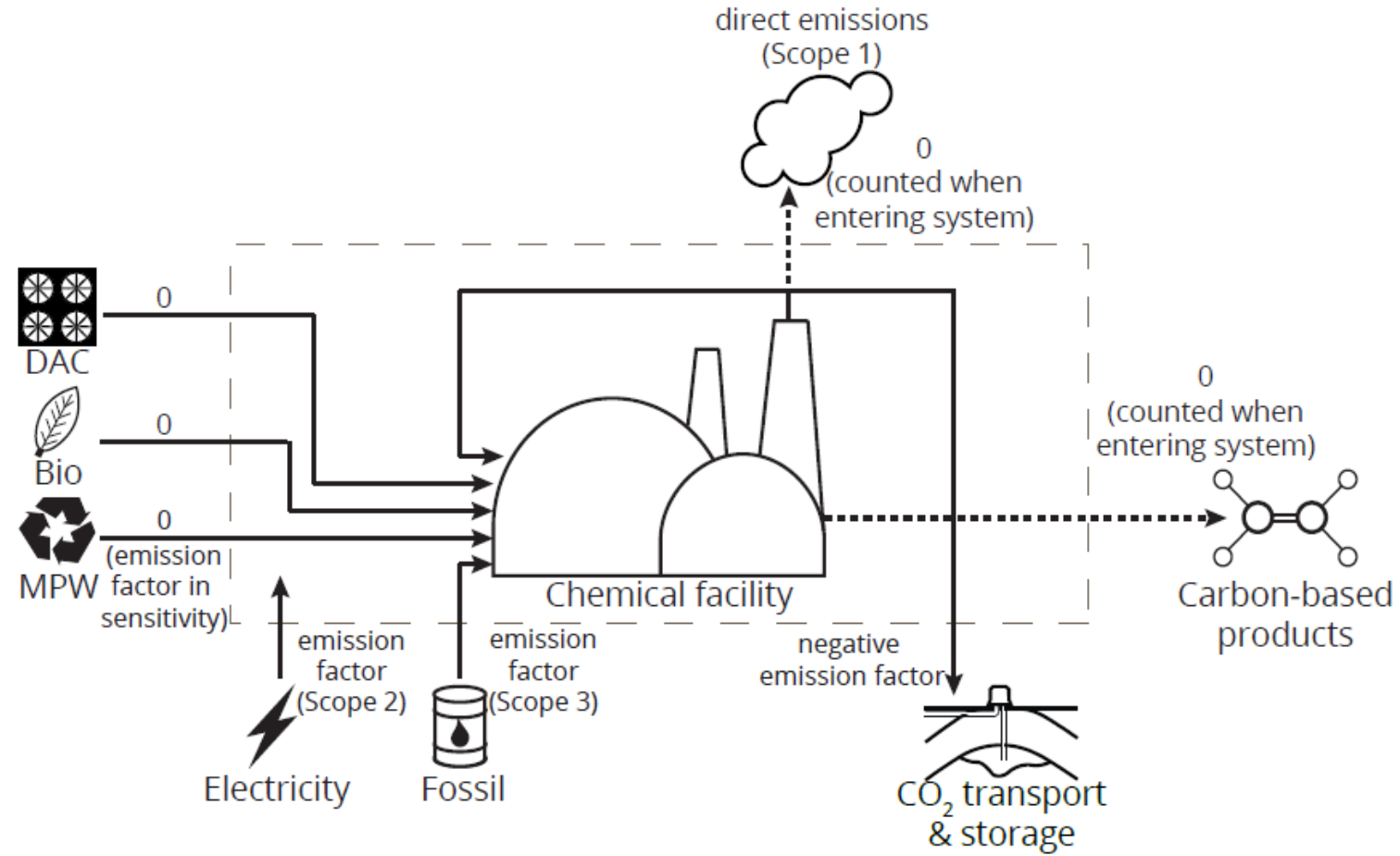
Tiggeloven, J. L., West, K., Mulder, A. J., Faaij, A. P. C., Kramer, G. J., Koning, V., & Gazzani, M. (in preparation). *Supporting the transition to a net-zero chemical industry by coupling national and cluster-level models*.

Data and code availability:

The AdOpT-NET0 optimization framework and all data supporting the research are open source.

- <https://github.com/julia1071>
- Wiegner, J. F., Tiggeloven, J. L., Bertoni, L., Ossentjuk, I. M., & Gazzani, M. (2025). AdOpT-NET0: A technology-focused Python package for the optimization of multi-energy systems. *Journal of Open Source Software*, 10(106), 7402

Emission scope



Optimization framework

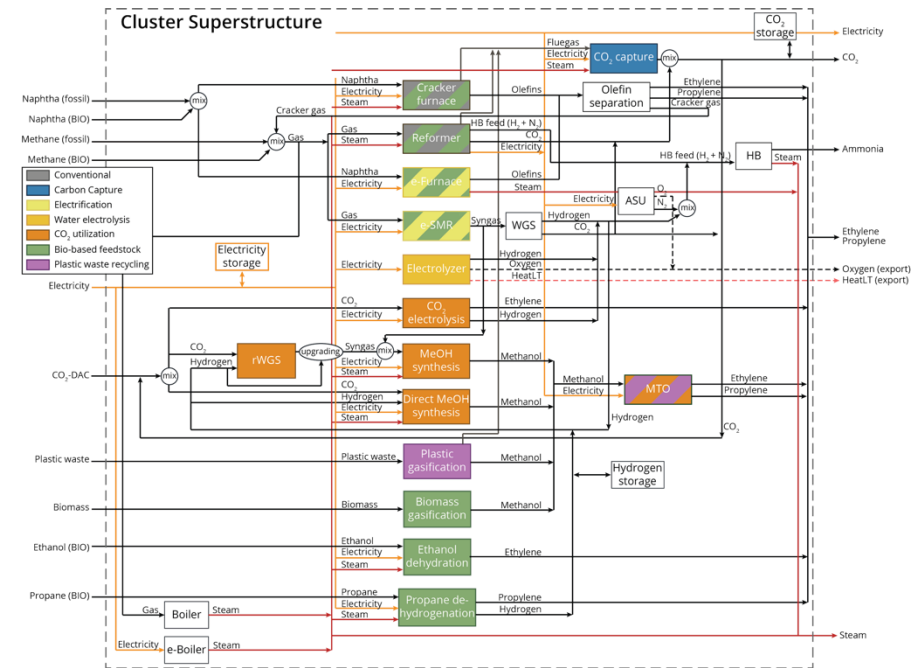
- MILP optimization framework
- Include both design and operational variables in optimization problem
- Full-year and hourly resolution (with design days)

$$\min_{\mathbf{x}, \mathbf{y}} (\mathbf{c}^T \mathbf{x} + \mathbf{d}^T \mathbf{y})$$

subject to

$$\mathbf{Ax} + \mathbf{By} = \mathbf{b}$$

$$\mathbf{x} \geq \mathbf{0} \in \mathbb{R}^N, \mathbf{y} \in \mathbb{N}^M$$



Decision var.

- Plant capacity
- Tech. selection & size
- Plant operation (production level)

Constraints

- Technology performance
- Energy balances
- Material balances

Optimization framework

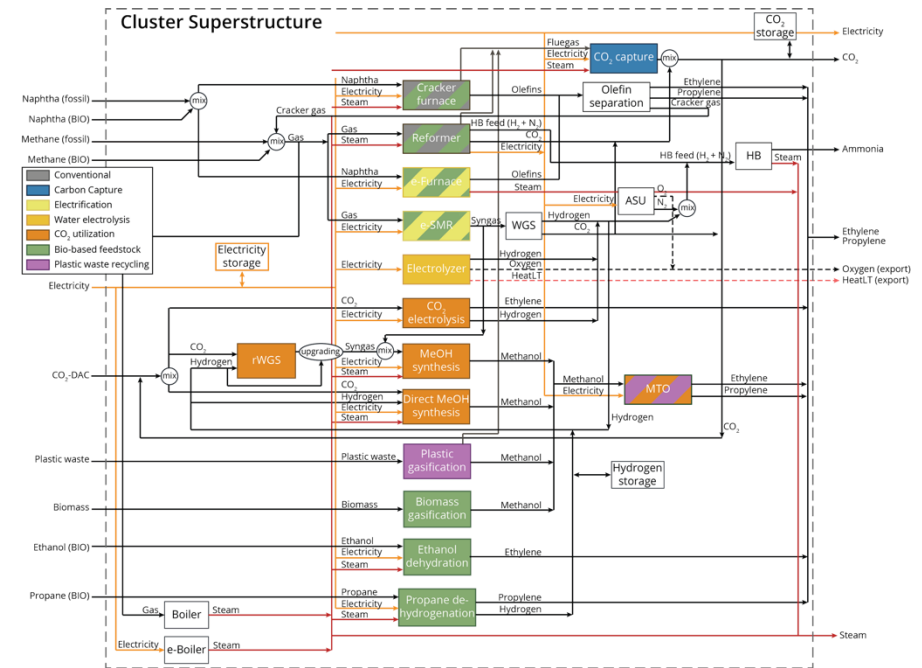
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Input Data	Decision var.	Constraints
<ul style="list-style-type: none"> • Weather conditions • Demands • Prices • Carbon rates • Technology cost and performance 	<ul style="list-style-type: none"> • Plant capacity • Tech. selection & size • Plant operation (production level) 	<ul style="list-style-type: none"> • Technology performance • Energy balances • Material balances

Optimization framework

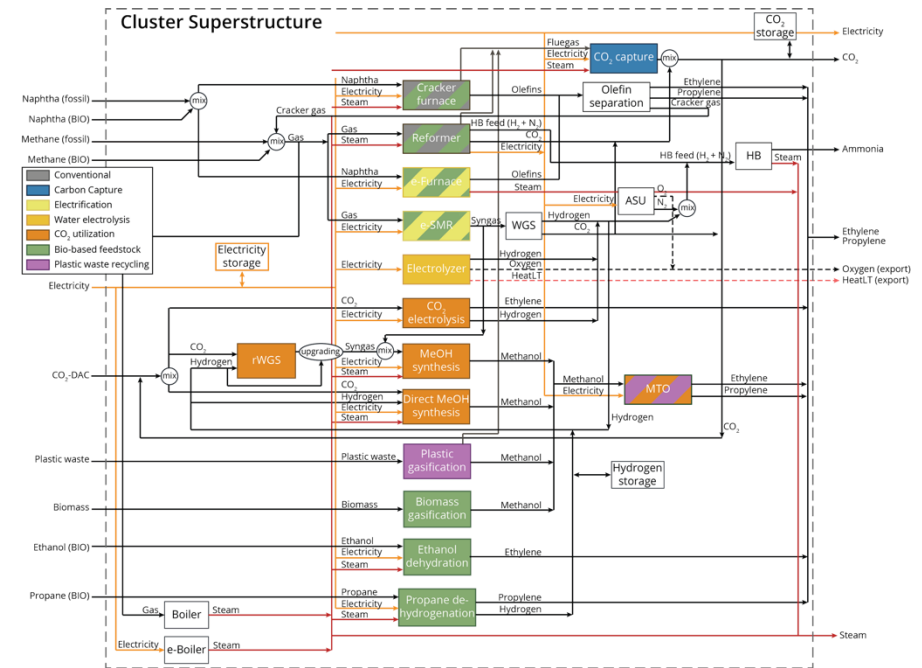
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subject to

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Objective funct.

- Total annual cost
- Total annual emissions