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### Linking Life Cycle and Integrated Assessment Modeling to Evaluate Technologies in an Evolving System Context: A Power-to-Hydrogen Case Study for the United States

Patrick Lamers,\* Tapajyoti Ghosh, Shubhankar Upasani, Romain Sacchi, and Vassilis Daioglou

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## Life-cycle Assessment Integration into Scalable Open-source Numerical models (LiAISON) for analyzing emerging low-carbon technologies



TJ Ghosh  
(PI)



Patrick Lamers  
(Senior Advisor)



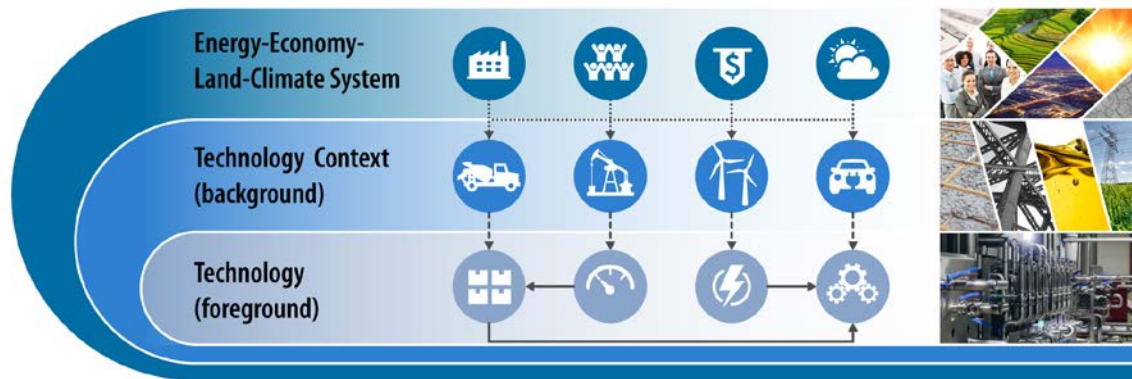
Shubhankar  
Upasani  
(inventories)

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# Lifecycle Analysis Integration into Scalable Open-source Numerical models



**Research Question:** What are the future impacts and tradeoffs of present-day novel technologies accounting for transitions in the energy and manufacturing sectors as well as technology improvements?

**Method:** Coded, prospective life cycle assessment using long-term, coherent scenarios of the energy-economy-land-climate system to quantify the effects of background system changes and foreground technology improvements for various technologies.

**Value-add:** Inform R&D prioritization for novel technologies and preemptively address potential tradeoffs and unintended consequences of their large-scale deployment.

Funding: Department of Energy

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Project staff



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(PI)



Patrick Lamers  
(Senior Advisor)



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(inventories)

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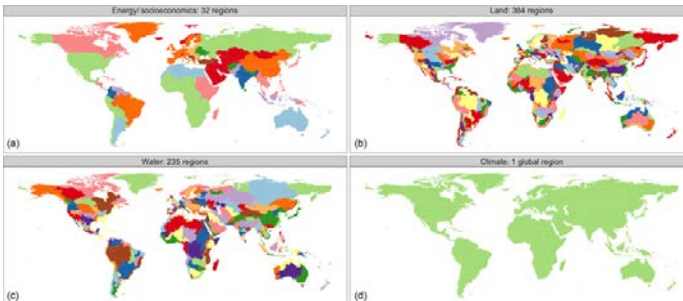
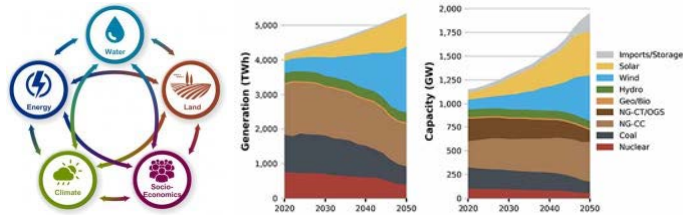
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Linkages: FECM, BOTTLE, others

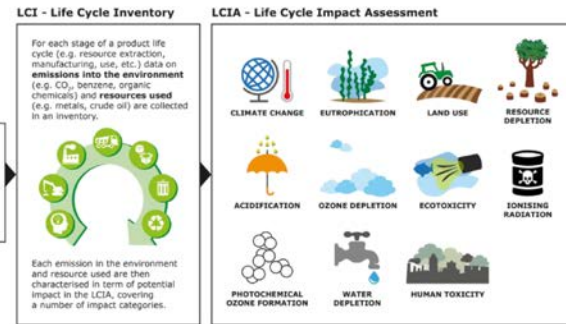
# Vision / Motivation

## Prospective system models



## Life Cycle Assessment

Ecosystem Impacts	Human Impacts	Resource Depletion
Climate Change Acid Rain Eutrophication Land Use Change Solid Waste Toxicity	Ozone Depletion Smog Particulate Matter Carcinogens Toxicity	Fossil Fuel Freshwater Soil Forest Grassland Minerals



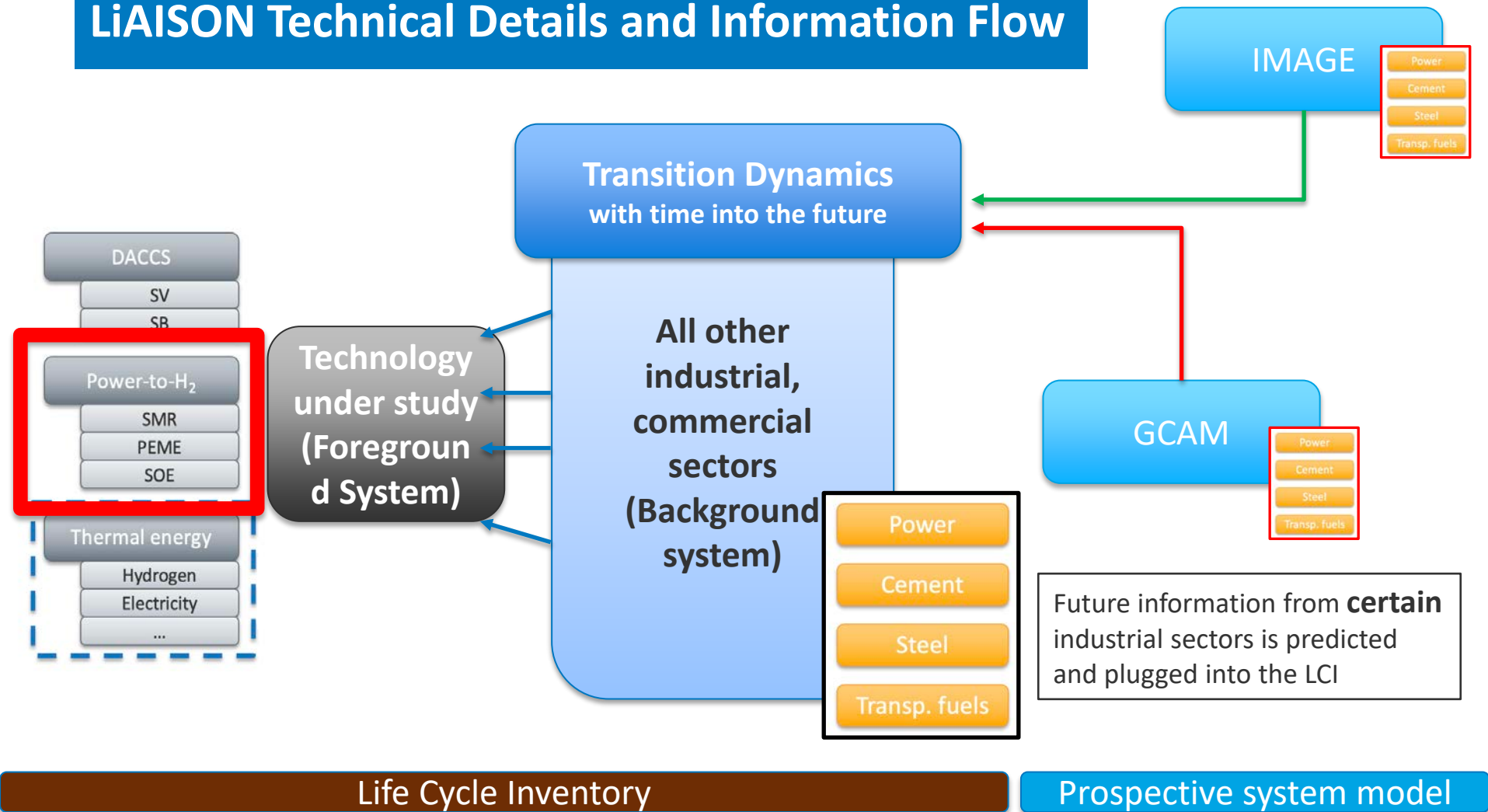
*Comprehensive but scenarios often depend on limited metrics. Models are primarily cost-driven.*

URL: [nrel.gov](http://nrel.gov); [pnnl.gov](http://pnnl.gov)

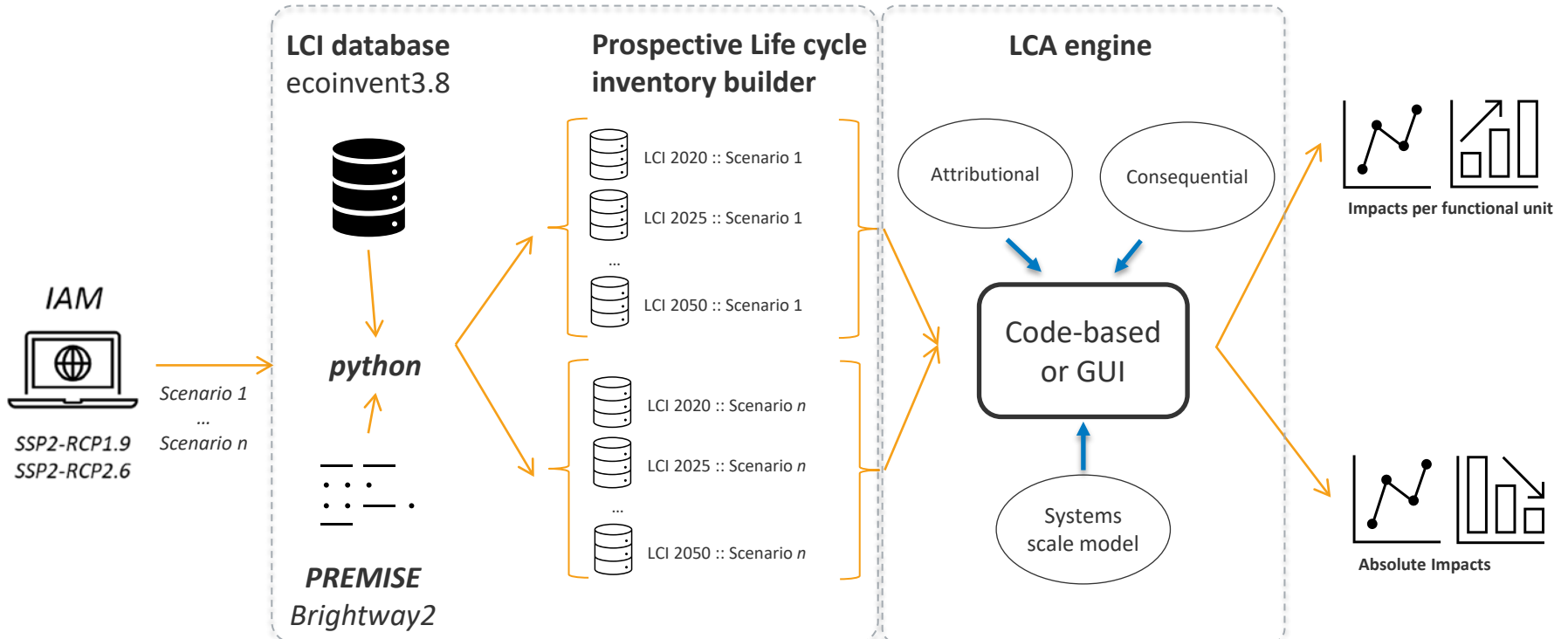
*Multi-metric but results are context-specific. Analyses have varying system boundaries (hard to compare).*

URL: <https://epca.jrc.ec.europa.eu/lifecycleassessment.html>;  
<https://www.sciencedirect.com/topics/engineering/life-cycle-impact-assessment>

# LiAISON Technical Details and Information Flow



# Methodology - LiAISON Technical Details



**NREL model inputs  
(scenarios)**

**Step 1:  
Code-based life cycle inventory creation**

**Step 2:  
User-defined analysis**

**Product- or System-  
level results**

**BACKGROUND**

**FOREGROUND**

**RESULT ANALYSIS**

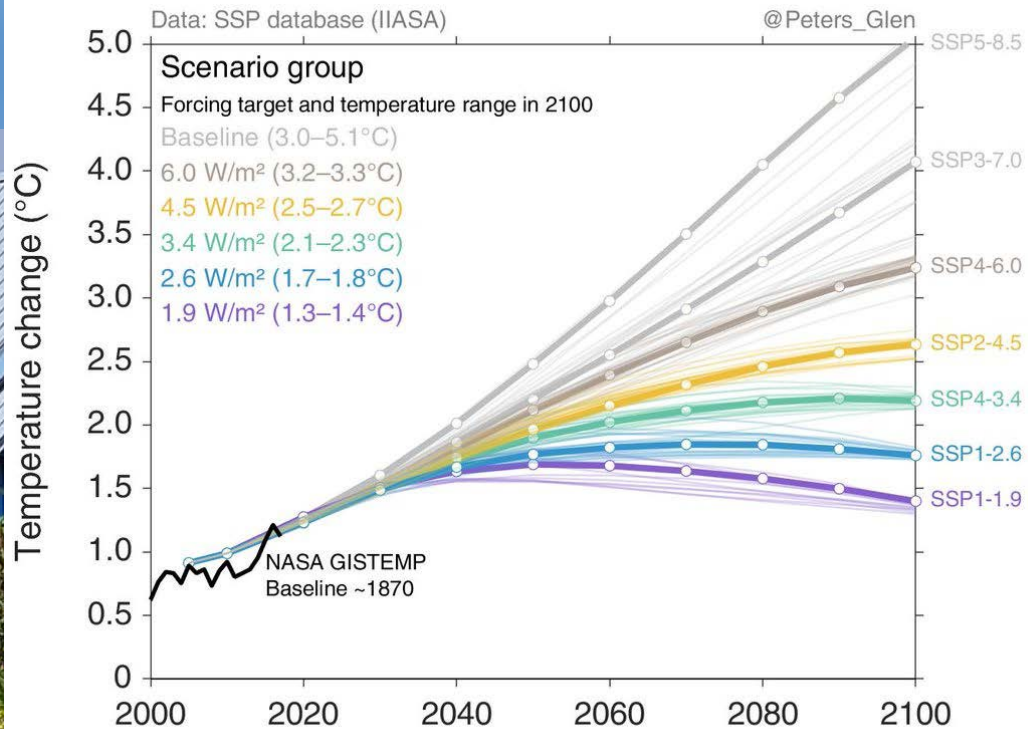




## Climate Change 2022 Mitigation of Climate Change

**Shared-Socioeconomic Pathways (SSP):**  
socioeconomic challenges for mitigation and adaption

**Representative Concentration Pathways (RCP):**  
global radiative forcing levels ( $\text{W}/\text{m}^2$ )





# Sector projections

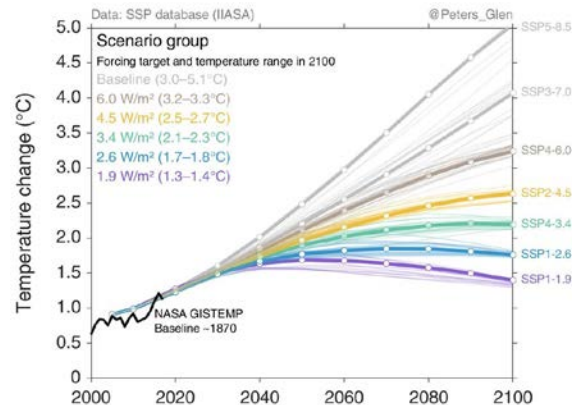
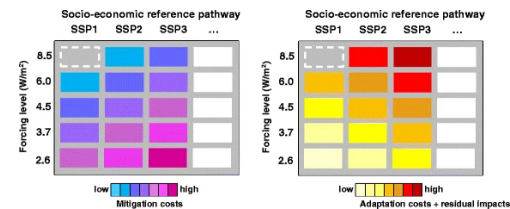
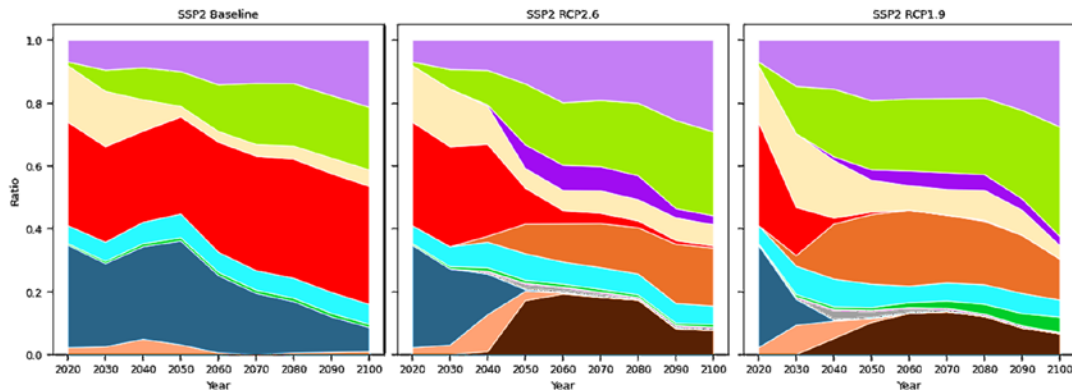


The background scenarios define technology compositions and efficiencies across four sectors:

- Power,
- Cement,
- Steel,
- Transport fuels.

## Example: power sector specifics – Shared Socio-Economic Pathways (IPCC)

- **SSP2-RCP1.9:** net zero GHG economy by 2035, net zero GHG economy by 2050
- **SSP2-RCP2.6:** delayed by ~ 20 years
- **SSP2-reference:** no targets





# Power-to-Hydrogen (PtH<sub>2</sub>)

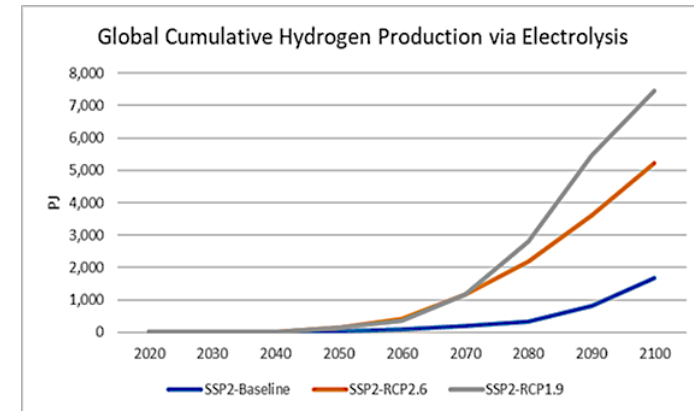
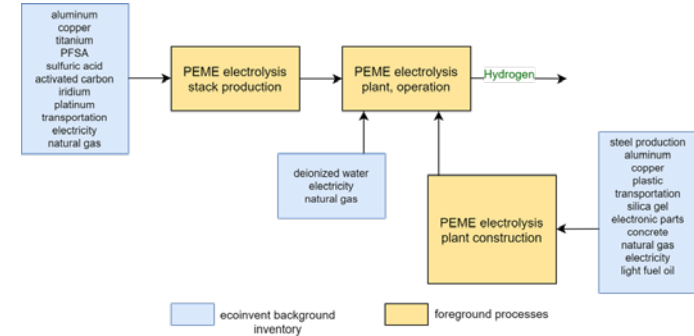


## Technologies:

- **Steam methane reforming (reference)** : H<sub>2</sub> generation via steam methane reforming of natural gas to produce syngas and then H<sub>2</sub>. (*Baseline*)
- **Solid-oxide electrolysis (SOE)**: H<sub>2</sub> generation via electrolysis in a fuel cell with a solid oxide/ ceramic electrolyte (adv: high efficiency).
- **Polymer-electrolyte-membrane electrolysis (PEME)**: H<sub>2</sub> generation via electrolysis in a cell with a solid polymer electrolyte (adv: low weight and volume).

*Adjusted to background deployment levels in the respective scenarios.*

*Foreground dynamics via learning-by-doing in the deployment stage.*

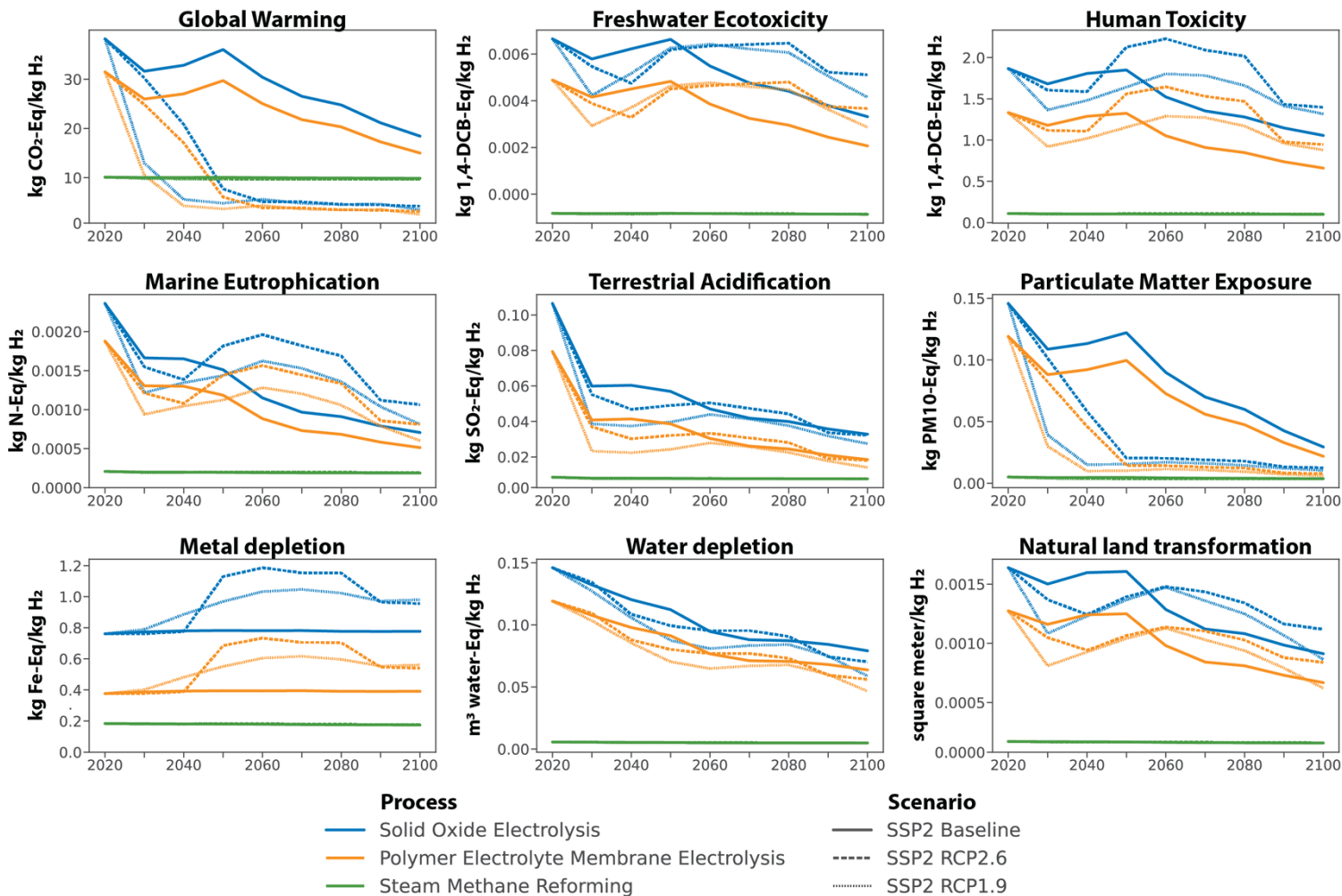


# Future impacts (ReCiPe; 1 kg H<sub>2</sub>) due to changes in the cement, steel, power, transport fuel sectors

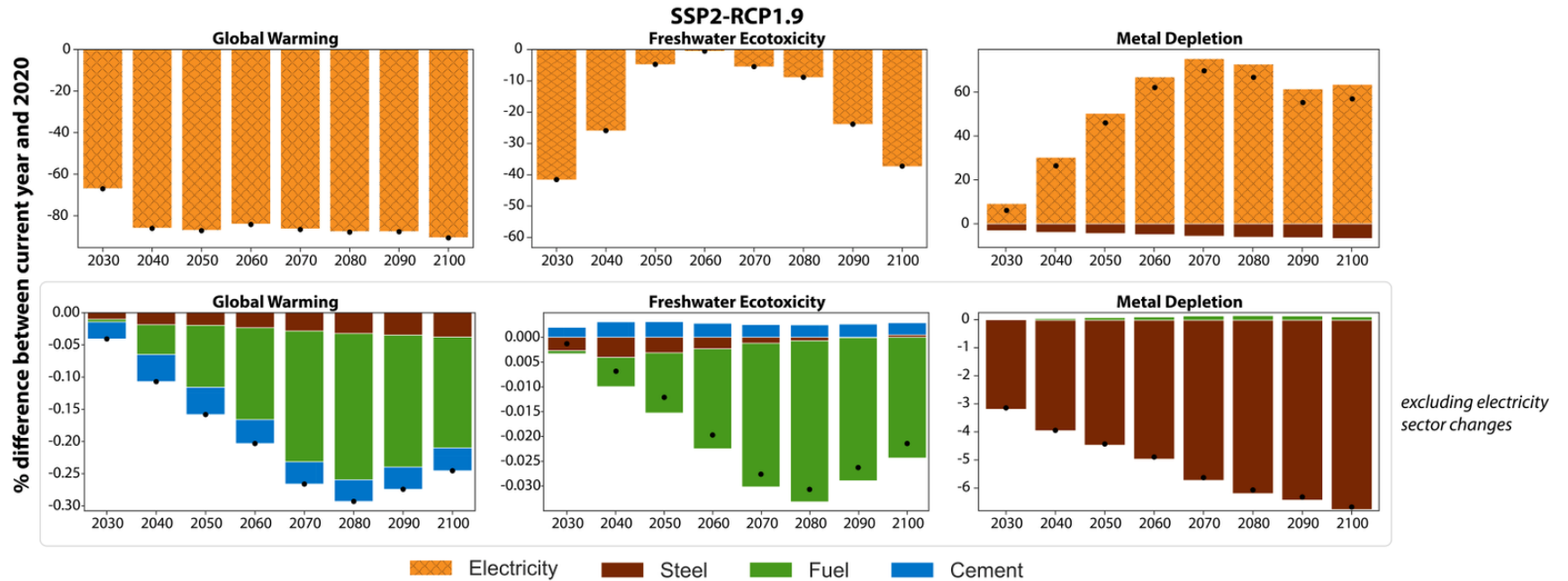
GWP, PEME and SOE reach parity with SMR in 2050 and 2030 for SSP2-RCP2.6 and SSP2-RCP1.9 respectively.

Do not reach parity with SMR for the baseline case.

Impacts such as marine eutrophication, metal depletion and human toxicity increase due to deployment of BECCS and increased used of solar and wind for the decarbonization pathways.



# Specifying the dynamics per sector (background)



## Example: PEME, SSP2-RCP1.9

- Power sector exhibits the largest influence (up to -80%; top left).
- Metal depletion is linked to steel sector dynamics (recycling rates and efficiencies).
- Dynamics for other sectors are still observed, but they do not contribute significantly (<1%; bottom right).
- Land and water impacts link back to transport fuel sector dynamics.

LiAISON computes these results for each technology-scenario combination allowing us to identify:

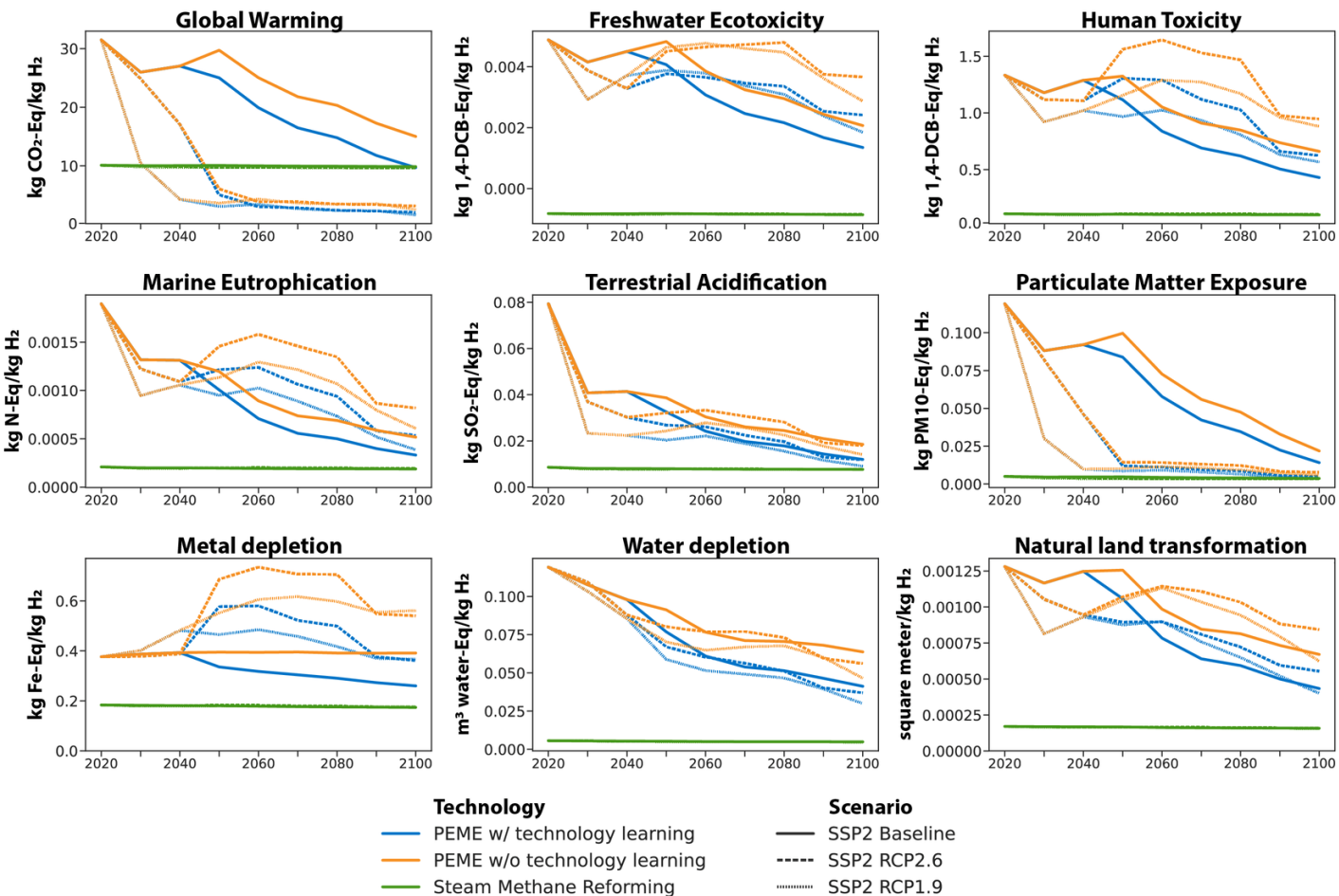
- 1) The influences of the individual sector dynamics;
- 2) Potential tradeoffs and underlying dependencies (e.g., hot spot analysis for power technologies)

Future impacts (ReCiPe; 1 kg H<sub>2</sub>) due to changes in the background superimposed with technology learning at 5% energy efficiency improvement per year

Beyond 2040 electrolysis is deployed globally on a large-scale, driving learning-by-doing improvements.

Learning-by-doing, further reduces impacts over time.

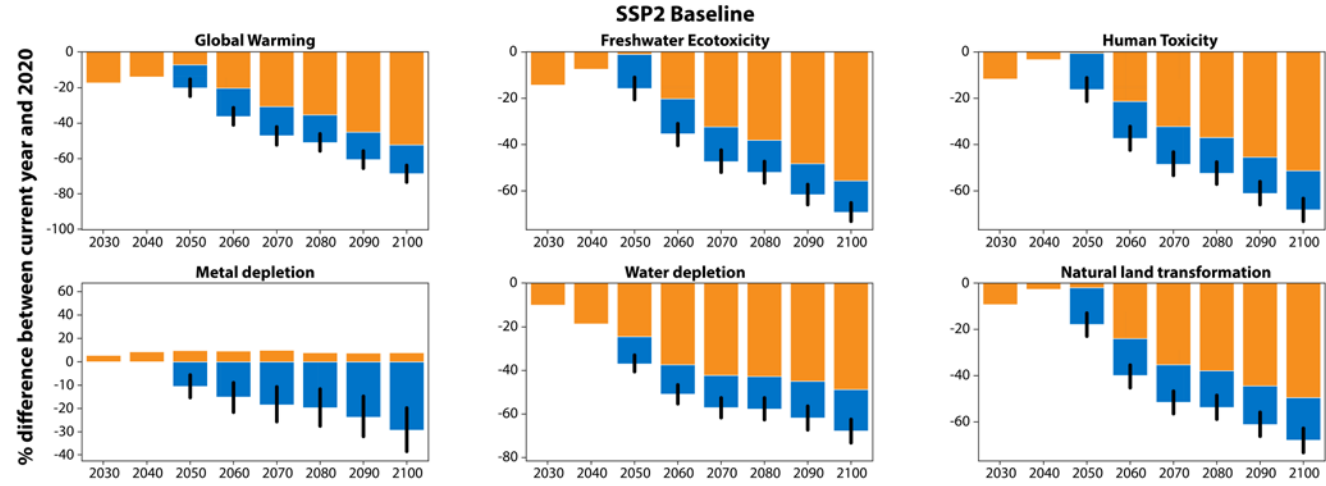
Benefits are largest for metrics that do not drop due to background changes, i.e., smaller benefits for GWP<sub>100</sub> in mitigation scenarios, larger ones for impacts that rise, e.g., eutrophication.



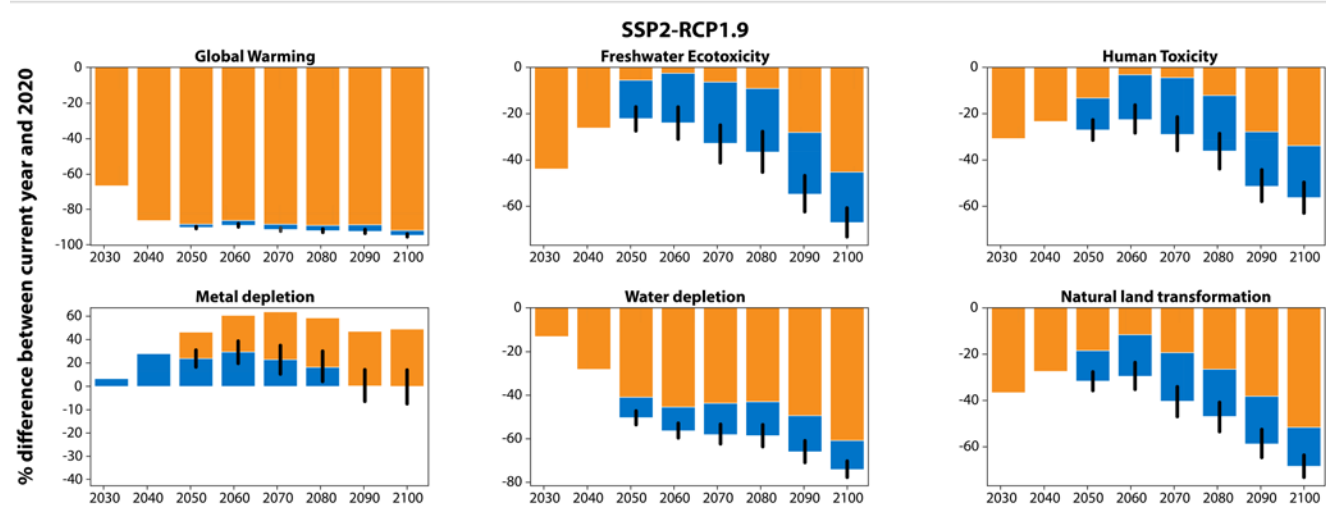
R&D  
influence  
vs. systems  
context

PEME  
ReCiPe  
SSP2-  
Baseline

PEME  
ReCiPe  
SSP2-  
RCP1.9



If the electricity grid evolves according to RCP1.9 pathway, PEME will have less contribution to decarbonization







































Thus, evolving the background rather than improving PEME efficiency will have more impact on decarbonization.

Only background      Background + foreground dynamics (learning)



# LCA tool comparison: Computation time

Functionality	Excel	openLCA	SimaPRO	LiAISON	
Collection of inventory data					
Foreground production system building					 NA
Linking with background inventory					 < 5 mins
Monte Carlo analysis					 < 1 hour
Regional sensitivity analysis					 > 1 hour
Technological learning sensitivity analysis					
Prospective LCA					
Plotting					

\*Assuming standard LCA with < 5 foreground processes, 500 MC runs, 3 regions

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