

Modeling multisectoral interactions in the TEMOA Canadian Open Energy Model (CANOE)

PIs: Sylvia Sleep (U of C), Daniel Posen (U of T)

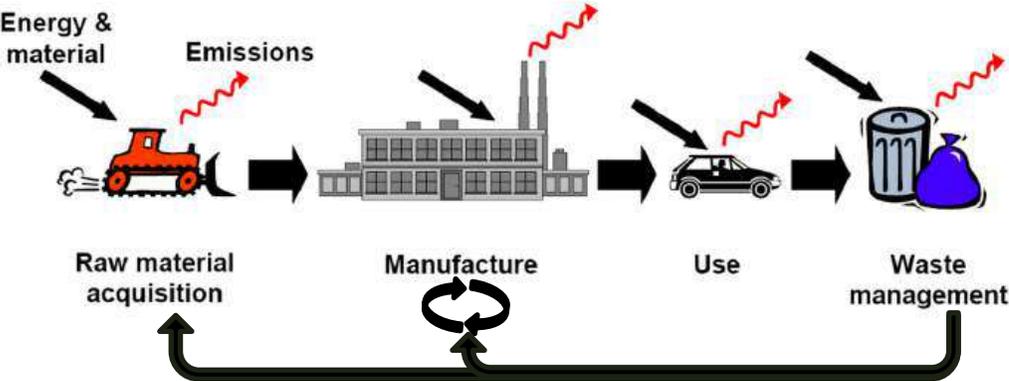
Co-leads: Sean McCoy (U of C), Sara Hastings Simon (U of C), Joule Bergerson (U of C / ND), Andrew Leach (U of A), Juan Moreno Cruz (Waterloo), Heather MacLean (U of T)

Partners: Canada Energy Regulator, Natural Resources Canada, ESMIA, Pollution Probe, Sutubra Consulting

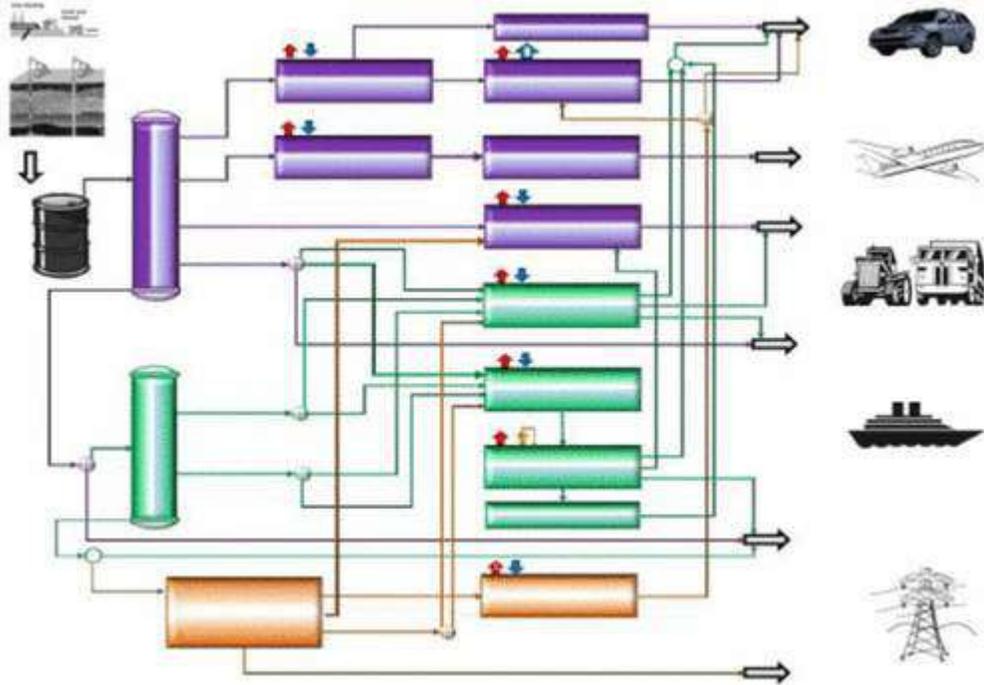


Our team comes from diverse areas adjacent to energy modeling

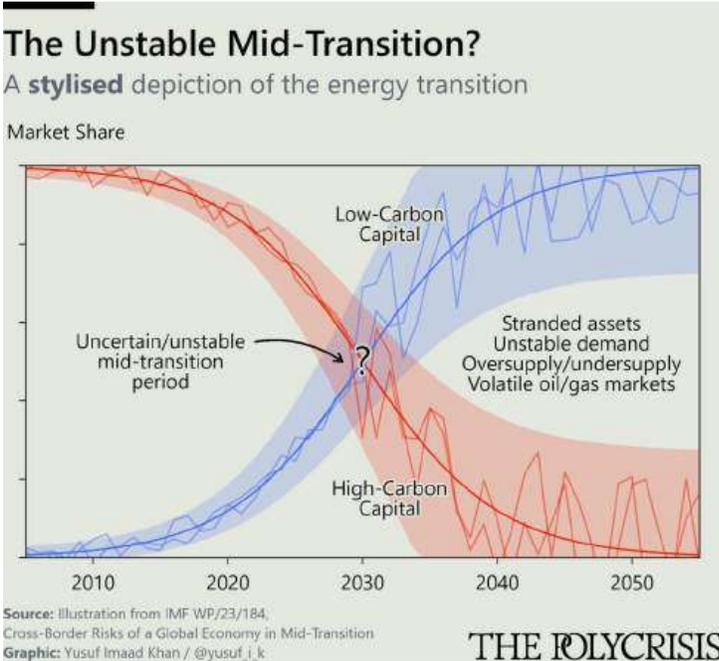
Life cycle assessment and Material flow analysis



Detailed technology modeling (e.g., refineries)



Abella and Bergerson 2012; Prelim model (updated 2022)



CANOE is a capacity expansion ESOM

Classification

- Energy systems optimisation model (LP)
- Least-cost capacity expansion
- Built on **new Temoa v4.0**

Spatial resolution

- **National-provincial** (one region per province)

(Typical) Temporal resolution

- 5-year periods, **2025 to 2050** (maybe 2070)
- Representative days with **hourly resolution**

Supply sectors

- Chemical fuels (liquid, gaseous fuels)
- Electricity

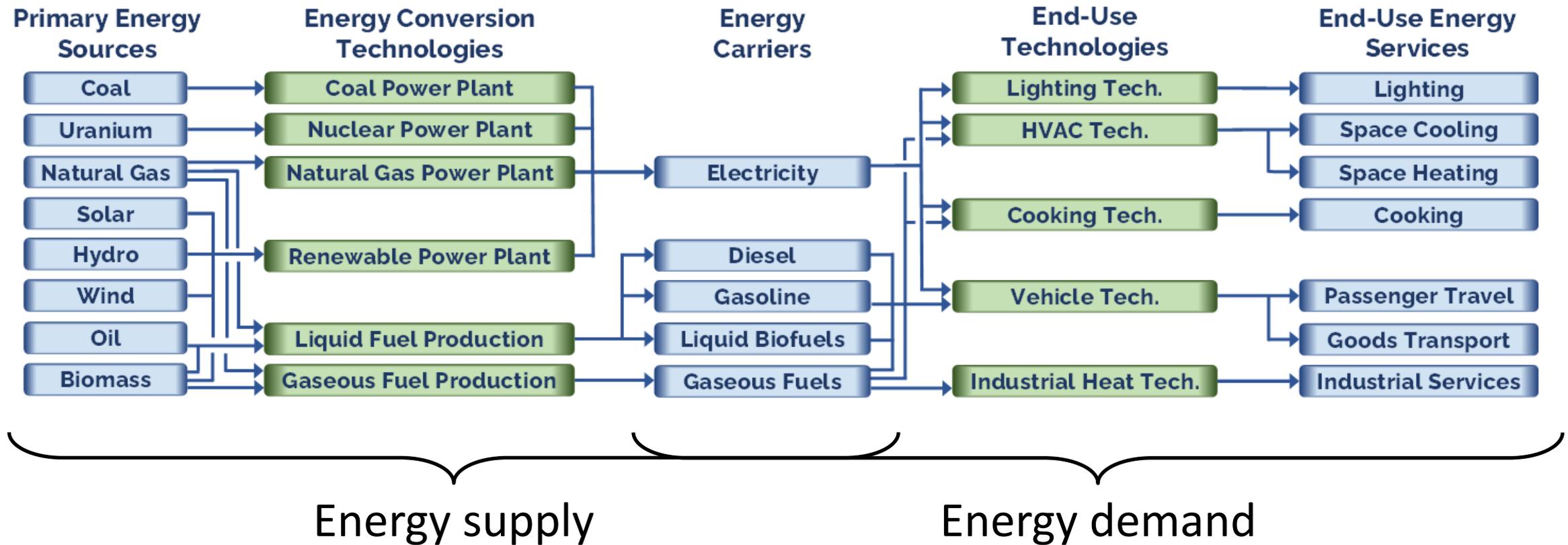
Demand sectors

- Industry
- Transportation
- Buildings

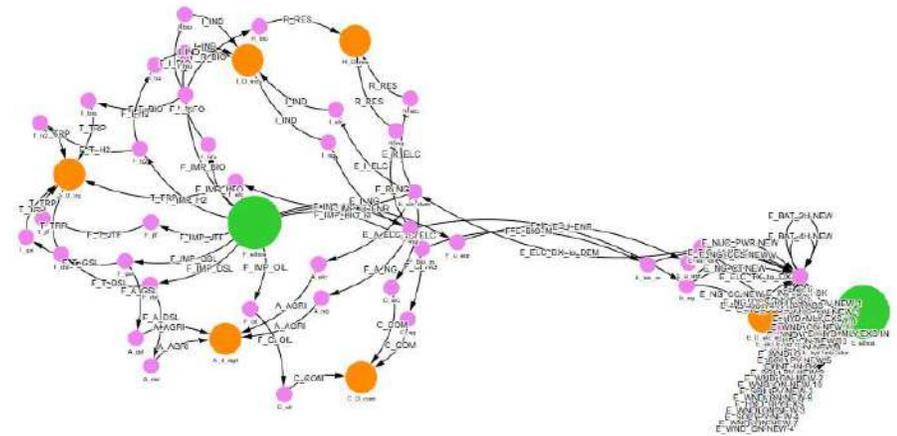
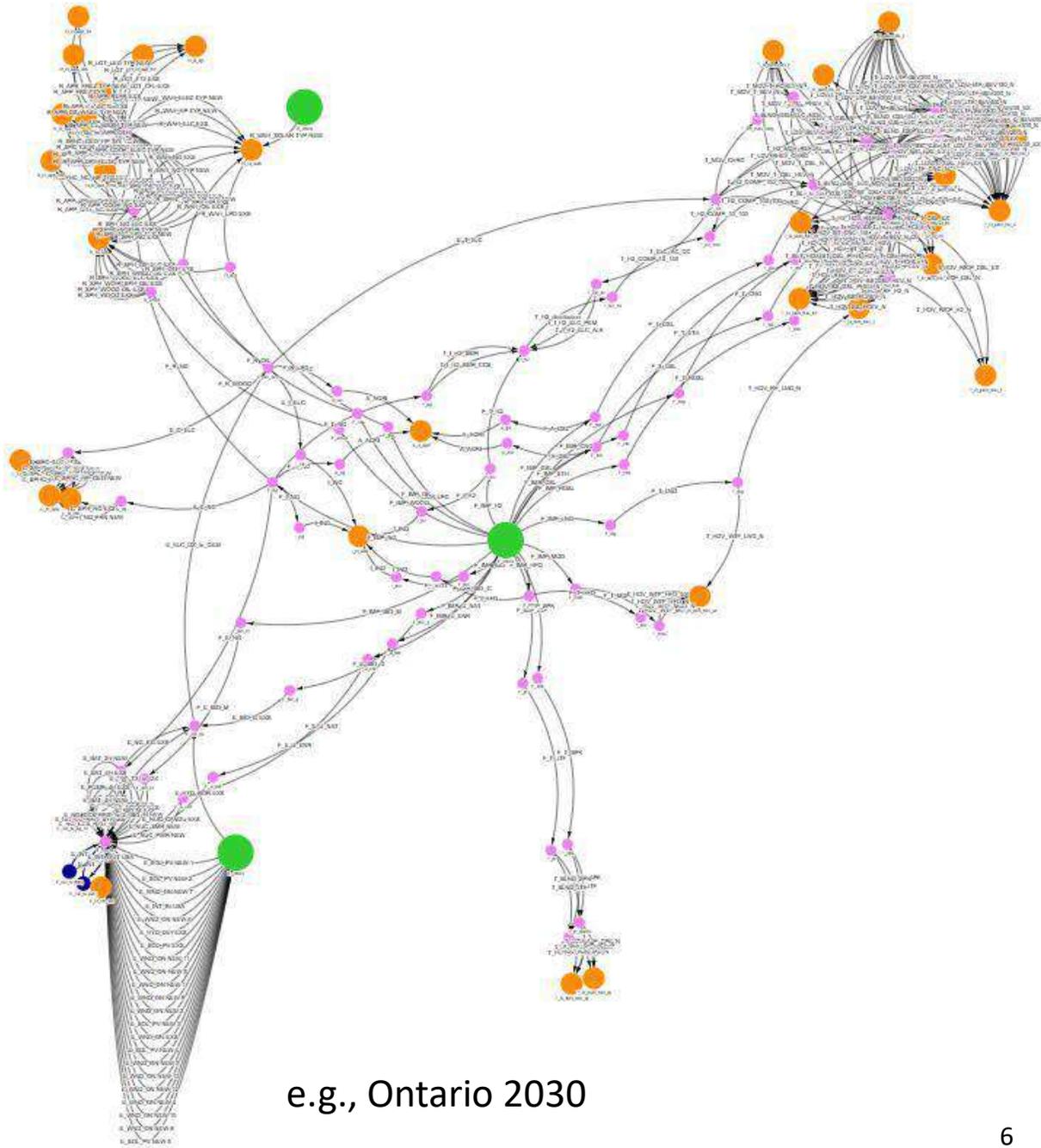
(Also carbon removal, e.g. direct air capture)

Systems representation – CANOE

(demonstrative subset of systems)



Snapshots of the real network



Building CANOE

Managing an open source, multi-sectoral ESOM

Problem

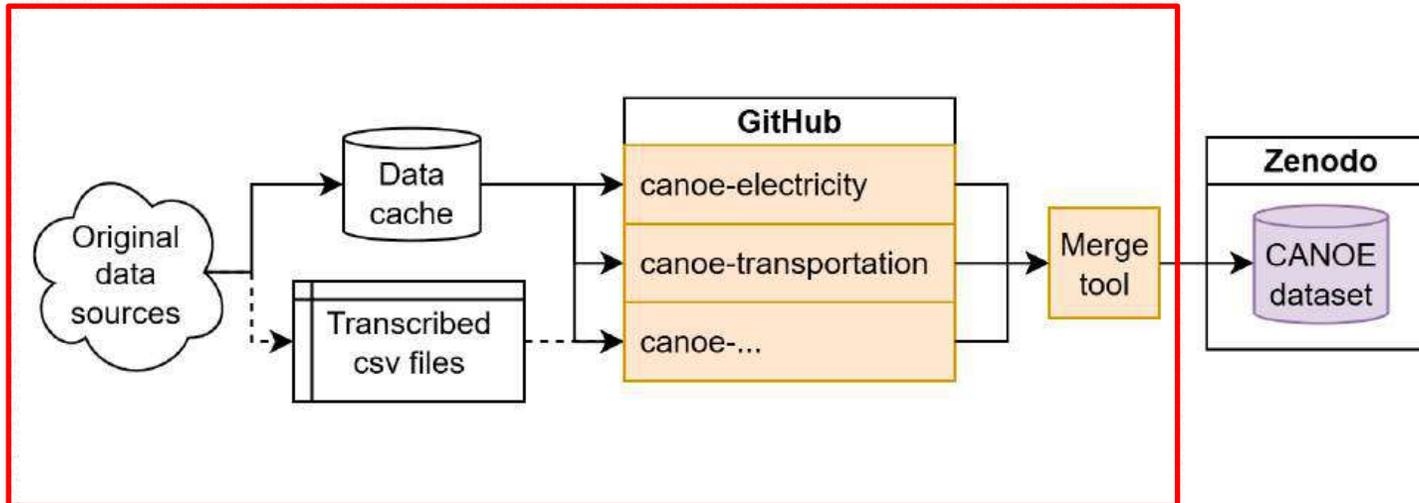
Huge amount of data. How do we mitigate input errors, keep track of assumptions and document all processing steps?

Solution

Python. Build the database via code, never manually edit data. Publish Python scripts on GitHub for transparency and collaboration.

Building CANOE

Backend



 Python-based

 SQLite



Mastermind: Davey Elder



Building CANOE

Managing an open source, multi-sectoral ESOM

Problem

Huge amount of data. How do we mitigate input errors, keep track of assumptions and document all processing steps?

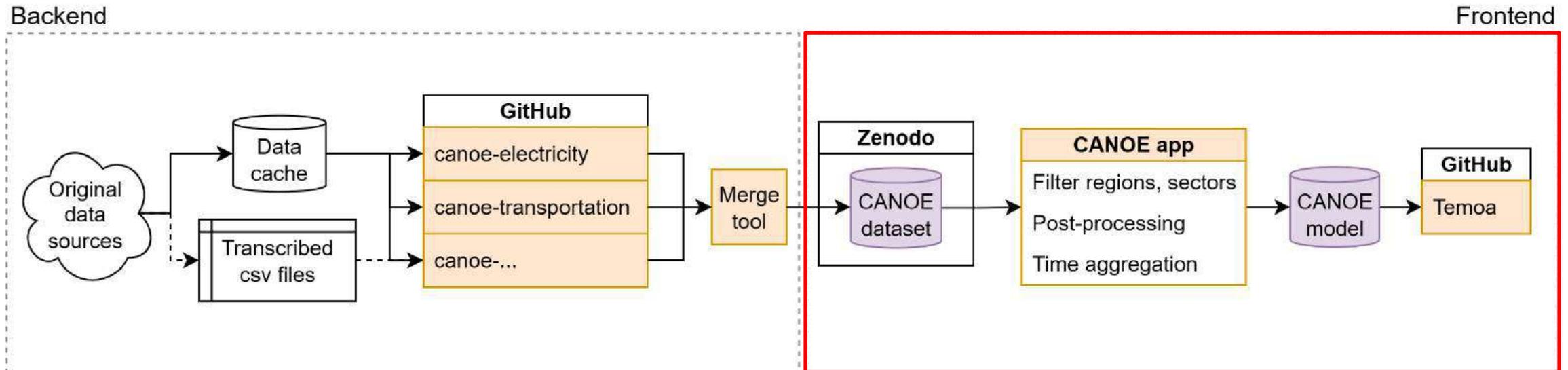
Model complexity. How do we model Canada's entire energy system but also tackle smaller research questions and explore uncertainty?

Solution

Python. Build the database via code, never manually edit data. Publish Python scripts on GitHub for transparency and collaboration.

Modularity. Construct each region, sector, and feature modularly with unique data IDs. Filter by these IDs to model only what you're interested in.

Building CANOE



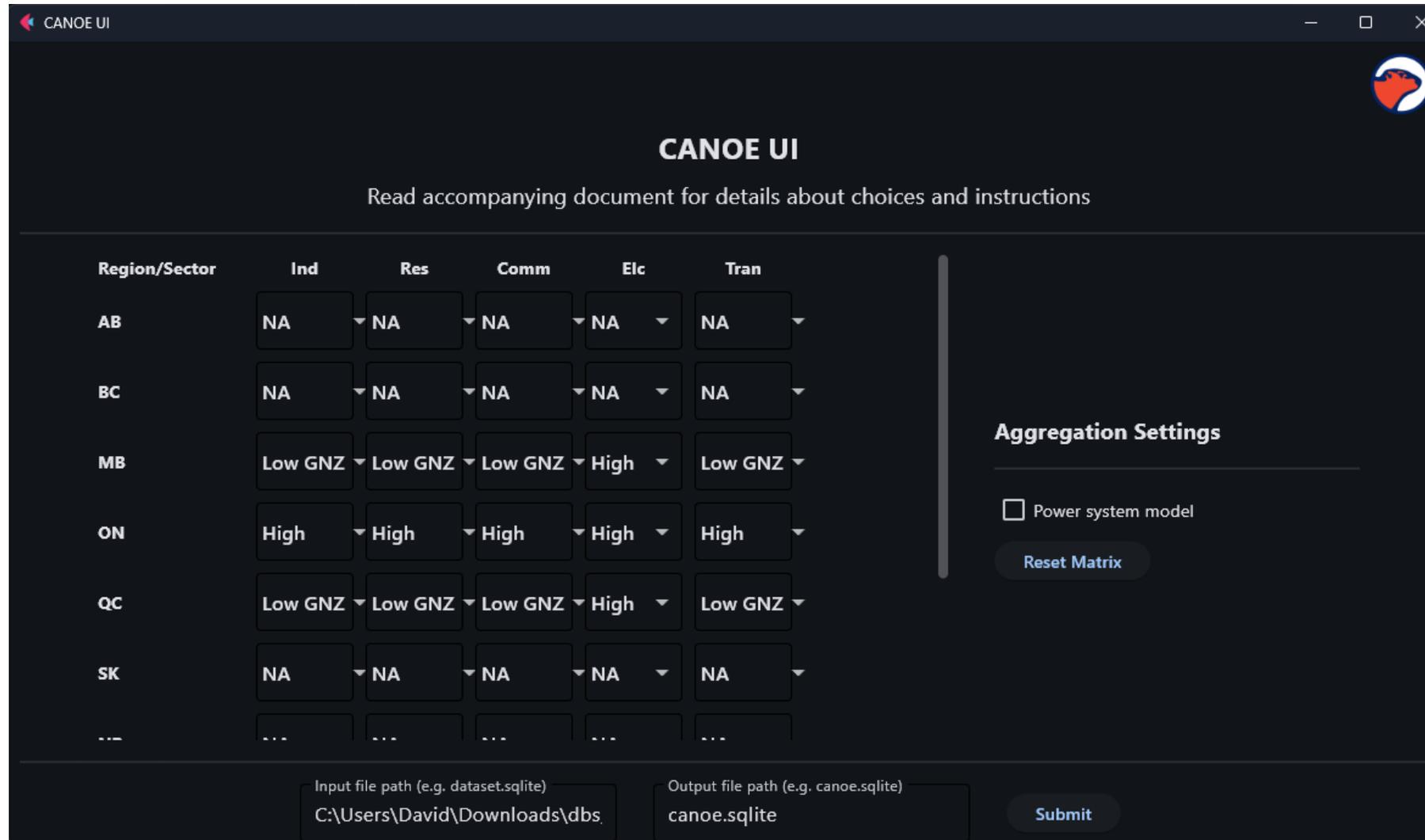
-  Python-based
-  SQLite



Mastermind: Davey Elder



Model resolution selection



The screenshot shows the CANOE UI interface. At the top, it says "CANOE UI" and "Read accompanying document for details about choices and instructions". Below this is a table for selecting model resolution for different regions and sectors. The table has columns for "Region/Sector", "Ind", "Res", "Comm", "Elc", and "Tran". The rows are labeled AB, BC, MB, ON, QC, SK, and "...". Each cell in the table contains a dropdown menu with a selection. To the right of the table is the "Aggregation Settings" section, which includes a checkbox for "Power system model" and a "Reset Matrix" button. At the bottom, there are input fields for "Input file path (e.g. dataset.sqlite)" and "Output file path (e.g. canoe.sqlite)", and a "Submit" button.

Region/Sector	Ind	Res	Comm	Elc	Tran
AB	NA	NA	NA	NA	NA
BC	NA	NA	NA	NA	NA
MB	Low GNZ	Low GNZ	Low GNZ	High	Low GNZ
ON	High	High	High	High	High
QC	Low GNZ	Low GNZ	Low GNZ	High	Low GNZ
SK	NA	NA	NA	NA	NA
...

Aggregation Settings

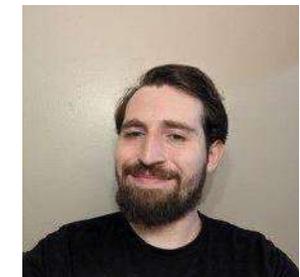
Power system model

Reset Matrix

Input file path (e.g. dataset.sqlite)
C:\Users\David\Downloads\dbs

Output file path (e.g. canoe.sqlite)
canoe.sqlite

Submit



David Turnbull



New model features in CANOE / Temoa

Embodied carbon

Full life cycle emissions accounting

Endogenous material flows

Critical minerals

Bulk materials

EV charging

Charging infrastructure for electric vehicles

Charging behaviours (hourly charging demand)

Oil refineries

Constraining realistic petroleum product fractions

Seasonal storage

Hydroelectric reservoirs

E-fuels, hydrogen

Endogenous technological learning

Declining capital costs with greater deployment

Survival curves

Realistic technology survivorship

Growth rate constraints

Realistic technology deployment and phaseout

Embodied carbon

Full lifecycle emission accounting



Cameron Wade

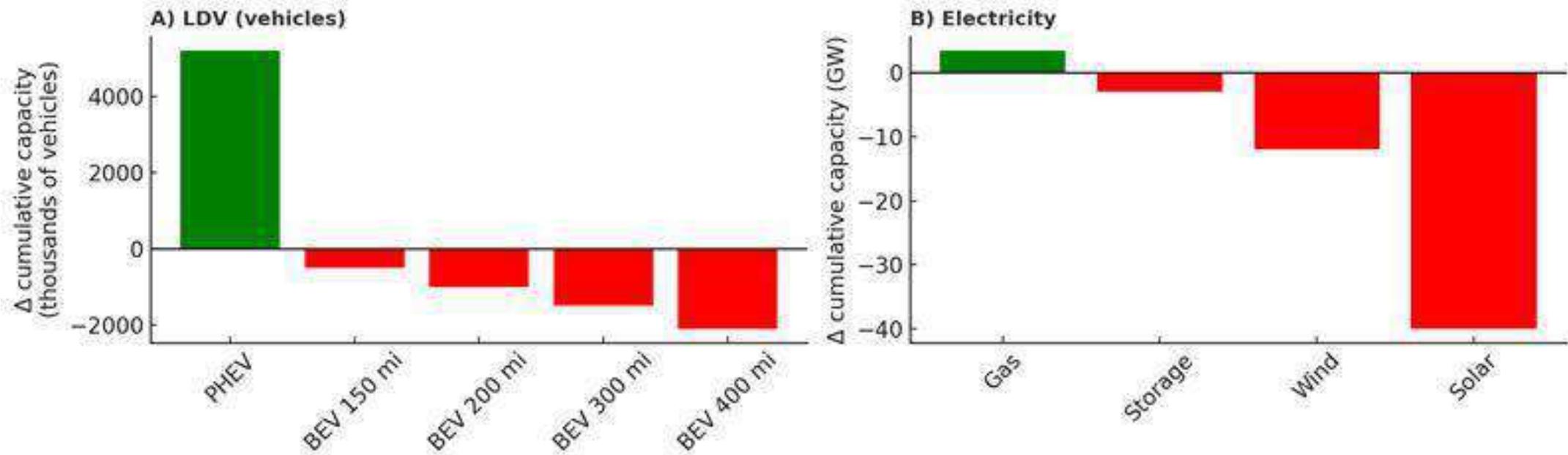
- Most frameworks account for operational emissions, indexed to energy flows.
- We now also account for **embodied emissions, indexed to construction and end of life** of technology capacity.
- Accounts for all upstream and downstream lifecycle emissions.
- Not just emissions. Can also be land use, job creation, etc.



Embodied carbon

Change in multiple sectors when accounting for embodied emission costs

Capacity Changes vs Reference (Embodied Scenario)



Shahid Hossaini

Endogenous material flows

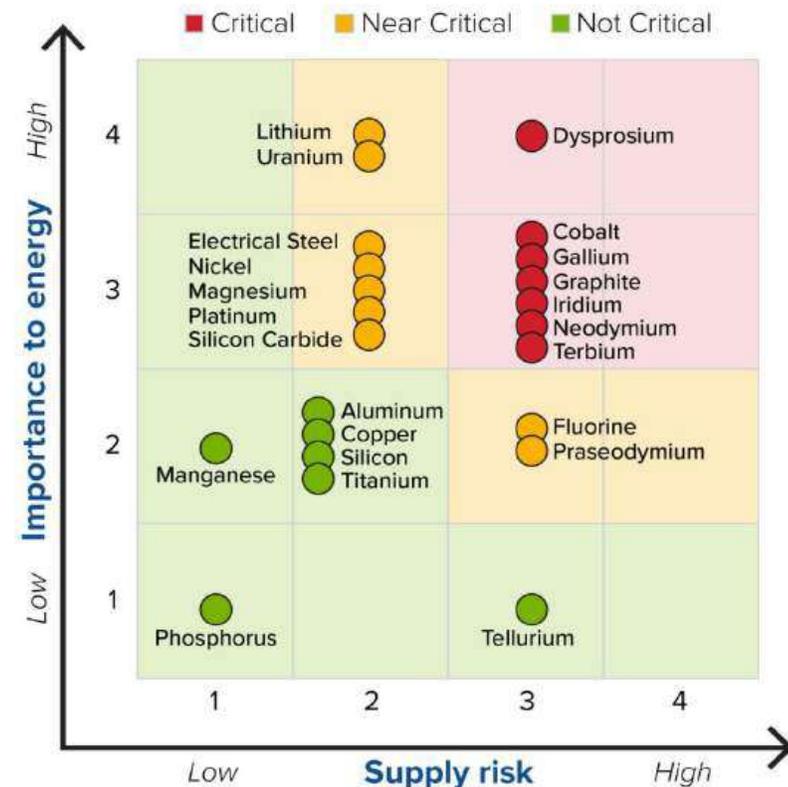
Endogenous process flows indexed to technology capacity

- Connect construction and end of life of technologies to commodity flows in the process network.
- Track **supply chains** and recycling for **bulk materials** (steel, concrete, ...), **critical minerals** (lithium, cobalt, ...), or components (e.g., **batteries**).
- Endogenously model energy consumption of these supply chains.



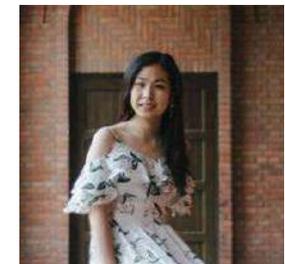
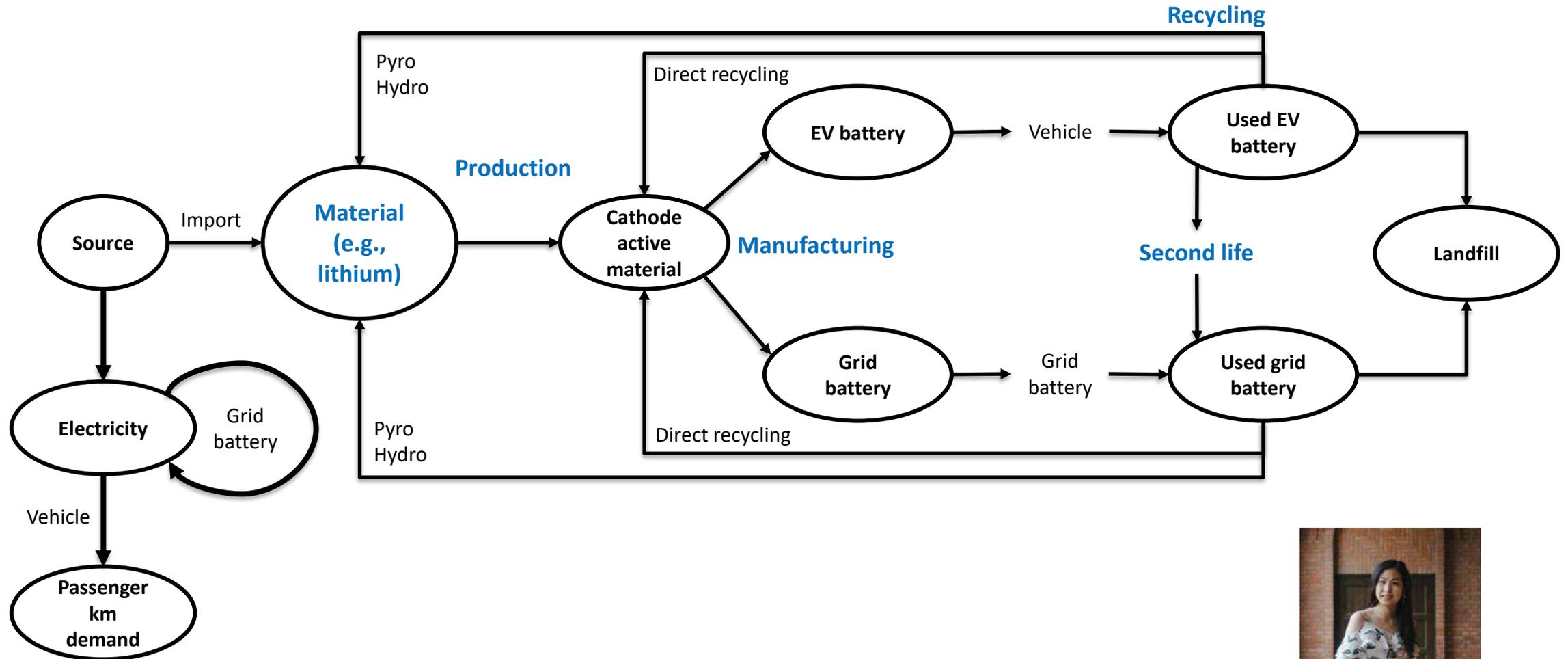
Davey Elder

SHORT TERM 2020-2025



Endogenous material flows

Application: battery supply chains

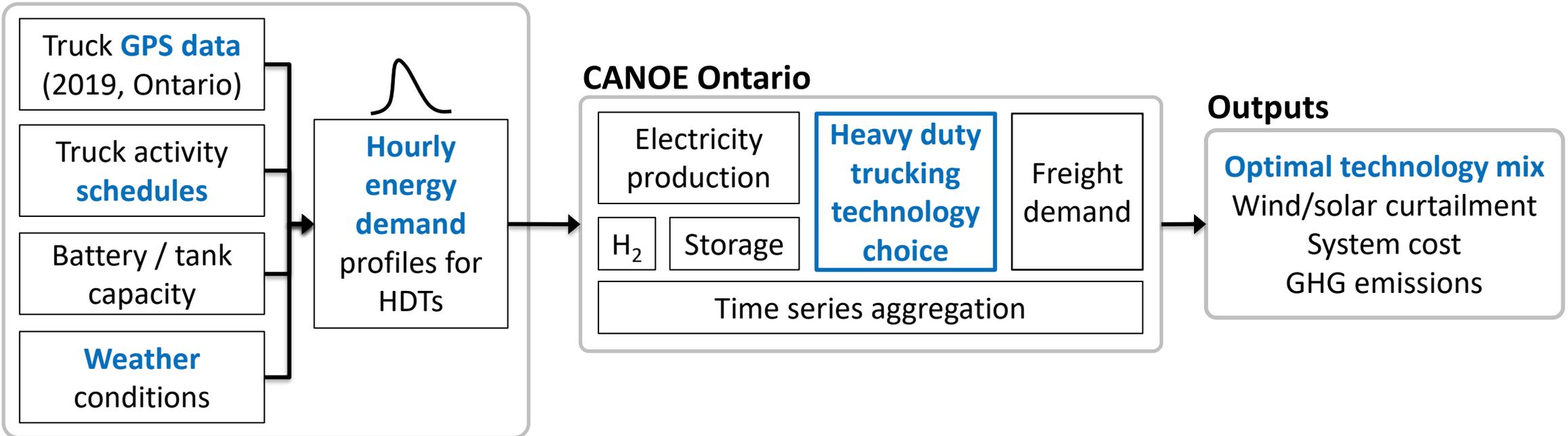


Dijuan Liang

Heavy duty EV truck charging

Demand profiles from GPS data + Charging infrastructure technology choice

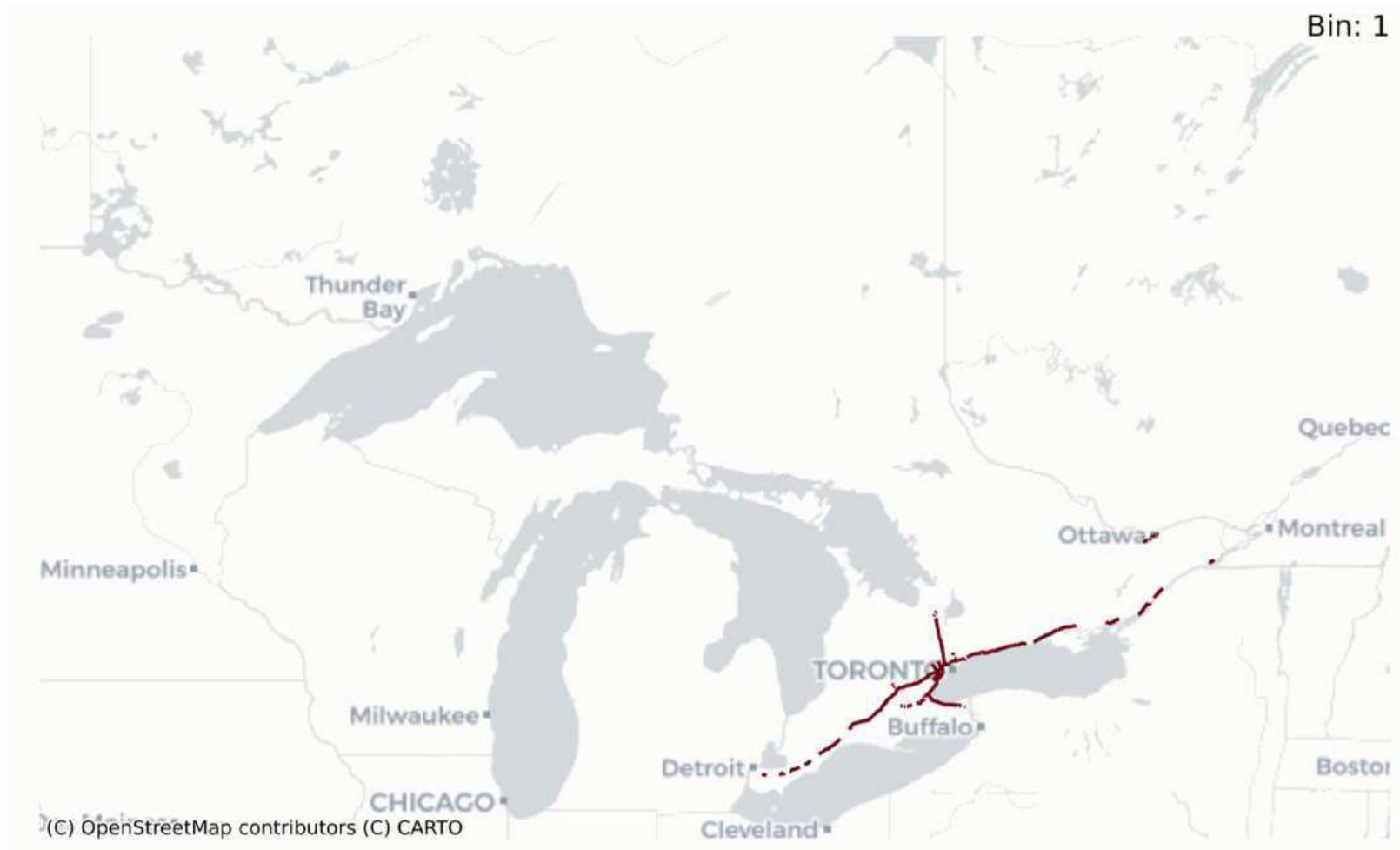
Simulate charging profiles



Sebastian Villada Rivera

Heavy duty EV truck charging

Spatially explicit E-road expansion

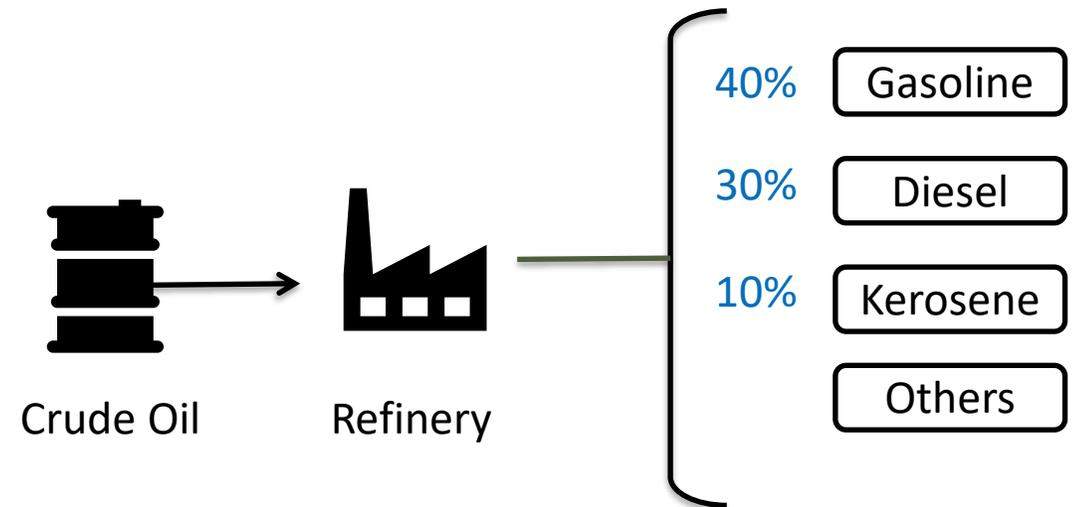


Lih Wei Yeow

Oil refineries

Constraining realistic product fractions

- Oil refineries are designed to produce a particular ratio of output products.
- Can we phase out one fuel at a time? (e.g., gasoline but not diesel)
- Model oil refineries to explore **constraints due to refinery design and operation.**



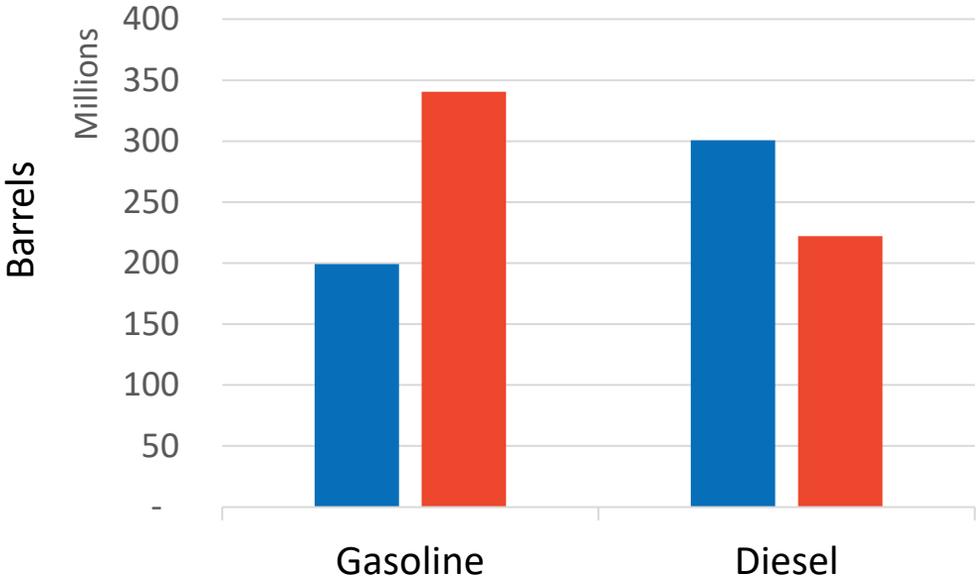
Thiago Rodrigues

Oil refineries: constraints on product mix may slow light-duty vehicle electrification (or speed up truck electrification)

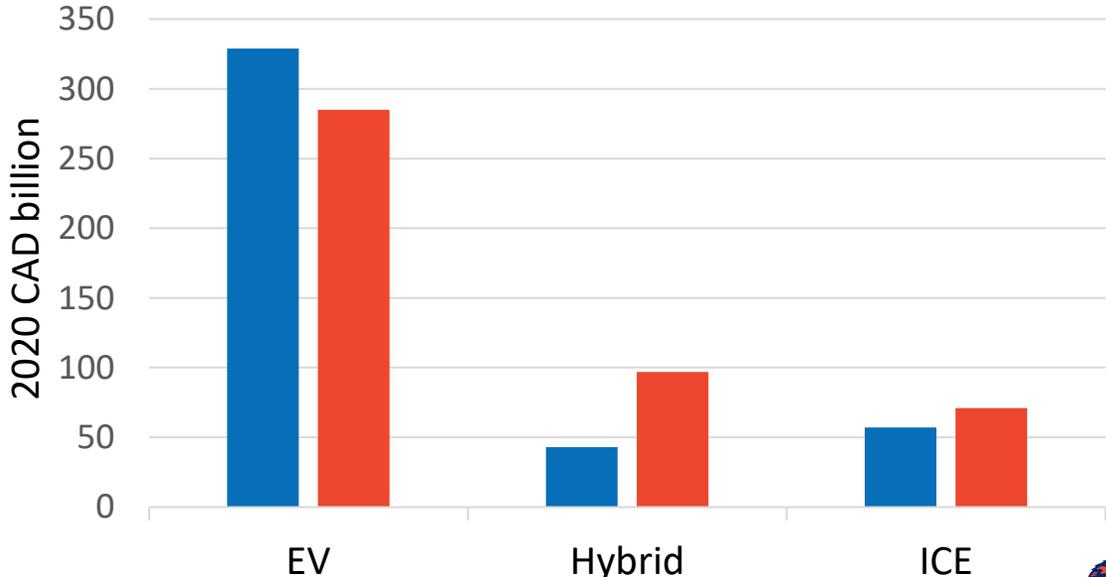
Testing results: before and after forcing existing product ratios

Scenario: **Free product ratios** ■ **Fixed product ratios** ■

Fuel consumption (2025 – 2050)



Investments in new vehicles (2025 – 2050)



Seasonal storage

Long duration storage in a time-aggregated model

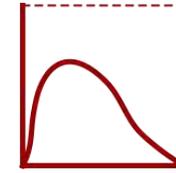
- Many energy models use **seasonal time slicing or representative days** to reduce the time dimension.
- Broken chronology leads to irrational inter-seasonal storage behaviour.
- We reconstruct the original sequence of our base weather year from these representative days. Restoring **rational seasonal storage** behaviour.
- Applies to large-reservoir hydroelectric, E-fuels, hydrogen.

Cameron Wade

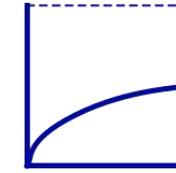


Davey Elder

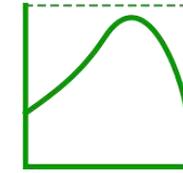
Representative days



3 days



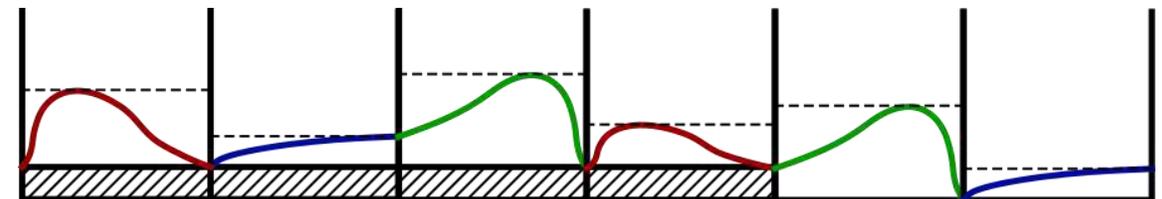
2 days



4 days

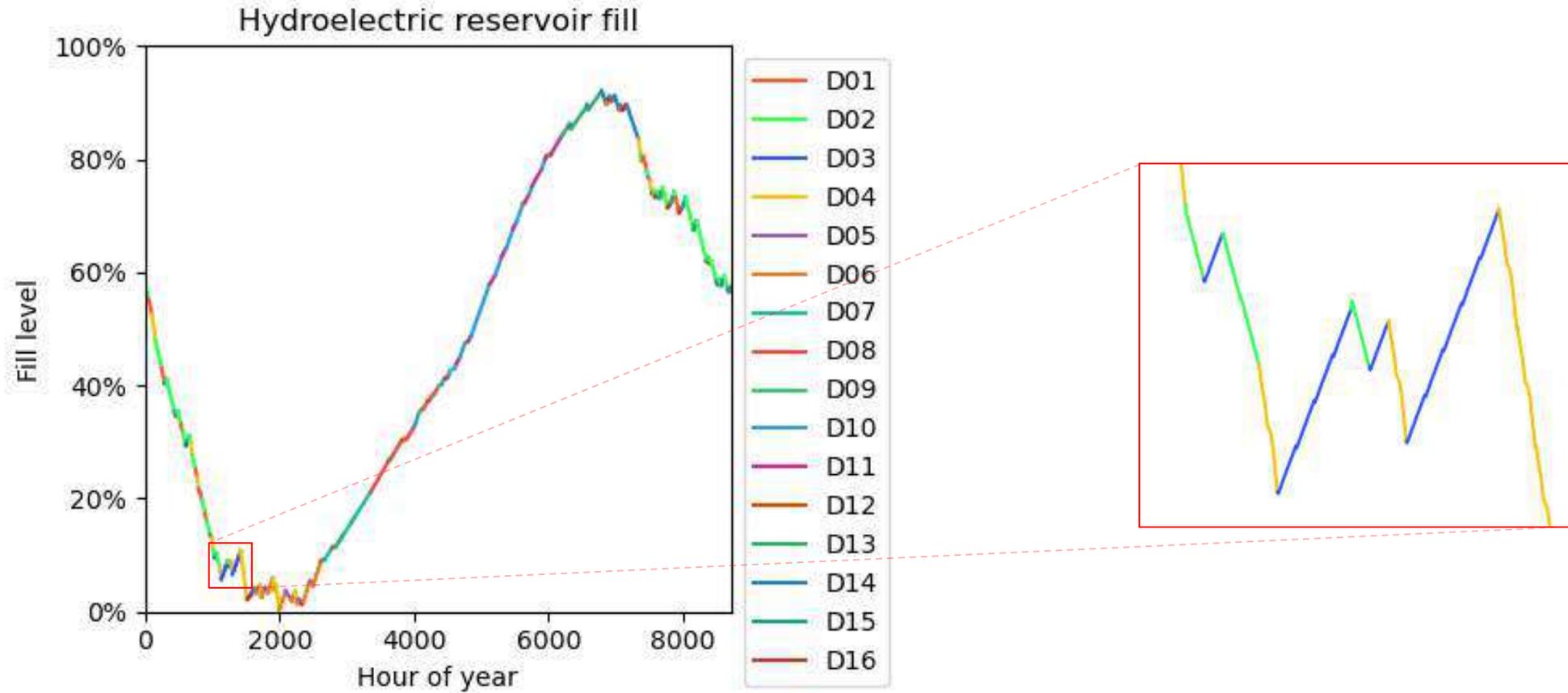
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Superimposed

Full-year seasonal storage levels



Seasonal storage

Example: Sixteen representative days. Hydroelectric reservoirs in Manitoba, 2050.



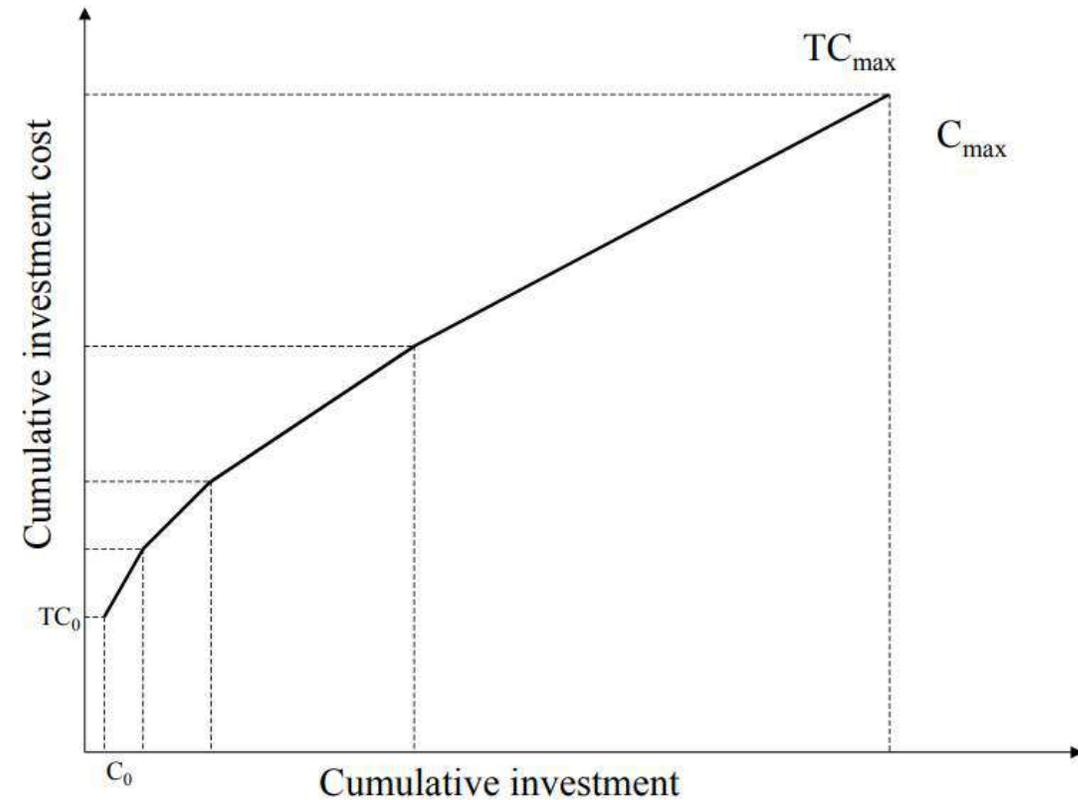
Endogenous technological learning

Declining capital costs with increasing cumulative production



Davey Elder

- New technology costs usually decline over time due to economies of scale and innovation.
- We can **endogenise declining capital costs** using a mixed integer linear (MILP) formulation (based on TIMES).
- Our implementation is flexible to summation over regions and groups of technologies.

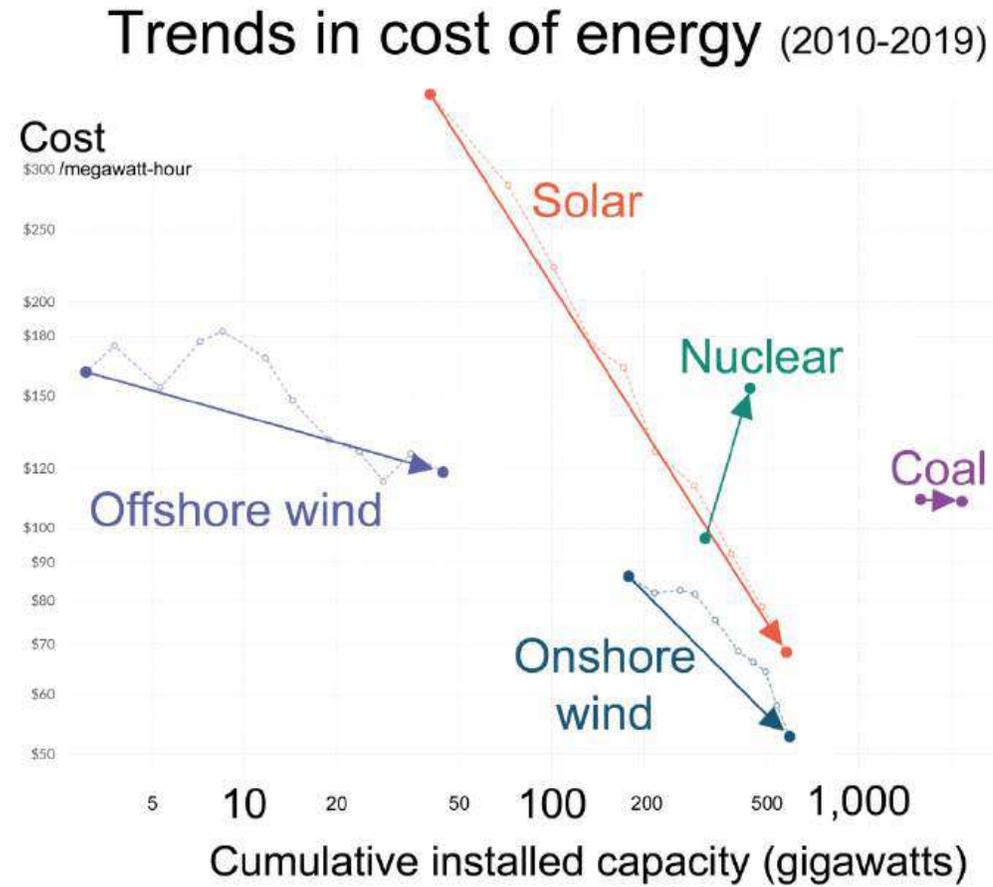


Endogenous technological learning

Application: emerging technologies

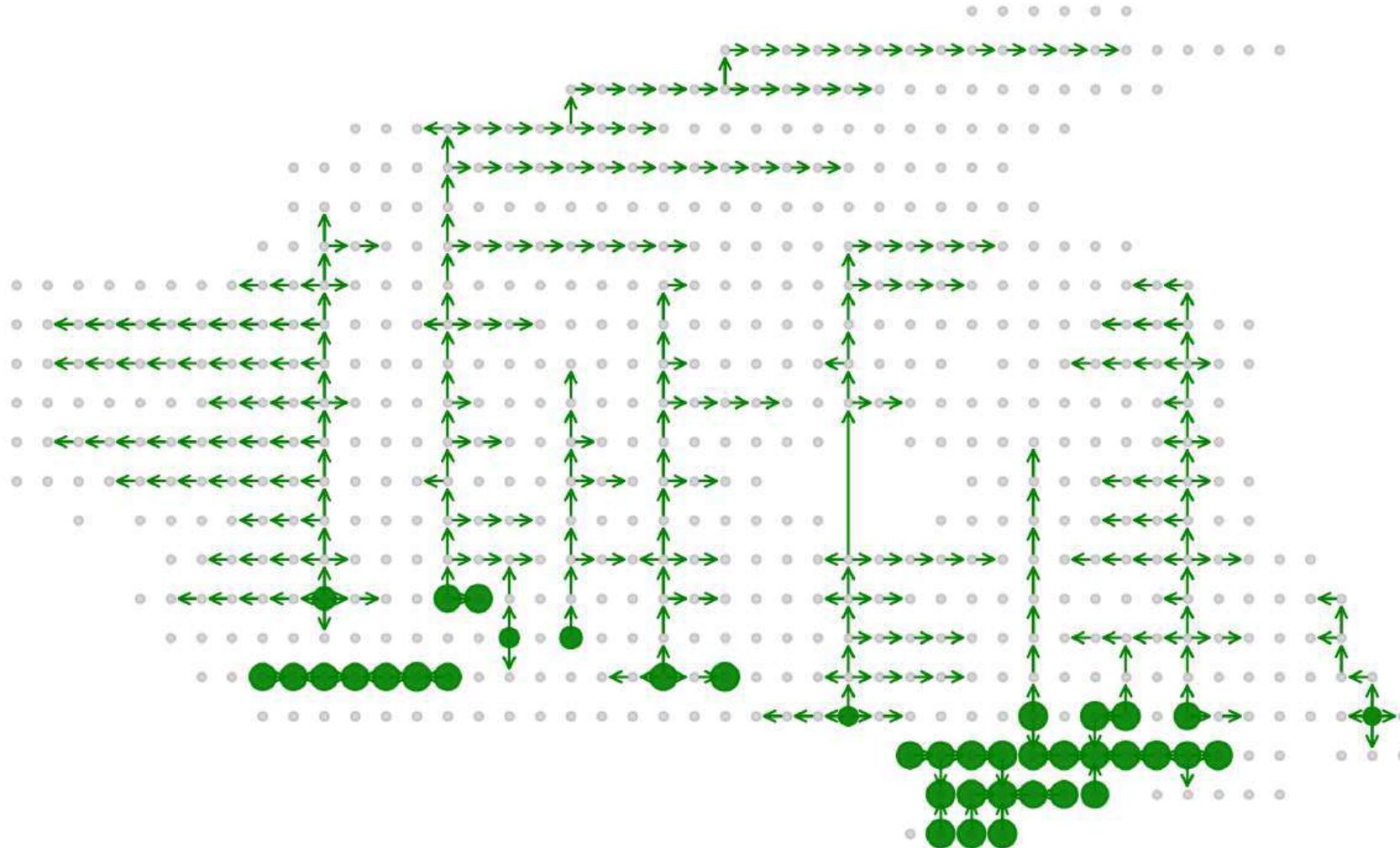


Davey Elder



Geospatial optimization with economies of scale

Application: spatially-explicit infrastructure expansion



Thiago Rodrigues

Uncertainty tools (Temoa)

Monte Carlo

Samples distributions of input parameters.

Outputs distributions of outcomes.

Method of Morris

Global sensitivity analysis.

Ranks sensitivity of input parameters.

Modelling to Generate Alternatives (MGA)

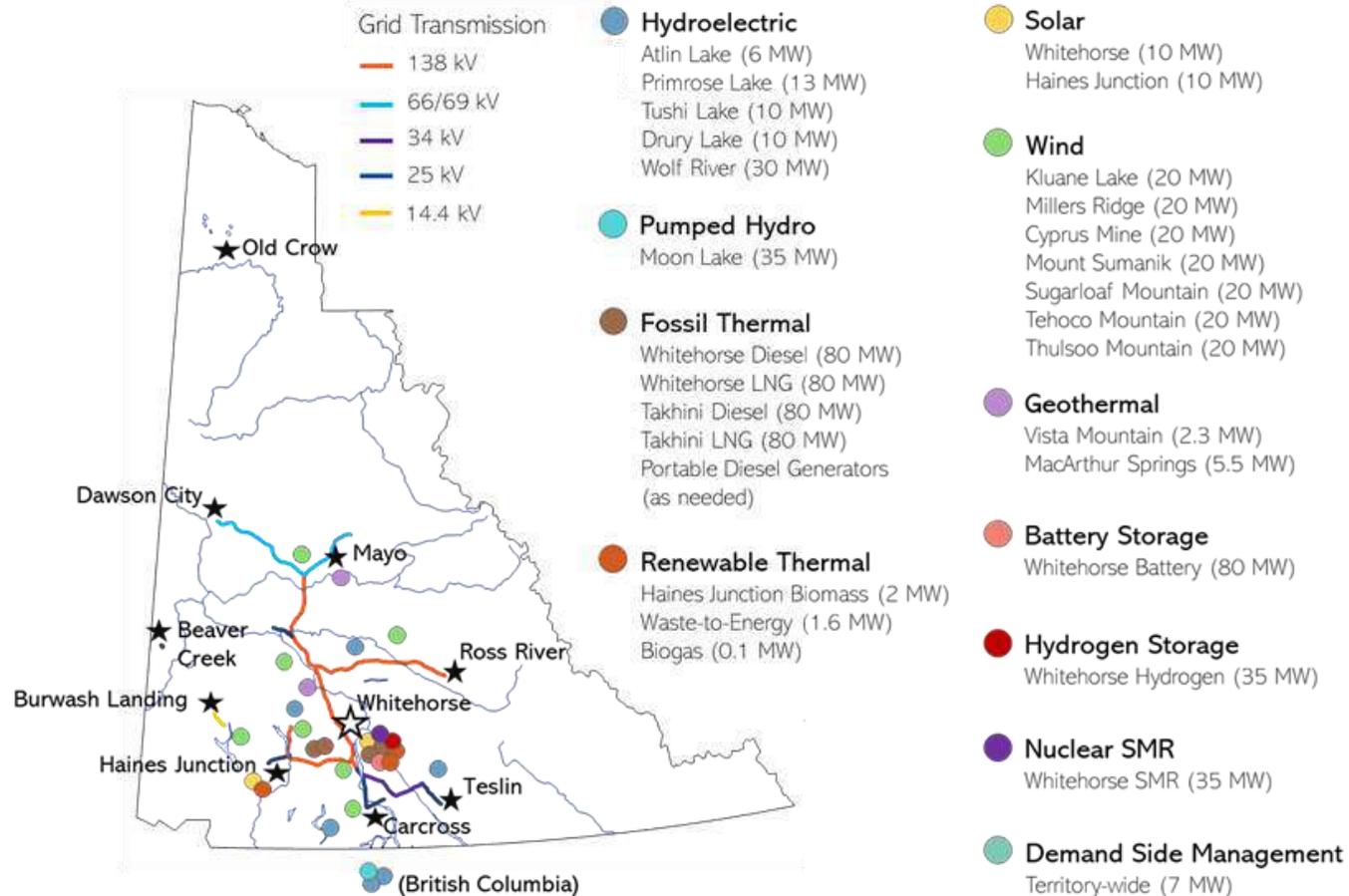
Maximally-different near-optimal solutions.

Uncertainty applications

Monte Carlo – Robust decision making for Yukon’s power system



Chris Fitzgibbon

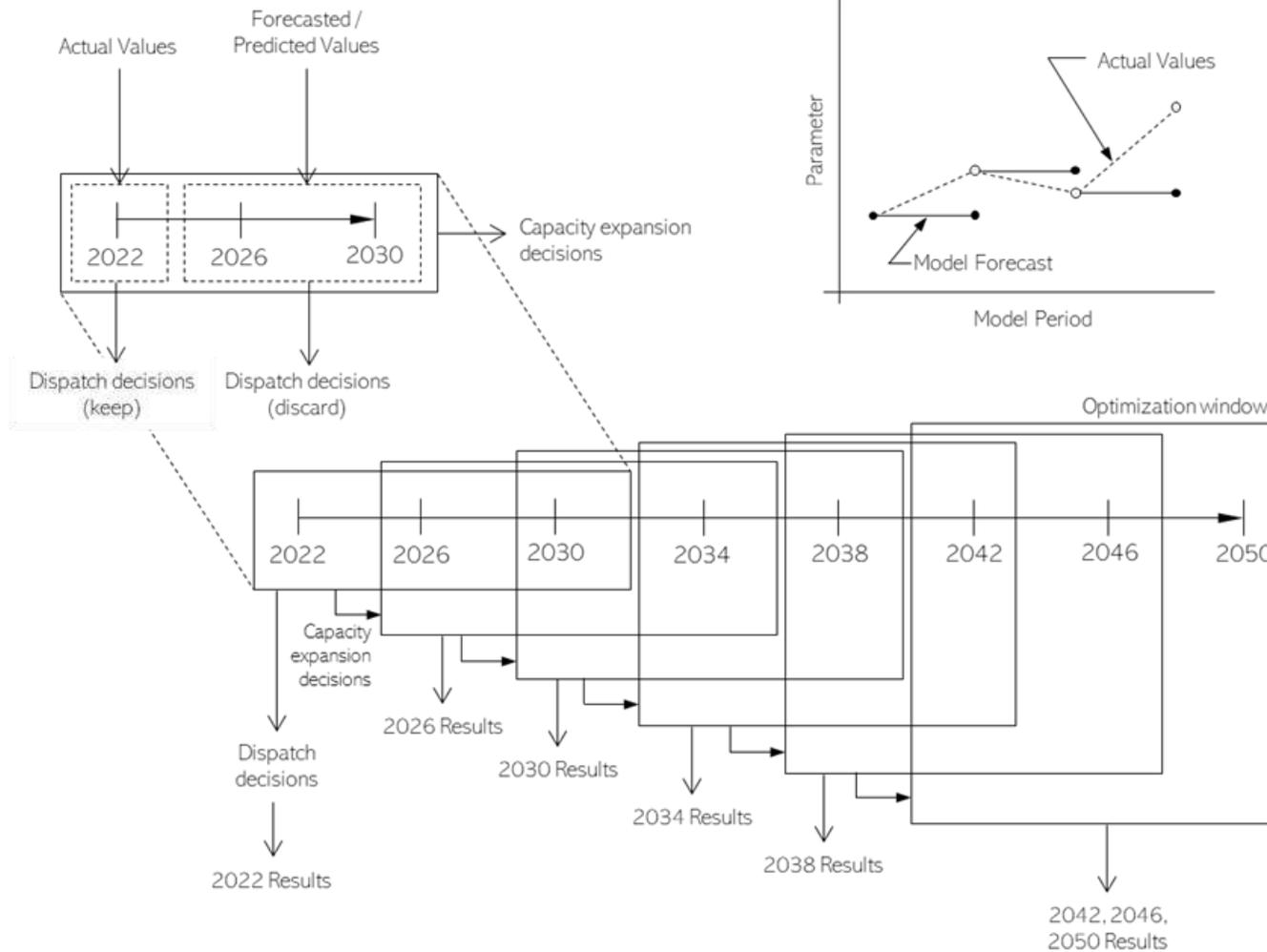


Uncertainty applications

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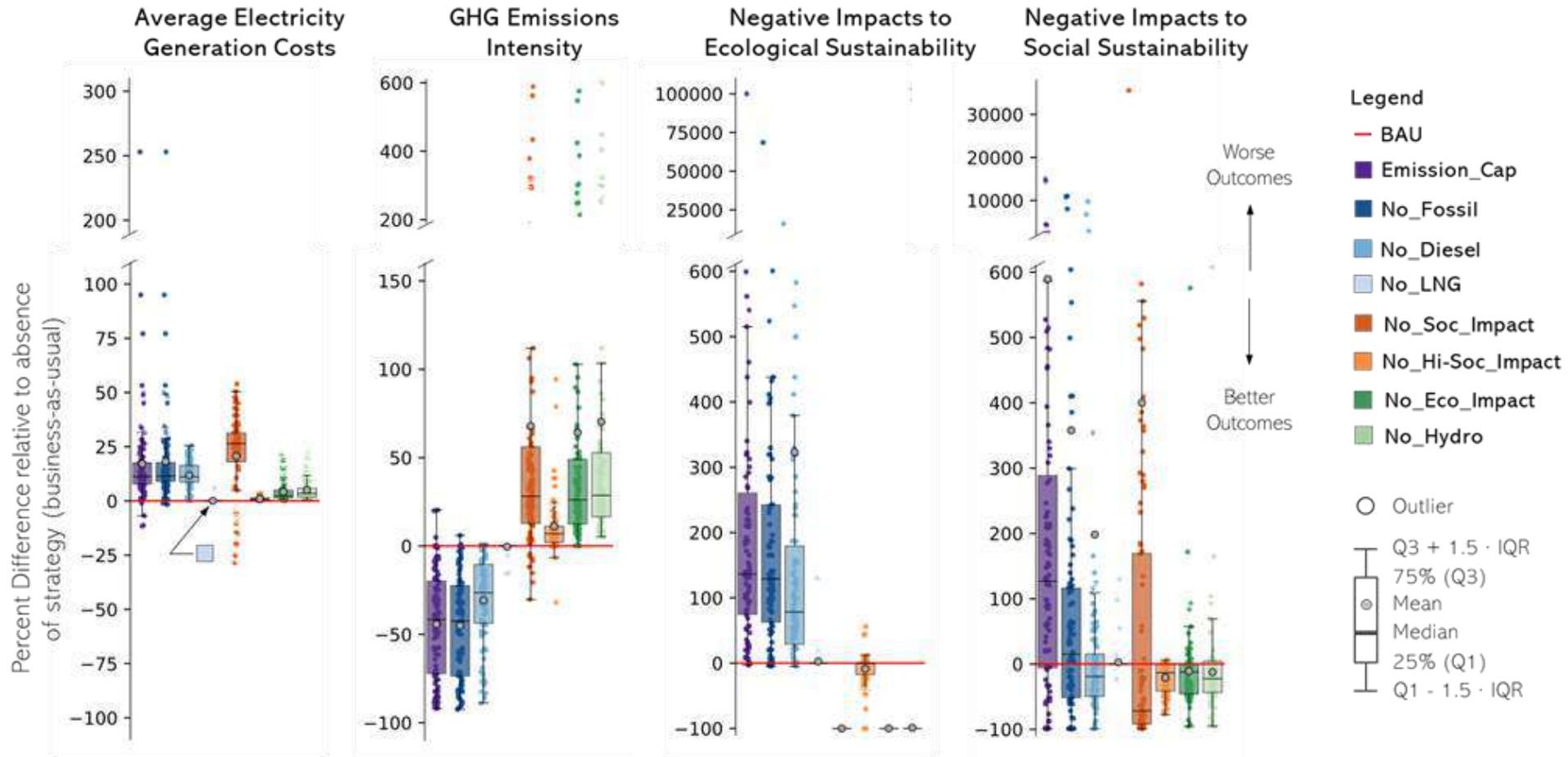


Uncertainty applications

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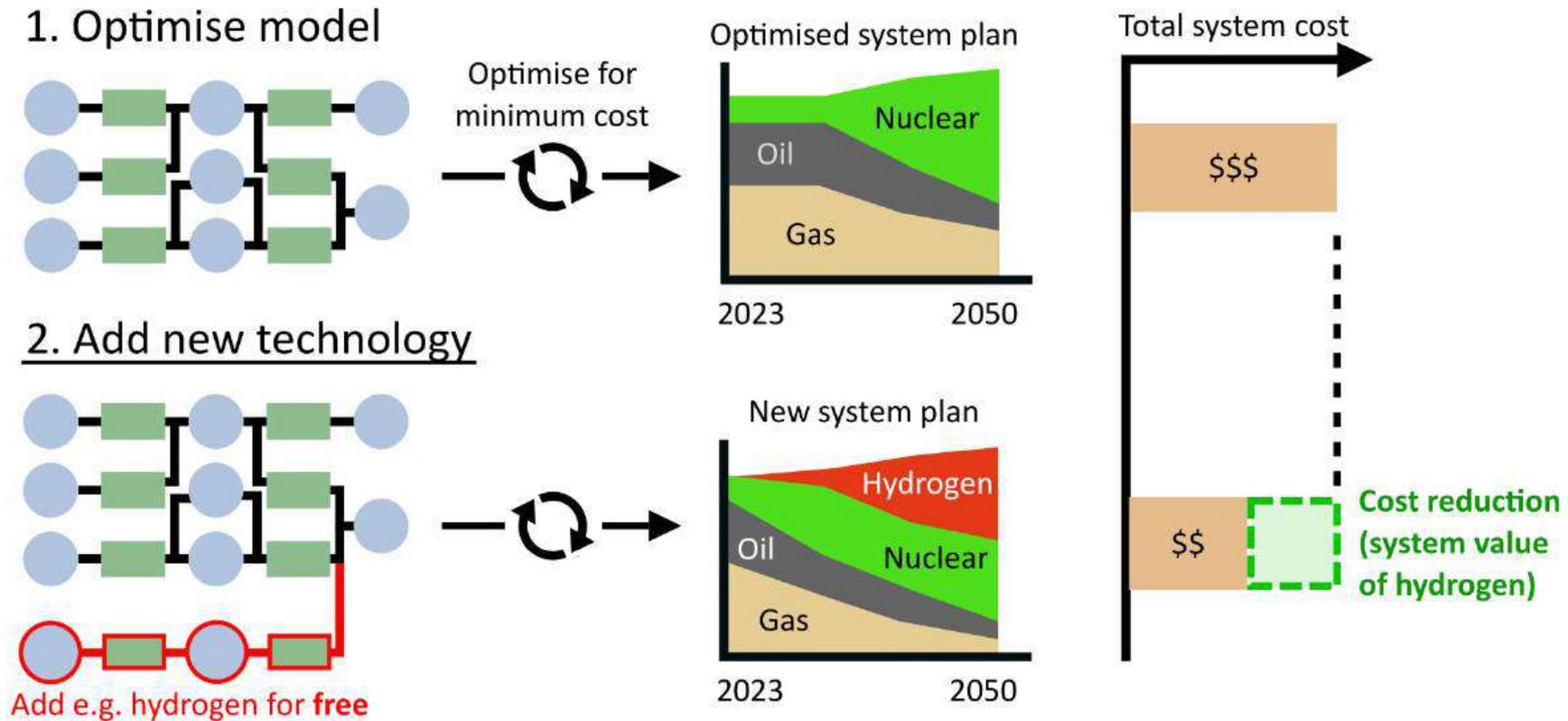


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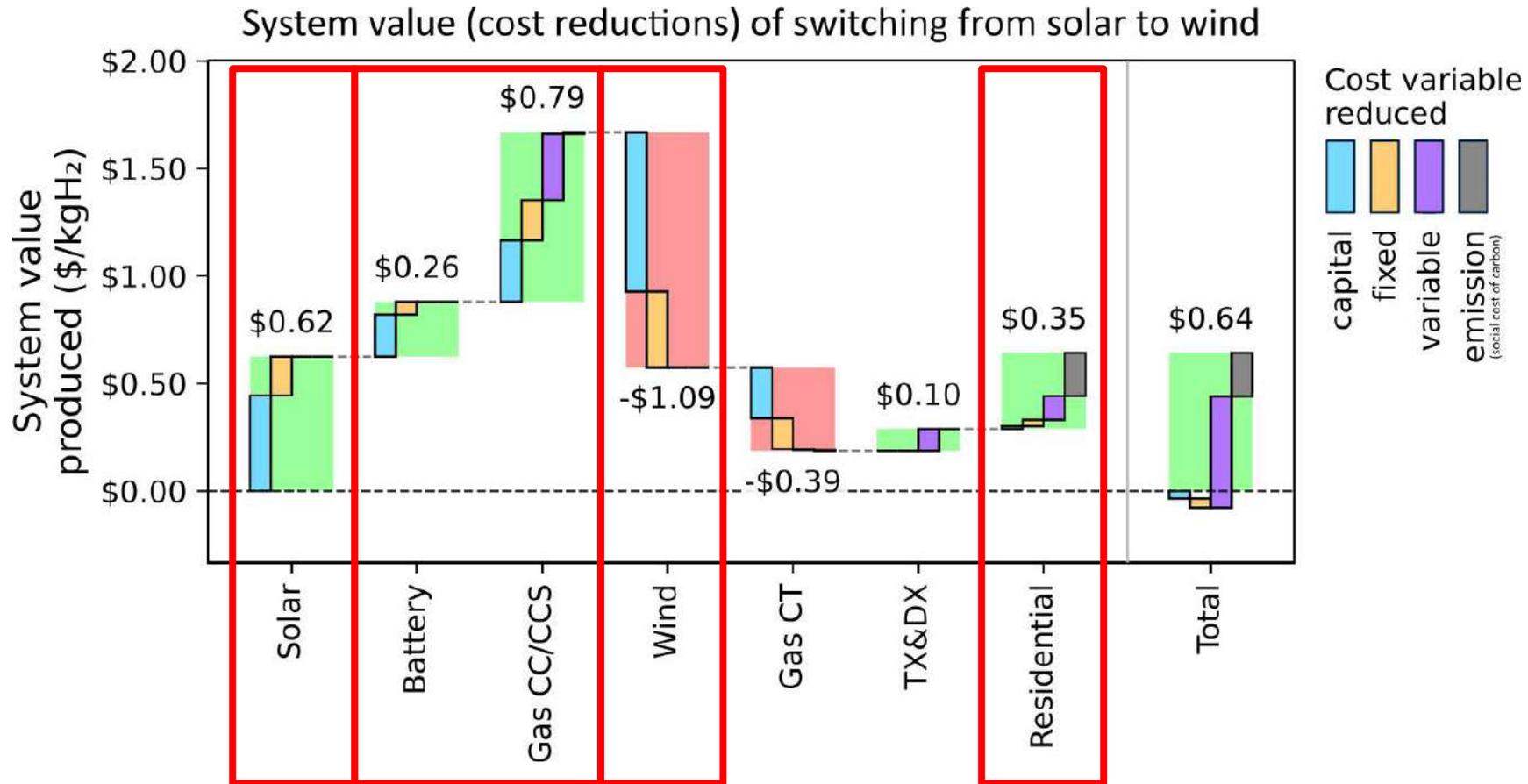
System Value Assessment (SVA)

Method



System Value Assessment (SVA)

Pairing H₂ production with wind outperforms solar – boosted by multi-sector interactions



The Canadian Open Energy Model (CANOE)

Summary

National-scale, multi-sectoral ESOM for Canada.

Open source and open access

Modular construction. Flexible scope and resolution.

Built for uncertainty analysis.

Model features:

Embodied carbon

Endogenous material flows

EV charging

Oil refineries

Seasonal storage

Endogenous technological learning

Survival curves

Growth rate constraints

More on CANOE



Our Temoa branch



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