

Workshop

PyPSA-Eur: A Sector-Coupled Open Optimisation Model of the European Energy System

Supported by:



on the basis of a decision
by the German Bundestag

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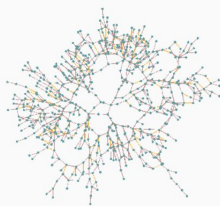
github.com/pypsa/pypsa-eur
pypsa-eur.readthedocs.io
resilient-project.github.io

What is PyPSA?

Our research focus:

- **Cost-effective pathways** to reduce greenhouse gas emissions
- **Evaluation** of grid expansion, hydrogen strategies, carbon management strategies
- **Co-optimisation** of generation, storage, conversion and transmission **infrastructure**
- **Algorithms** to improve the tractability of models
- **All open** source and open data

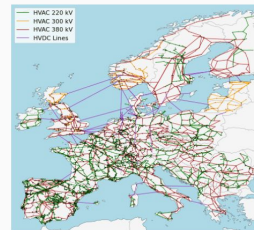
PyPSA



A python software toolbox for simulating and optimising modern power systems.

[Documentation »](#)

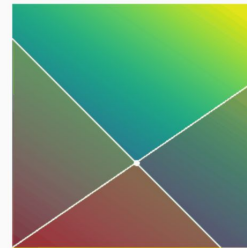
PyPSA-Eur



A Sector-Coupled Open Optimisation Model of the European Energy System

[Documentation »](#)

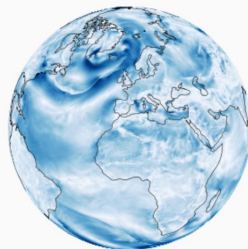
Linopy



Linear optimization interface for N-D labeled variables.

[Documentation »](#)

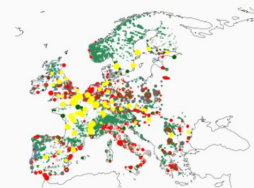
Atlite



A Lightweight Python Package for Calculating Renewable Power Potentials and Time Series

[Documentation »](#)

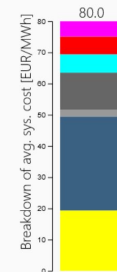
Powerplantmatching



A toolset for cleaning, standardizing and combining multiple power plant databases.

[Documentation »](#)

Model Energy



An online toolkit for calculating renewable electricity supplies.

Application examples

NGOs and international organisations



TSOs

Managing the Seasonal Variability of Electricity Demand and Supply

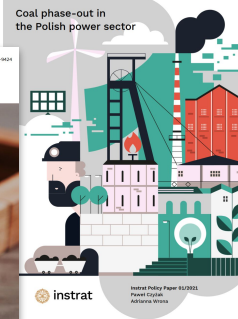


A path out of the gas crisis

New analysis shows Britain can cut gas from the power sector by the end of the decade, with huge cost savings from switching to renewables.
Publication date: 25th September 2022
Lead authors: Sarah Brown, Pavel Cypriak, Phil MacDonald
Other authors: Chelene Bruce-Lockhart, Ali Candell, Harriet Fox

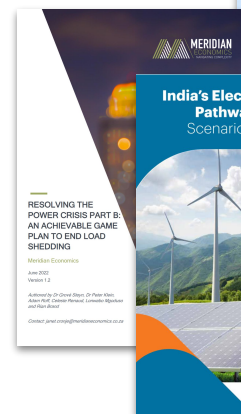
Achieving the goal

Coal phase-out in the Polish power sector



International

Regulators

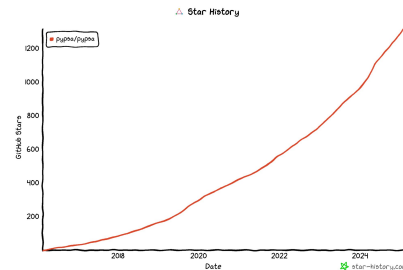


<https://pypsa.readthedocs.io/en/latest/references/users.html>



PyPSA:

Python for Power System Analysis



Capabilities

Capacity expansion (linear)

- single-horizon
- multi-horizon

Market modelling (linear)

- Linear optimal power flow
- Security-constrained LOPF
- Unit commitment
- Dispatch & redispatch

Non-linear power flow

- Newton-Raphson

With components for

- Electricity transmission networks and pipelines.
- Generators with **unit commitment constraints**
- **Variable** generation with time series (e.g. wind and solar)
- **Storage** with efficiency losses and inflow/spillage for hydro
- **Conversion** between energy carriers (PtX, CHP, BEV, DAC)

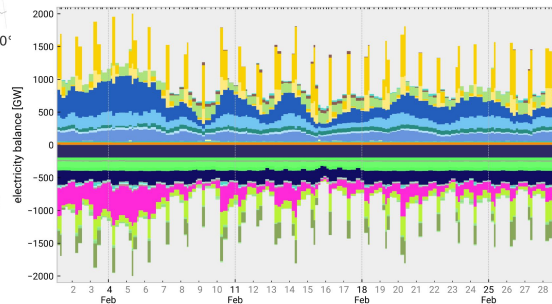
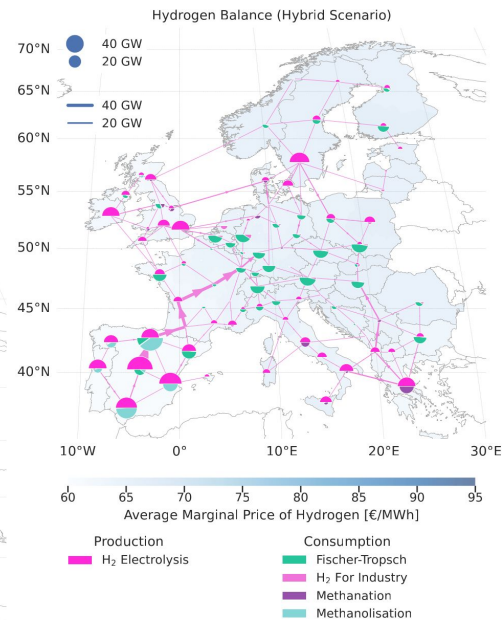
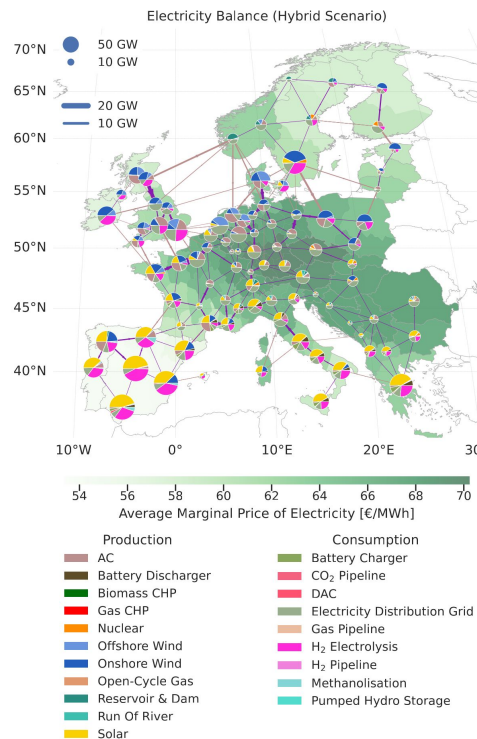
Backend

- all data stored in **pandas**
- framework built for performance with large networks and time series
- interfaces to major **solvers** (Gurobi, CPLEX, HiGHS, Xpress), with **linopy** (by PyPSA devs)
- highly **customisable**, but **no GUI**
- Suitable for greenfield, brownfield & pathway studies

PyPSA-Eur: A sector-coupled open model of the European energy system

Automated **workflow** to build energy system model of Europe from raw open data with high spatial and temporal resolution:

1. OSM transmission lines (>220 kV) + TYNDP
2. a database of existing power plants,
3. time series for electricity demand,
4. time series for wind/solar availability, and
5. geographic wind/solar potentials
6. cost and efficiency assumptions
7. methods for model simplification
8. more for sector-coupled networks like pipelines, LNG terminals, electric vehicles, industry locations,



Energy infrastructure planning in PyPSA as an optimisation problem

Find the long-term cost-optimal energy system, including investments and short-term costs:

$$\text{Min} \left[\begin{array}{c} \text{Yearly} \\ \text{system costs} \end{array} \right] = \text{Min} \left[\sum_n \left(\begin{array}{c} \text{Annualised} \\ \text{capital costs} \end{array} \right) + \sum_{n,t} \left(\begin{array}{c} \text{Marginal} \\ \text{costs} \end{array} \right) \right]$$

subject to

- meeting energy demand at each node n (e.g. region) and time t (e.g. hour of year)
- transmission constraints between nodes and linearised power flow
- wind, solar, hydro (variable renewables) availability time series $\forall n, t$
- installed capacity \leq geographical potentials for renewables
- fulfilling CO₂ emission reduction targets
- Flexibility from gas turbines, battery/hydrogen storage, HVDC links

data

data

data

data

More on that later!

Challenges with data-driven modelling

Create a full pipeline of data processing from raw data to results.

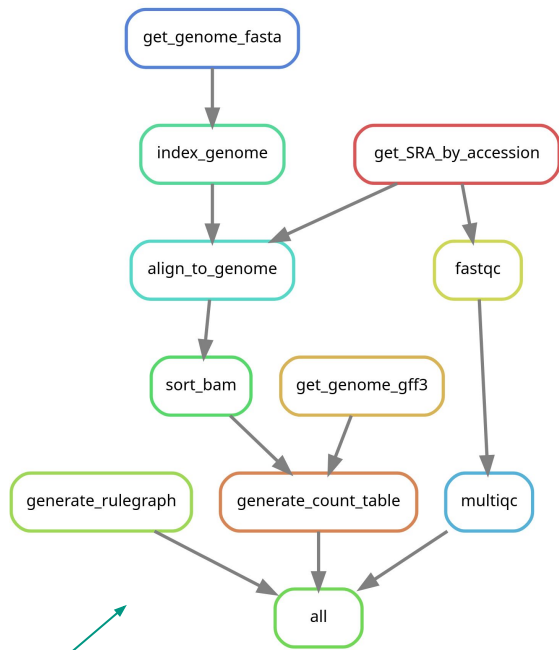
- Many different data **sources**
- Many data sources need **cleaning** and **processing**
- Many **intermediate** scripts and datasets
- Data and software **dependencies** need to be managed
- Data and code **change** over time
- Want to be able to **reproduce** results
- Want to run many different **scenarios**

Requires a scalable **workflow management tool!**



snakemake

Originally comes from bioinformatics field.



Miniature example of snakemake

Snakefile

```
rule mytask:  
    input:  
        "data/{sample}.txt"  
    output:  
        "result/{sample}.txt"  
    script:  
        "scripts/mytask.py"
```

```
rule myplot:  
    input:  
        "result/{sample}.txt"  
    output:  
        "figures/{sample}.pdf"  
    script:  
        "scripts/myplot.py"
```

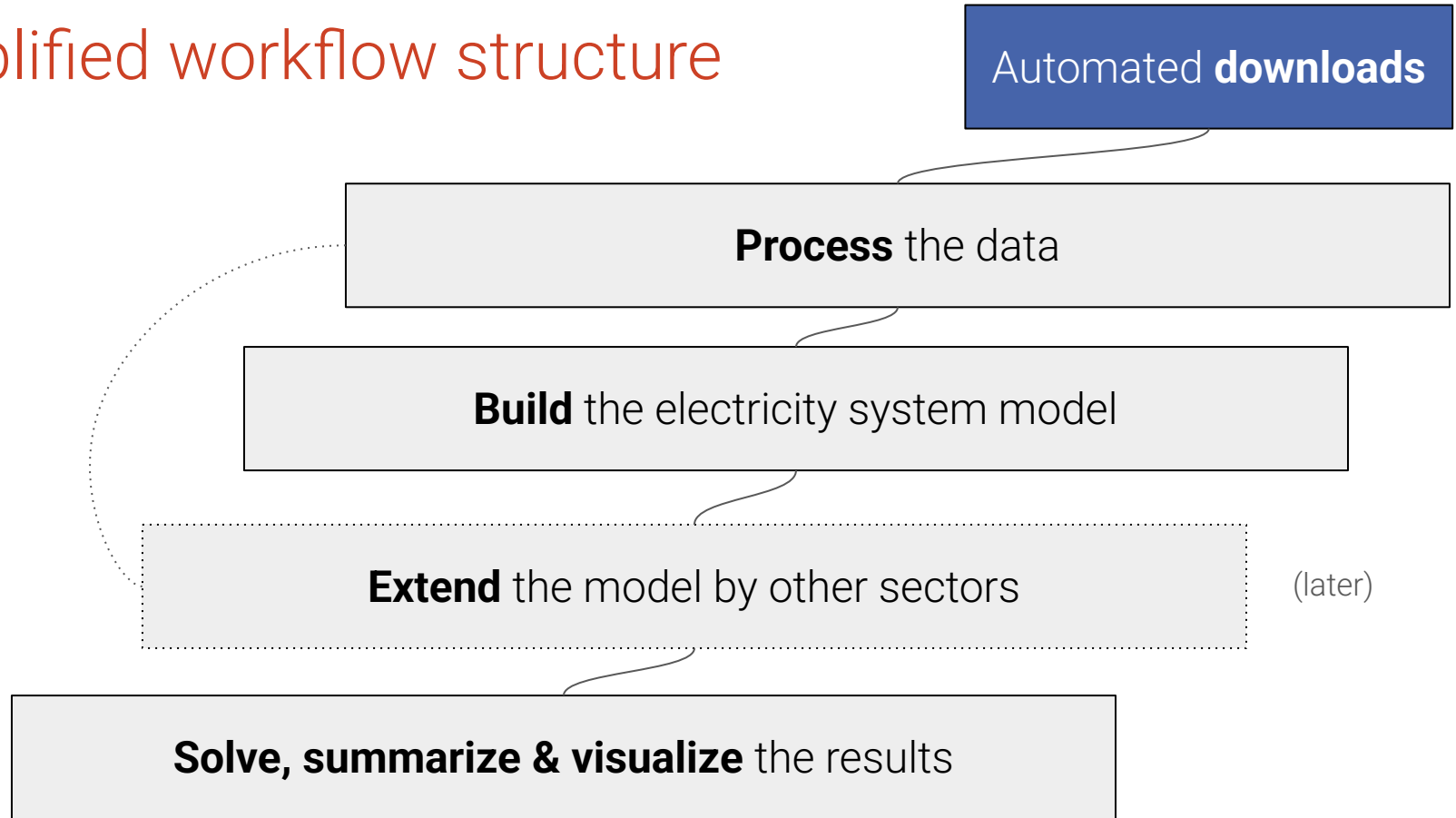


command: `$ snakemake figures/myfigure.pdf`



snakemake workflow for the electricity sector

Simplified workflow structure

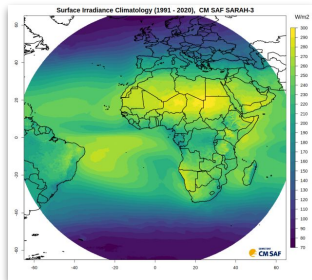


First, raw data is automatically downloaded.

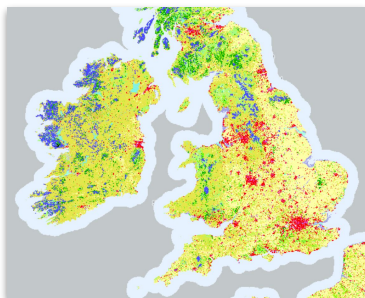
WDPA



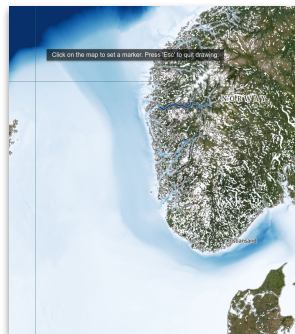
SARAH-3



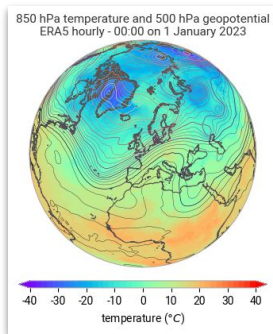
CORINE



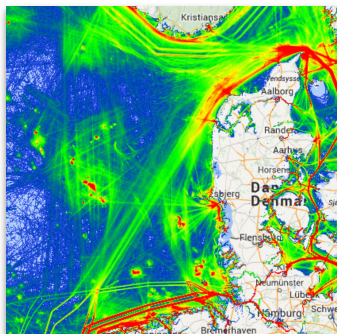
GEBCO



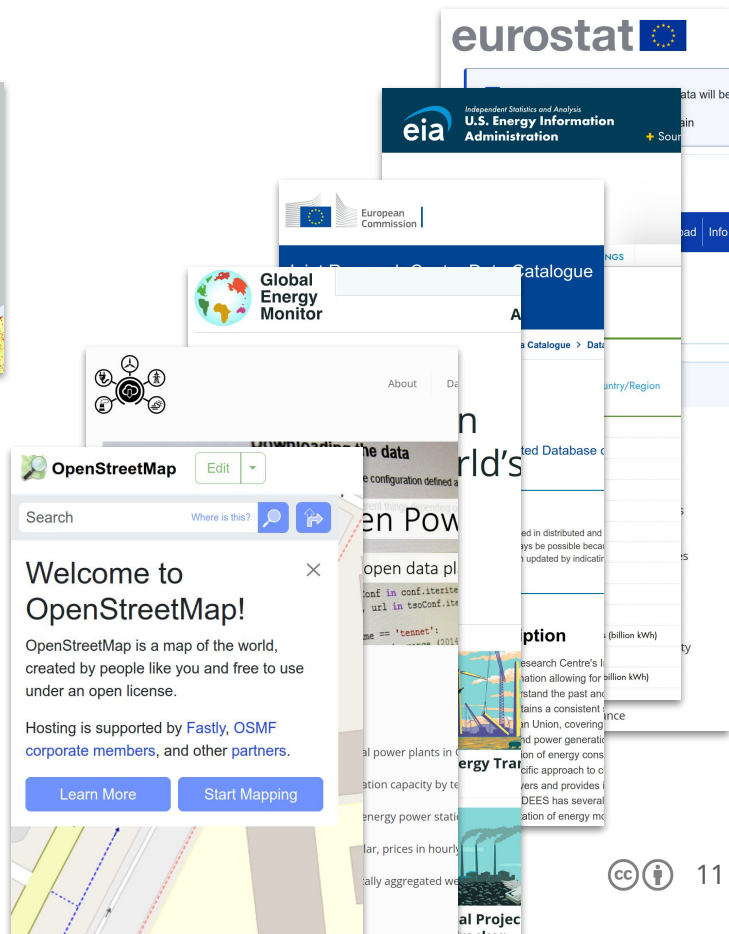
ERA5



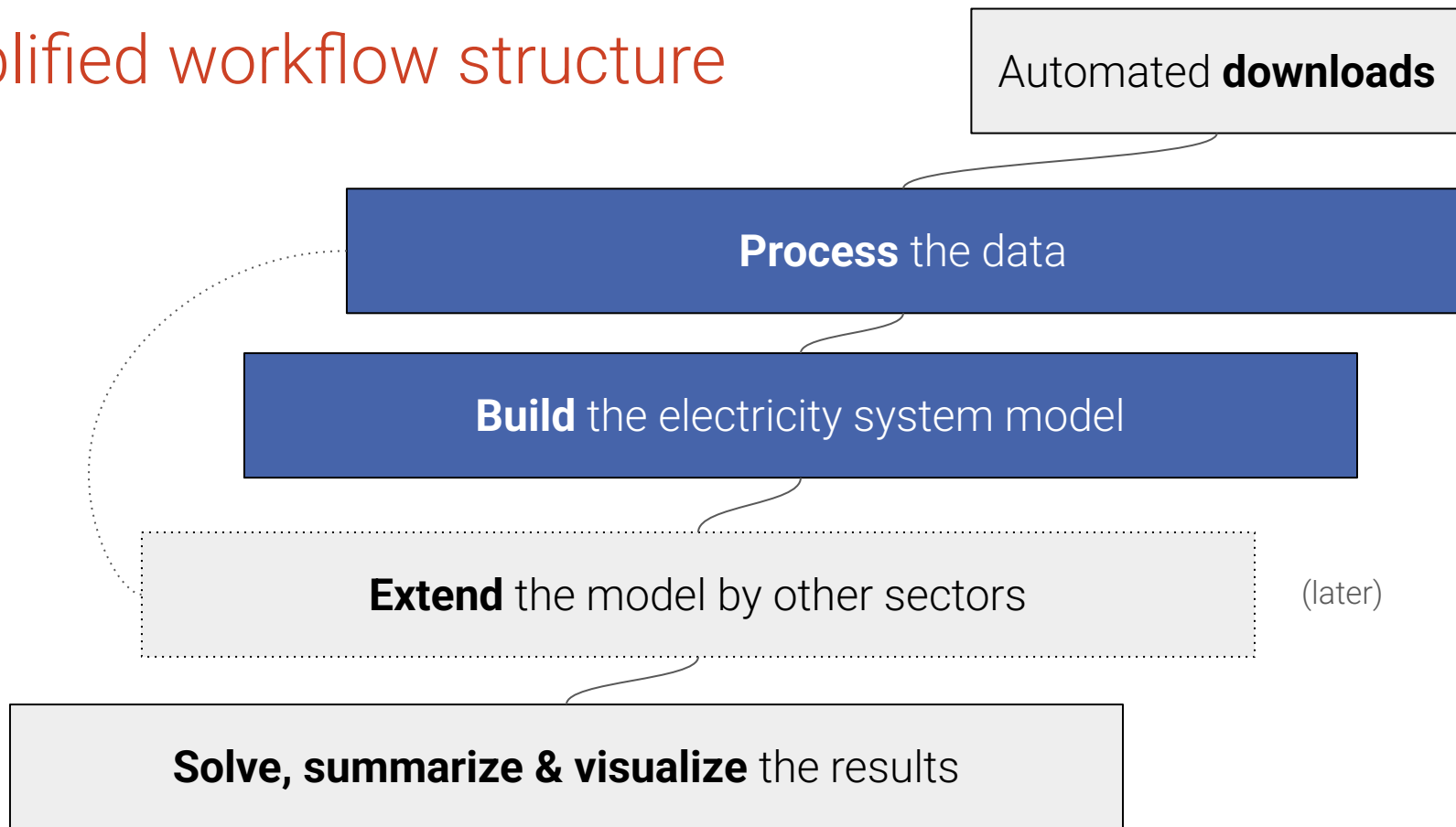
World Bank



https://pypsa-eur.readthedocs.io/en/latest/data_sources.html



Simplified workflow structure



Steps to building PyPSA-Eur electricity system

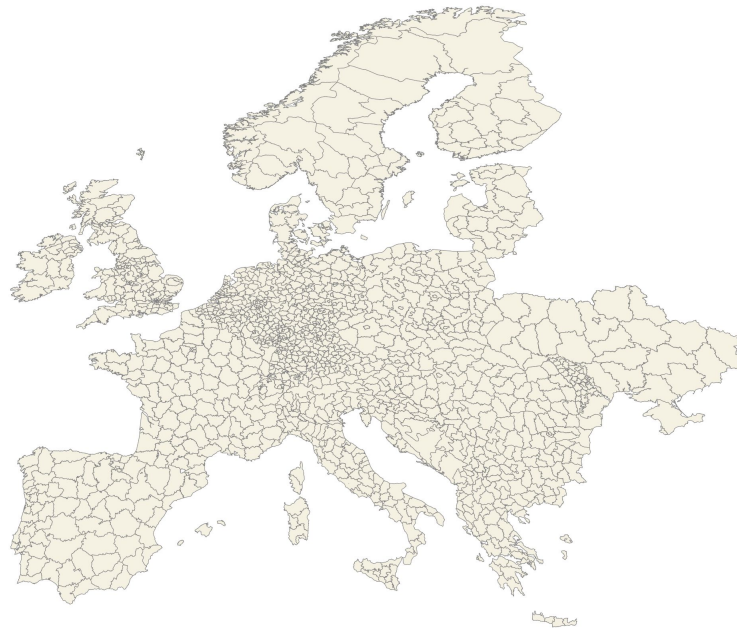
Retrieve onshore & offshore **polygons** for each country

build_shapes

Country shapes & exclusive economic zones (EEZ)



NUTS administrative regions (NUTS3)

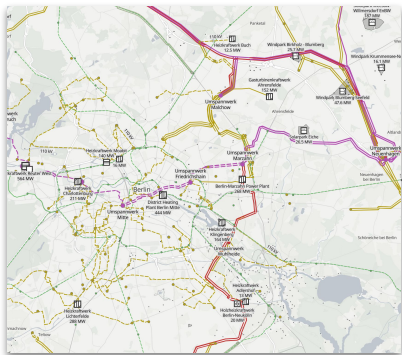


Steps to building PyPSA-Eur electricity system

Retrieve onshore & offshore polygons for each country	build_shapes
Construct a base high-voltage network with buses, transformers, AC & DC lines with DLR & TYNDP	base_network, build_transmission_projects

Power grid topology

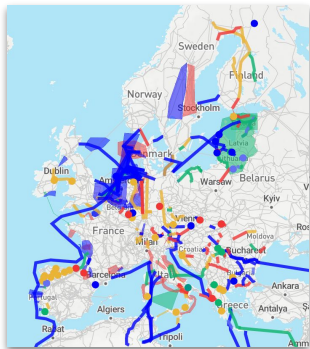
OpenStreetMap data



Apply **standard line types** for capacity and parameters.

Calculate **dynamic line rating** potential from weather data.

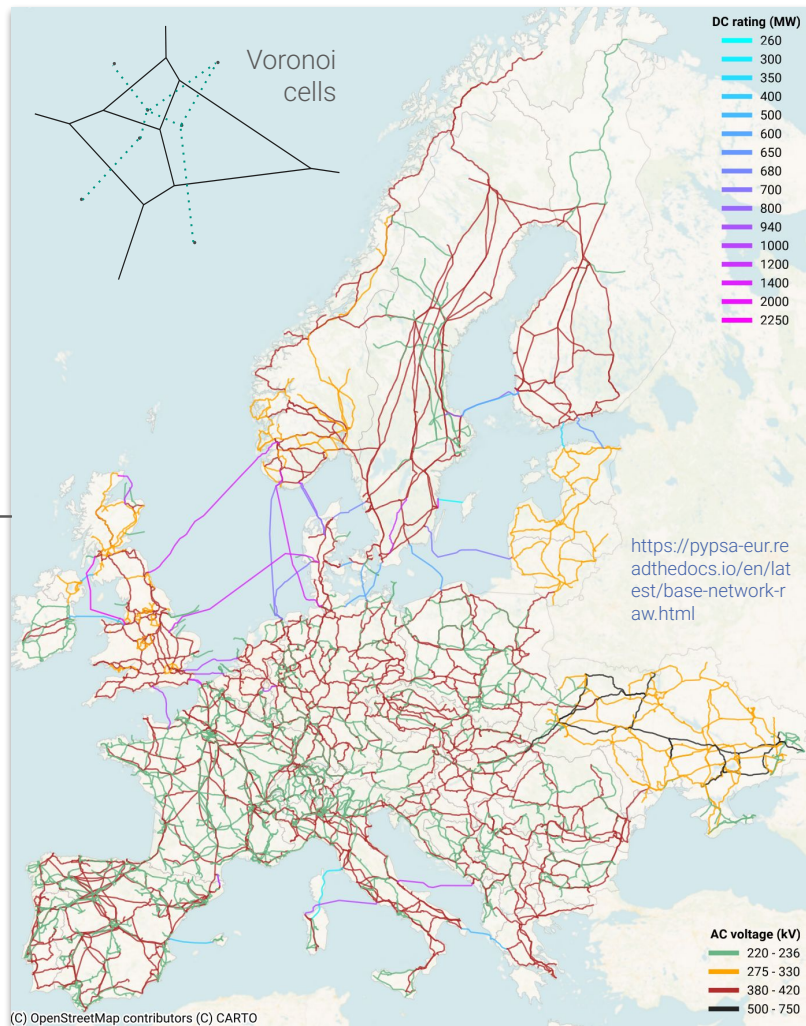
TYNDP projects



European network with

- ~5,800 buses
- ~7,300 AC lines (>220 kV)
- 36 HVDC links (+TYNDP)

<https://www.nature.com/articles/s41597-025-04550-7>



DC rating (MW)

- 260
- 300
- 350
- 400
- 500
- 600
- 650
- 680
- 700
- 800
- 940
- 1000
- 1200
- 1400
- 2000
- 2250

AC voltage (kV)

- 220 - 236
- 275 - 330
- 380 - 420
- 500 - 750

(C) OpenStreetMap contributors (C) CARTO

Steps to building PyPSA-Eur electricity system

Retrieve onshore & offshore polygons for each country	build_shapes
Construct a base high-voltage network with buses, transformers, AC & DC lines with DLR & TYNDP	base_network, build_transmission_projects
Transform all transmission lines to 380kV, remove dead ends & cluster with k-means or hierarchical clustering	simplify_network, cluster_network

Clustering the electricity network: `simplify_network`

Need to make the optimization problem less **computationally challenging**...

...if we want to **co-optimize** generation, storage, PtX conversion and transmission infrastructure:

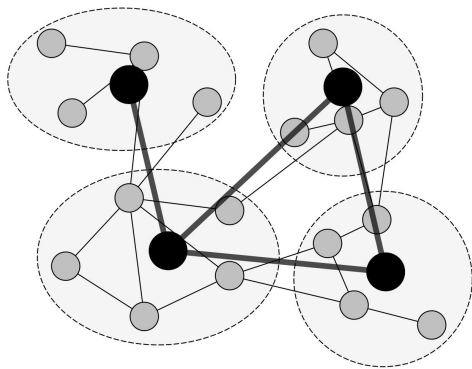
1. Lift all lines to **common voltage** level of 380 kV.
2. Remove **dead ends**.



Clustering the electricity network: `cluster_network`

Transformed
to **380 kV**

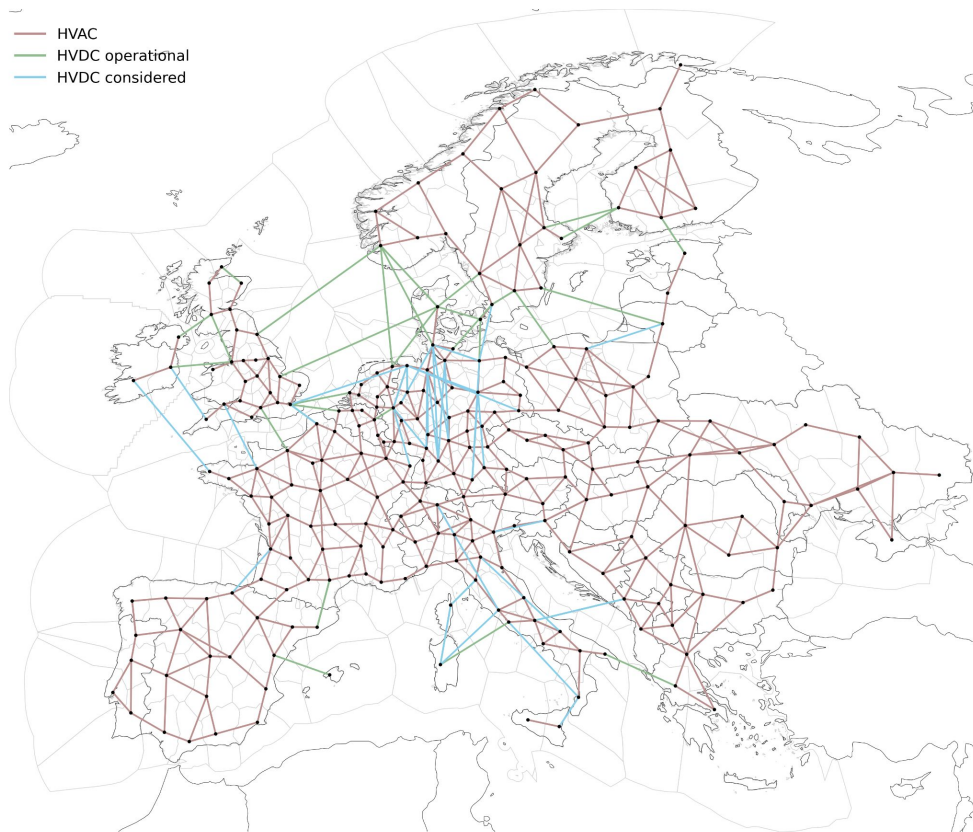
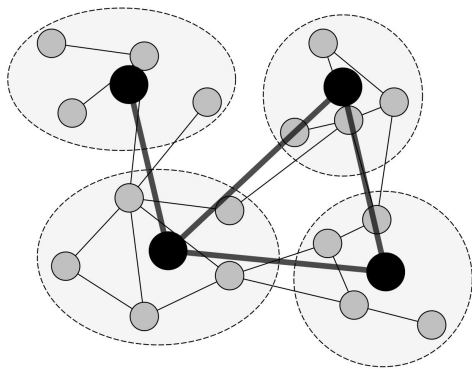
Clustered to
512 regions



Clustering the electricity network: `cluster_network`

Transformed
to **380 kV**

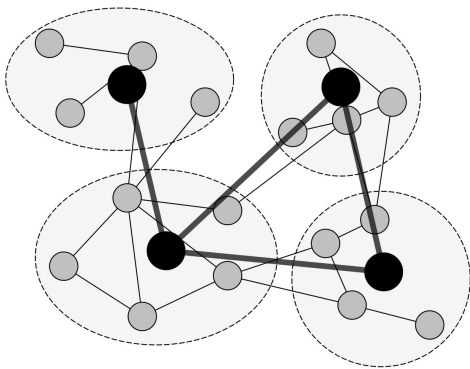
Clustered to
256 regions



Clustering the electricity network: `cluster_network`

Transformed
to **380 kV**

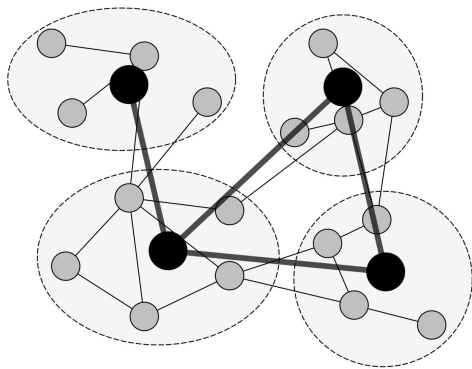
Clustered to
128 regions



Clustering the electricity network: `cluster_network`

Transformed
to **380 kV**

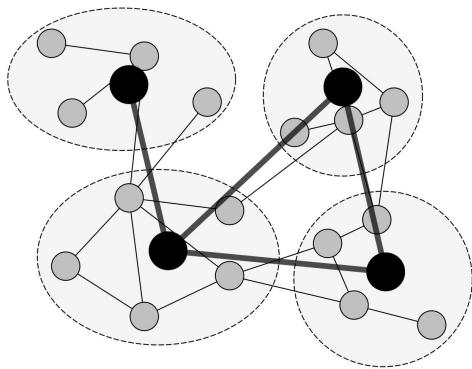
Clustered to
64 regions



Clustering the electricity network: `cluster_network`

Transformed
to **380 kV**

Clustered to
41 regions



Steps to building PyPSA-Eur electricity system

Retrieve onshore & offshore polygons for each country	build_shapes
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Transform all transmission lines to 380kV, remove dead ends & cluster with k-means or hierarchical clustering	simplify_network, cluster_network
Determine eligible areas for utility-scale PV & onshore/offshore wind park development	determine_availability_matrix
Build renewable capacity factor profiles for each clustered region based on land availability	build_renewable_profiles, build_hydro_profile

atlite: Convert weather data to energy systems data

Rule: build_renewable profiles

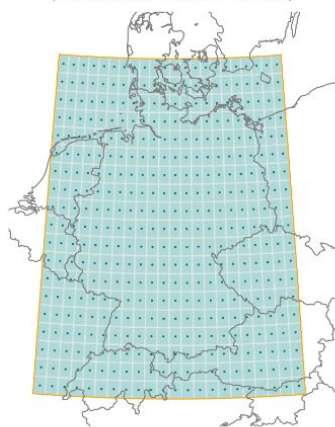
pypi v0.3.0 conda-forge v0.3.0 Tests passing codecov 72% docs passing license MIT
REUSE compliant JOSS 10.21105/joss.03294 chat 52 online stackoverflow pypsa questions 44

Python library for converting **weather data** (e.g. wind, solar radiation, temperature, precipitation) into **energy systems data**:

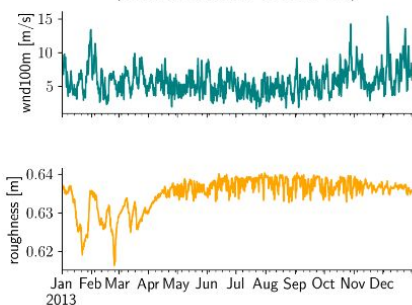
- solar photovoltaics
- solar thermal collectors
- wind turbines
- hydro run-off, reservoir, dams
- heat pump COPs
- dynamic line rating (DLR)
- heating and cooling demand (HDD/CDD)

It can also perform **land eligibility analyses**.

1. Create Cutout (Select spatio-temporal bounds)

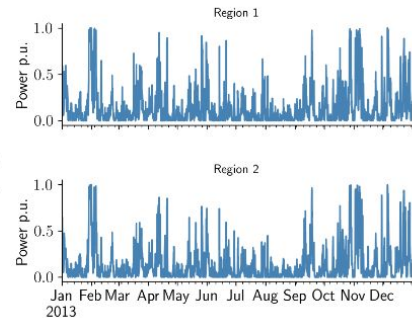
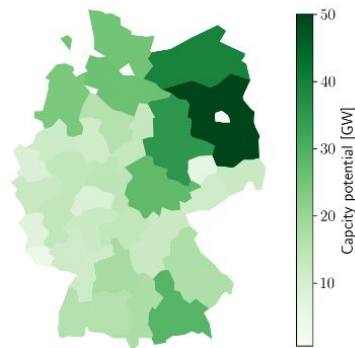


2. Prepare Cutout (Retrieve data per weather cell)



⋮

3. Convert Cutout (Calculate potentials and timeseries per region)



⋮

Time series for renewables

Historical meteorological weather data from ERA5 and SARA3 (up to 84 years, 30x30 km)

attr: influx_direct - time: 2013-03-01T00:00:00



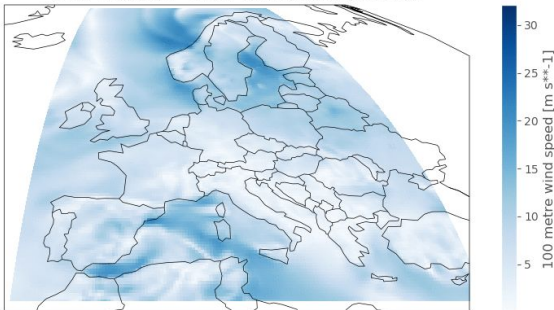
Solar panel models

- orientation
- material

atlite: Convert weather data to energy systems data

pypl v0.3.0 conda-forge v0.3.0 Tests passing codecov 72% docs passing license MIT
REUSE compliant JOSS 10.21105/joss.03294 chat 52 online stackoverflow pypsa questions 44

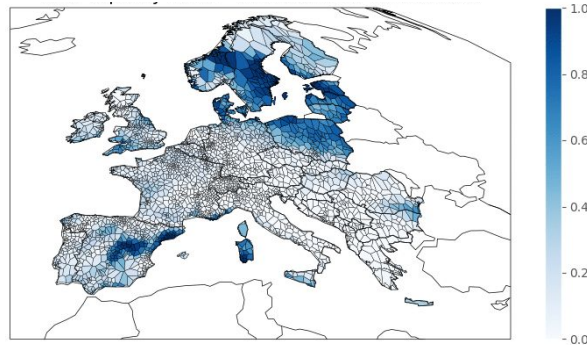
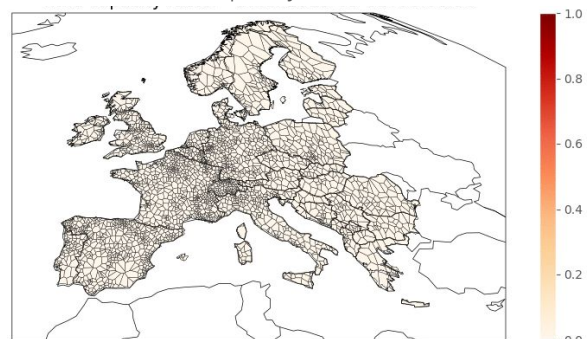
attr: wnd100m - time: 2013-03-01T00:00:00



Wind turbine models

- power curve
- surface roughness

Wind and solar capacity factors



Land availability for renewables



Example:
Onshore wind
in one clustered
region



atlite: Convert weather data to energy systems data

pypl v0.3.0 conda-forge v0.3.0 Tests passing codecov 72% docs passing license MIT
REUSE compliant JOSS 10.21105/joss.03294 chat 52 online stackoverflow pypsa questions 44

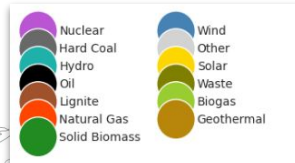
- CORINE / LUISA land cover
 - eligible land types
 - distance requirements
- NATURA / WDPA natural protection areas
- GEBCO bathymetry data
- Shipping lanes
- Distance to shore

Steps to building PyPSA-Eur electricity system

Retrieve onshore & offshore polygons for each country	<code>build_shapes</code>
Construct a base high-voltage network with buses, transformers, AC & DC lines with DLR & TYNDP	<code>base_network,</code> <code>build_transmission_projects</code>
Transform all transmission lines to 380kV, remove dead ends & cluster with k-means or hierarchical clustering	<code>simplify_network,</code> <code>cluster_network</code>
Determine eligible areas for utility-scale PV & onshore/offshore wind park development	<code>determine_availability_matrix</code>
Build renewable capacity factor profiles for each clustered region based on land availability	<code>build_renewable_profiles,</code> <code>build_hydro_profile</code>
Prepare existing renewables and fossil power plants	<code>build_powerplants</code>

Welcome to powerplantmatching's documentation!

<https://globalenergymonitor.org/projects/global-integrated-power-tracker/tracker-map/>



pypi v0.7.0 conda-forge v0.7.0 python >=3.9 Tests failing docs passing pre-commit.ci passed Ruff
license GPLv3+ DOI 10.5281/zenodo.3358985 stackoverflow pypsa questions 44

A toolset for cleaning, standardizing and combining multiple power plant databases.

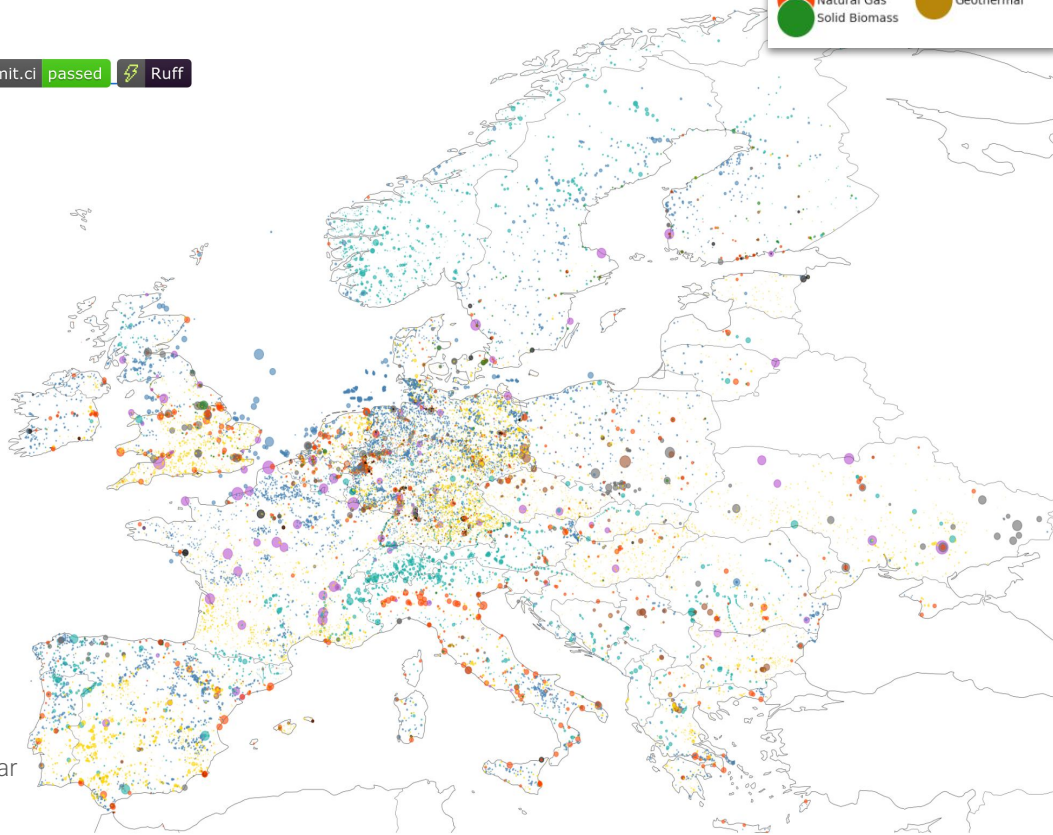
```
import powerplantmatching as pm
df = pm.powerplants(from_url=True)
df.query("DateIn > 2000")
```

Sources

- Global Energy Monitor (GEM)
- Open Power System Data (OPSD)
- Global Energy Observatory
- World Resources Institute
- Marktstammdatenregister (MaStR)
- CARMA
- ENTSO-E, BNetzA, UBA, IRENA
- JRC for hydro power plants

Attributes

- name
- fuel type
- technology
- country
- capacity
- commissioning year
- retirement year
- coordinates



github.com/pypsa/powerplantmatching



Steps to building PyPSA-Eur electricity system

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Build renewable capacity factor profiles for each clustered region based on land availability	<code>build_renewable_profiles,</code> <code>build_hydro_profile</code>
Prepare existing renewables and fossil power plants	<code>build_powerplants</code>
Add generation, storage and demand to the network with techno-economic assumptions on costs and efficiencies, ...	<code>add_electricity,</code> <code>prepare_network</code>

Open database of techno-economic assumptions

- compiles **techno-economic assumptions** on energy system components
 - investment costs, FOM/VOM costs, efficiencies, lifetimes
 - for given years, e.g. 2020, 2030, 2040, 2050
 - from mixed sources, but prioritising **Danish Energy Agency** where available (and sensible)

Preview Code Blame 1097 lines (1097 loc) · 213 KB						Raw Copy Download Edit	
Q fischer-tropsch							
1	technology	parameter	value	unit	source		
217	Fischer-Tropsch	FOM	3.0	%/year	Agora Energiewende (2018): The Future Cost of Electricity-Based Synthetic Fuels (https://www.agora-energiewende.de/en/publications/the-future-cost-of-electricity-based-synthetic-fuels-1/), section 6.3.2.1.		
218	Fischer-Tropsch	VOM	4.4663	EUR/MWh_FT	Danish Energy Agency, data_sheets_for_renewable_fuels.xlsx		
219	Fischer-Tropsch	capture rate	0.9	per unit	Assumption based on doi:10.1016/j.biombioe.2015.01.006		
220	Fischer-Tropsch	carbondioxide-input	0.326	t_CO2/MWh_FT	DEA (2022): Technology Data for Renewable Fuels (https://ens.dk/en/our-services/projections-and-models/technology-data/technology-data-renewable-fuels), Hydrogen to Jet Fuel, Table 10 / pg. 267.		
221	Fischer-Tropsch	efficiency	0.799	per unit	Agora Energiewende (2018): The Future Cost of Electricity-Based Synthetic Fuels (https://www.agora-energiewende.de/en/publications/the-future-cost-of-electricity-based-synthetic-fuels-1/), section 6.3.2.2.		
222	Fischer-Tropsch	electricity-input	0.007	MWh_el/MWh_FT	DEA (2022): Technology Data for Renewable Fuels (https://ens.dk/en/our-services/projections-and-models/technology-data/technology-data-renewable-fuels), Hydrogen to Jet Fuel, Table 10 / pg. 267.		
223	Fischer-Tropsch	hydrogen-input	1.421	MWh_H2/MWh_FT	DEA (2022): Technology Data for Renewable Fuels (https://ens.dk/en/our-services/projections-and-models/technology-data/technology-data-renewable-fuels), Hydrogen to Jet Fuel, Table 10 / pg. 267.		
224	Fischer-Tropsch	investment	703726.4462	EUR/MW_FT	Agora Energiewende (2018): The Future Cost of Electricity-Based Synthetic Fuels (https://www.agora-energiewende.de/en/publications/the-future-cost-of-electricity-based-synthetic-fuels-1/), table 8: "Reference scenario".		
225	Fischer-Tropsch	lifetime	20.0	years	Danish Energy Agency, Technology Data for Renewable Fuels (04/2022), Data sheet "Methanol to Power".		
956	methanation	lifetime	20.0	years	Guesstimate.		

https://github.com/PyPSA/technology-data/blob/master/outputs/costs_2030.csv

Temporal aggregation

Multiple options:

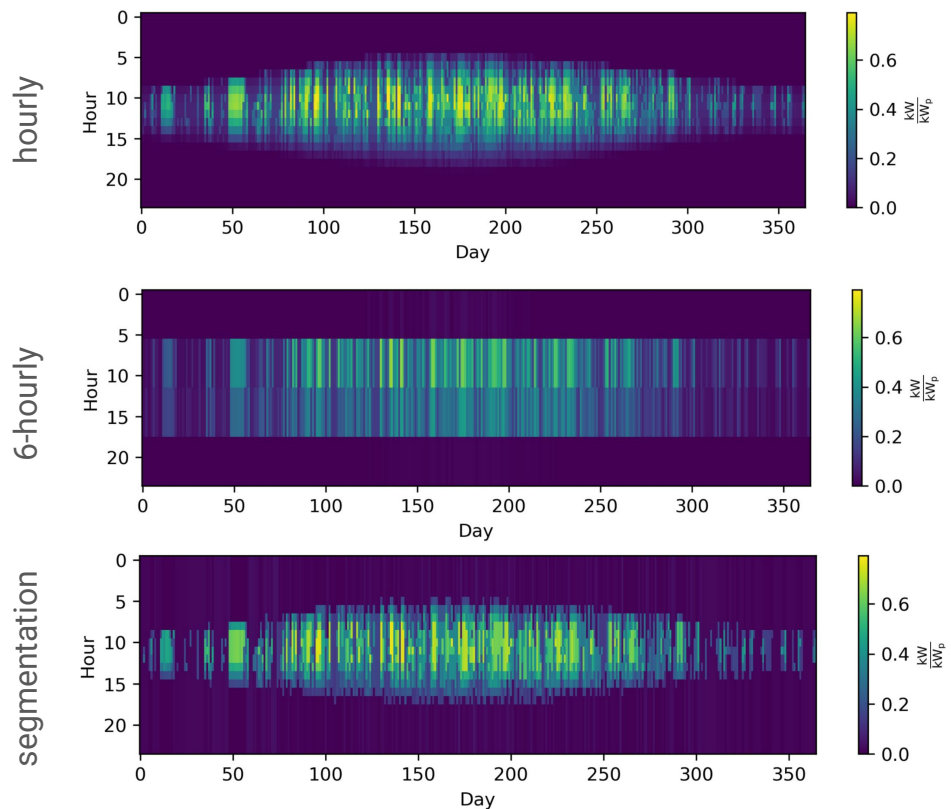
1. **averaging** of every Nth hour
2. **sampling** every Nth hour (e.g. 3-hourly)
3. Non-equidistant **segmentation** with pre-defined number of segments using the **tsam** Python library from **FZ Jülich**

Introduction

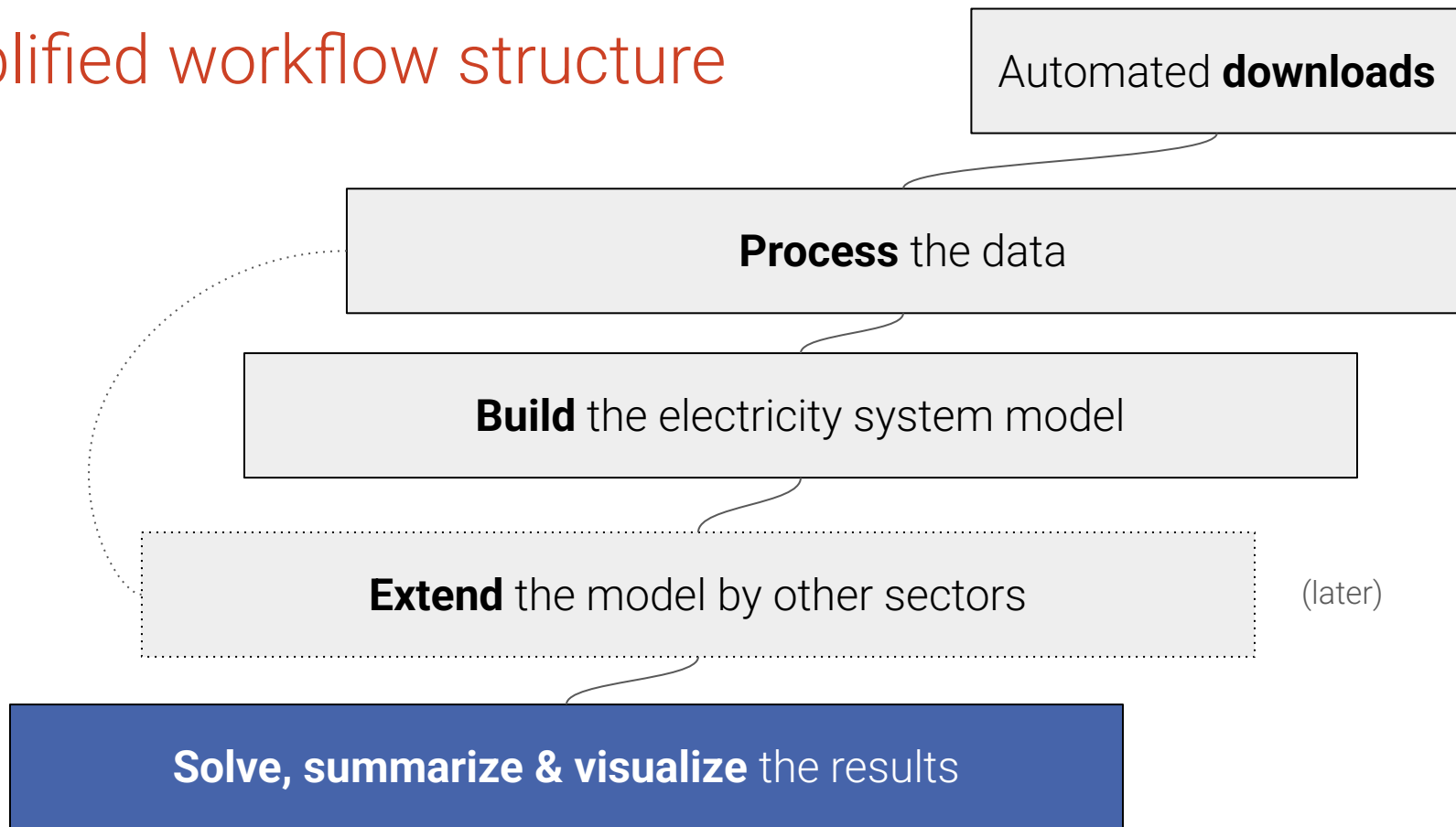


tsam - time series aggregation module

<https://tsam.readthedocs.io/en/latest/newsDoc.html>



Simplified workflow structure



linopy: Linear optimization with N-D labeled variables

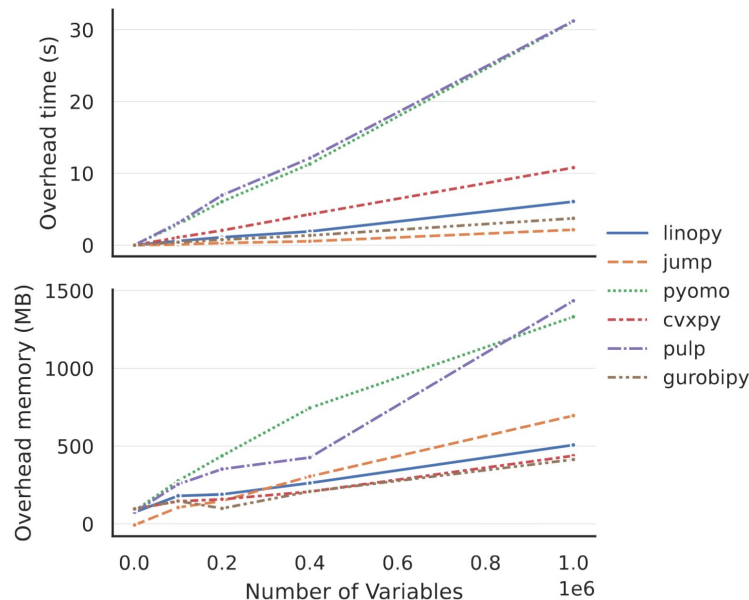
pypi v0.5.0  license MIT

Python library that facilitates **optimization** with real-world, large-scale data.

It supports:

- Linear (LP),
- Mixed-Integer (MILP),
- Quadratic programming (QP).

It has been developed to make linear programming in Python easy, highly-flexible and – most importantly – **highly performant**.



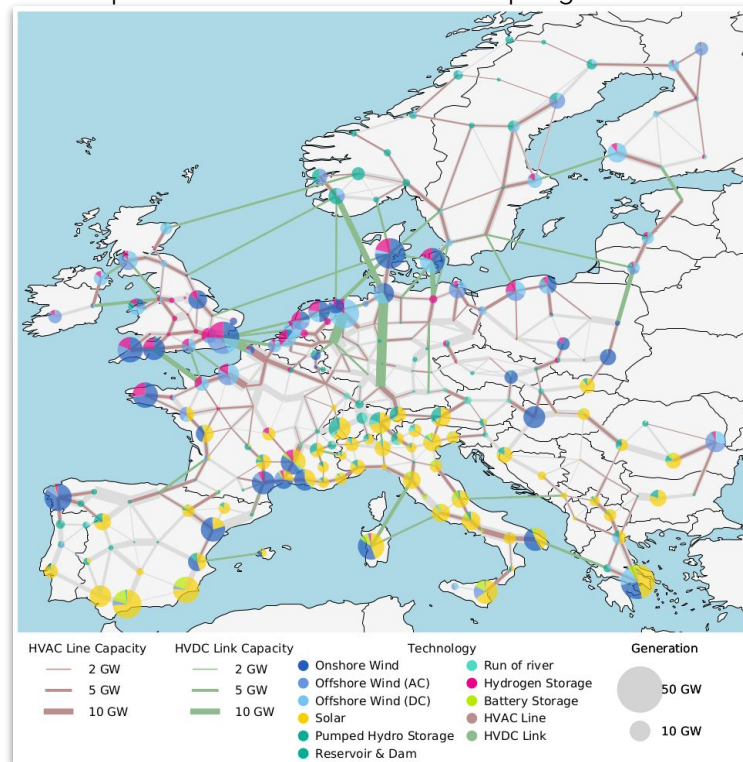
Solving and summarising networks

Hardware requirements:

- Building the model **can run locally** on most modern laptops. Very simple models can run with HiGHS solver.
- But access to a **commercial solver** and a larger **cluster/workstation** is required for solving problems (~250 GB RAM per scenario if resolution is very high)!

There is a **statistics module** in PyPSA designed to help with analysing solved networks and several **figures/maps** are created automatically.

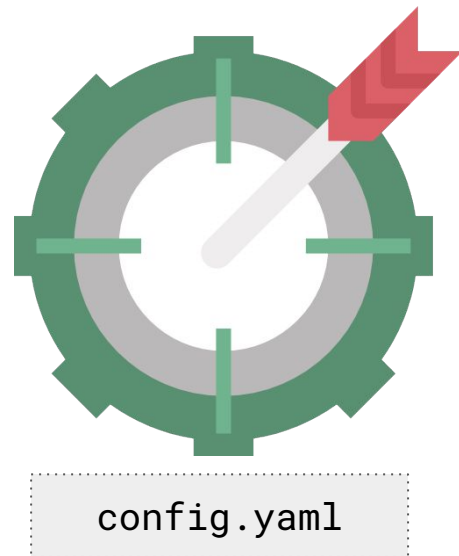
Example result without sector-coupling



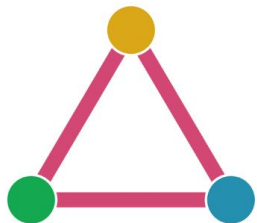
What is configurable?

electricity-only
examples

- Select subset of countries and focus countries (e.g. only DE)
- Select weather year (1940 - 2024 for ERA5)
- Specify CO₂ constraint and gas usage limit
- Tweak spatial resolution (between 41 and >1000 nodes)
- Tweak temporal resolution (from hourly to N-hourly)
- Customize cost assumptions (e.g. 2020, 2030, 2050)
- Parametrize technologies (e.g. wind turbine type, panel orientation)
- Define land use eligibility criteria (e.g. distance requirements)
- Pick a solver (HiGHS, Gurobi, CPLEX, Xpress...)
- Choose between greenfield or brownfield expansion



Let's look at this in more detail!



Q Search ctrl + K

Getting Started

- Introduction
- Installation
- Tutorial: Electricity-Only
- Tutorial: Sector-Coupled

Configuration

- Wildcards
- Configuration**
- Foresight Options
- Techno-Economic Assumptions

Rules Overview

- Retrieving Data
- Building Electricity Networks
- Building Sector-Coupled Networks
- Solving Networks
- Plotting and Summaries



Configuration

PyPSA-Eur has several configuration options which are documented in this section and are collected in a `config/config.yaml` file. This file defines deviations from the default configuration (`config/config.default.yaml`); confer installation instructions at [Handling Configuration Files](#).

Top-level configuration

“Private” refers to local, machine-specific settings or data meant for personal use, not to be shared. “Remote” indicates the address of a server used for data exchange, often for clusters and data pushing/pulling.

```
version: v2025.01.0
tutorial: false

logging:
  level: INFO
  format: '%(levelname)s:%(name)s:%(message)s'

private:
  keys:
    entsoe_api:

remote:
  ssh: ""
  path: ""
```

	Unit	Values	Description
version	–	0.x.x	Version of PyPSA-Eur. Descriptive only.
tutorial	bool	{true, false}	Switch to retrieve the tutorial data set instead of the full data set.
logging			



Contents

Top-level configuration

run

foresight

scenario

countries

snapshots

enable

co2 budget

electricity

atlite

renewable

conventional

lines

links

transmission projects

transformers

load

energy

biomass

solar_thermal

existing_capacities

sector

industry

costs

clustering

<https://pypsa-eur.readthedocs.io/en/latest/configuration.html>

Live Demo – Belgium / electricity-only / few days

Start with a dry-run:

Don't forget to activate your conda environment first!

```
$ snakemake -j1 solve_elec_networks --configfile config/test/config.electricity.yaml -n
```

Then execute the same command “for real” by dropping “-n” flag:

The “-j1” flag tells snakemake to run one job at a time.

```
$ snakemake -j1 solve_elec_networks --configfile config/test/config.electricity.yaml
```

To explore results, start a Jupyter notebook:

```
$ jupyter notebook
```

Practical Phase

(electricity-only)

2) Install conda environment

Installation links:

- [Anaconda](#) (bigger download):
- [Miniconda](#) (recommended):

```
$ conda update conda
$ conda env create -f envs/environment.yaml
$ conda activate pypsa-eur
```

4) Explore PyPSA network in a Jupyter notebook

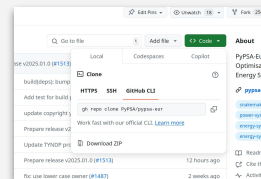
```
import pypsa
fn = "results/test-elec/networks/base_s_5_elec.nc"
n = pypsa.Network(fn)
n.statistics()
n.plot()
```

1) Download the repository

Open a terminal / CMD and type:

```
$ cd ~/path/to/my/directory
$ git clone https://github.com/PyPSA/pypsa-eur.git
$ cd pypsa-eur
```

You can also download
the repository as a ZIP by
hand.



3) Run PyPSA-Eur tutorial with **snakemake**

Guide:

<https://pypsa-eur.readthedocs.io/en/latest/tutorial.html>

```
$ snakemake solve_elec_networks
--configfile
config/test/config.electricity.yaml -j1
```

Users of Windows, add two lines to YAML:

```
run:
  use_shadow_directory: false
```

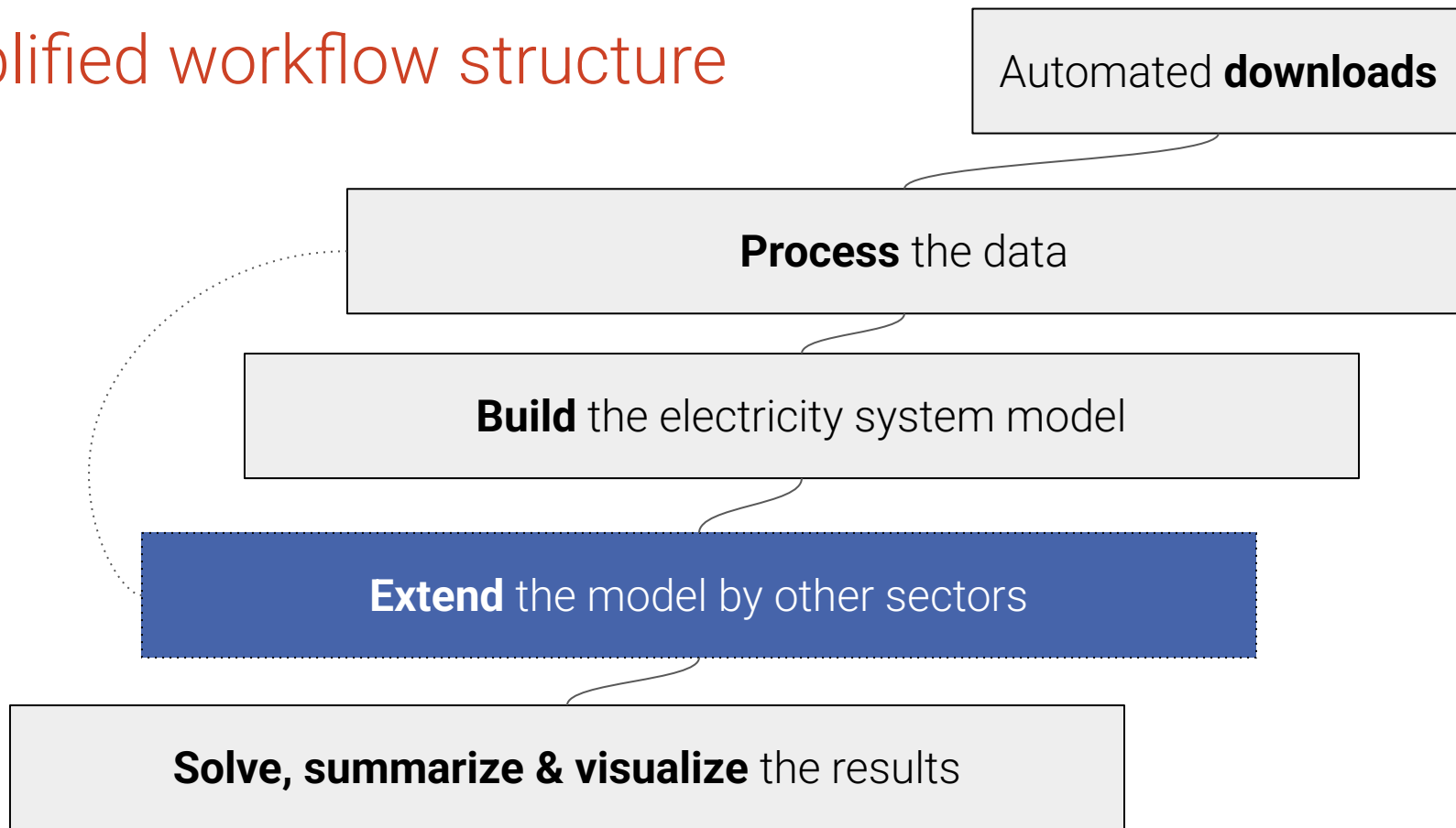
Small exploratory configuration tasks

(electricity-only)

Go to <https://pypsa-eur.readthedocs.io/en/latest/configuration.html> and <https://github.com/PyPSA/pypsa-eur/blob/master/config/config.default.yaml> and try to find out how to configure some of the settings for **electricity-only models** listed below:

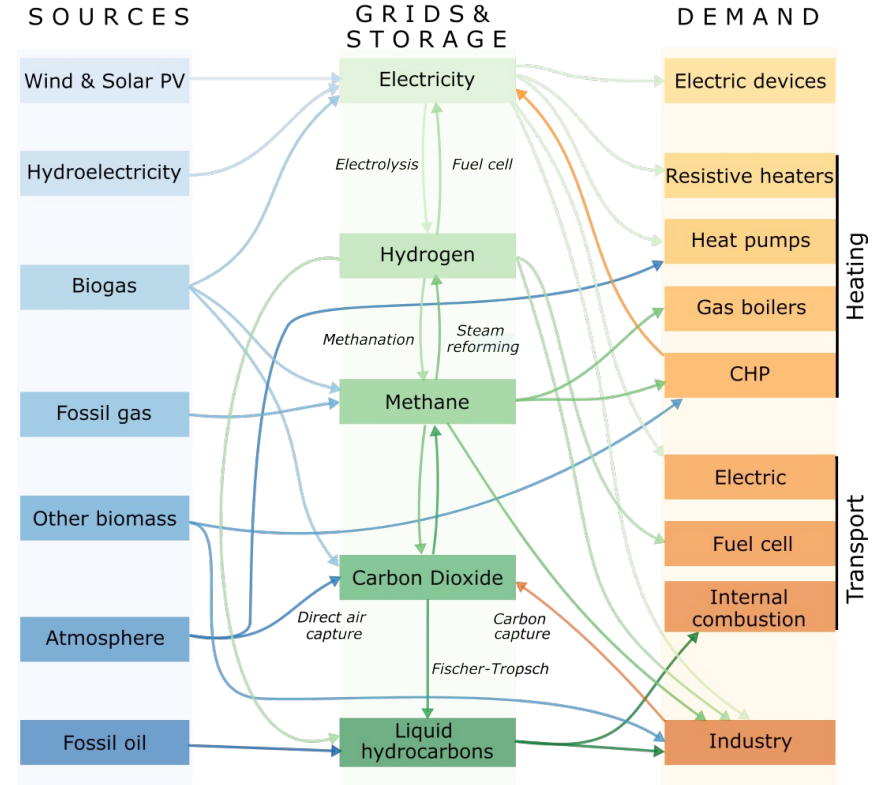
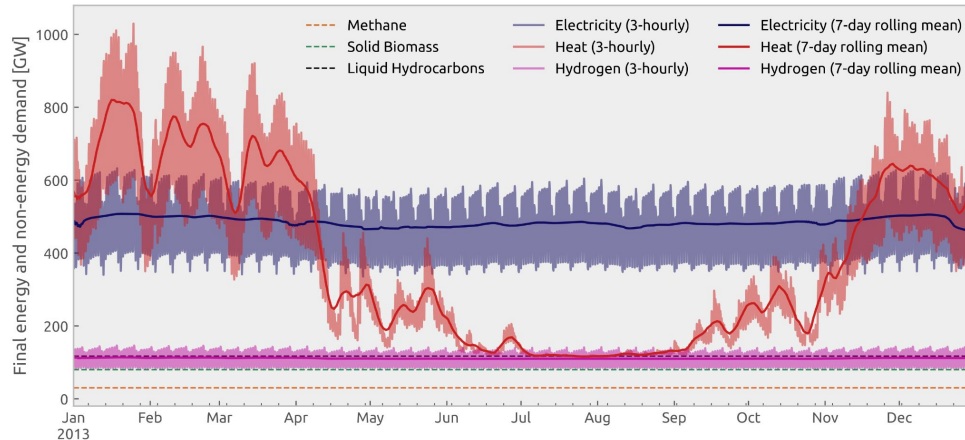
1. Increase the maximum line loading from 70% to 100%.
2. Disable power transmission grid reinforcements.
3. Activate dynamic line rating with default settings.
4. Activate linearised transmission loss approximation.
5. Deactivate the estimation of existing renewable capacities.
6. Change the techno-economic assumptions to the year 2020.
7. Remove the option to build hydrogen or battery storage.

Simplified workflow structure

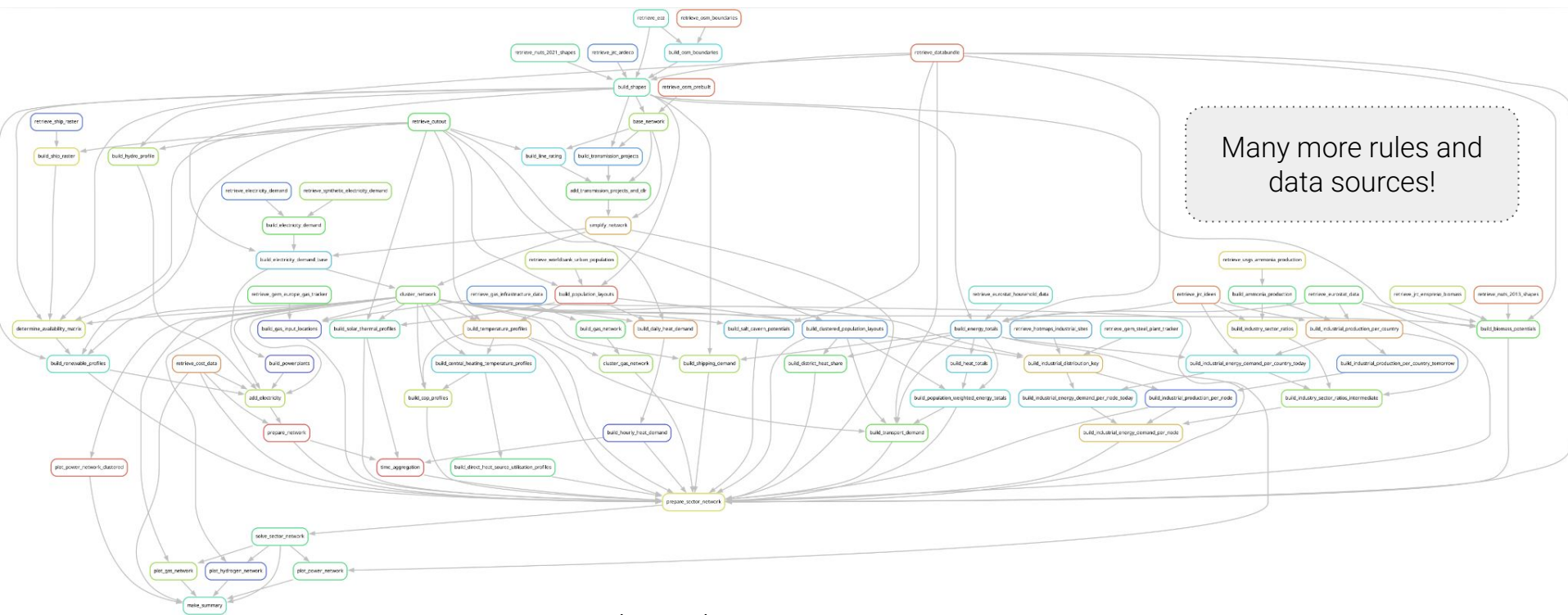


Coupling with other sectors

- transport sector (EVs, shipping, aviation)
- heating sector (district heating, individual)
- industry sector (steel, chemicals, ammonia, ...)
- carbon management (CCUTS) + biomass
- hydrogen, CO₂ and gas networks



Extension by other sectors requires more data!

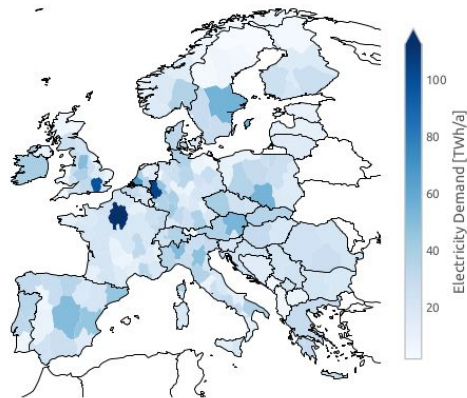


High-resolution version:

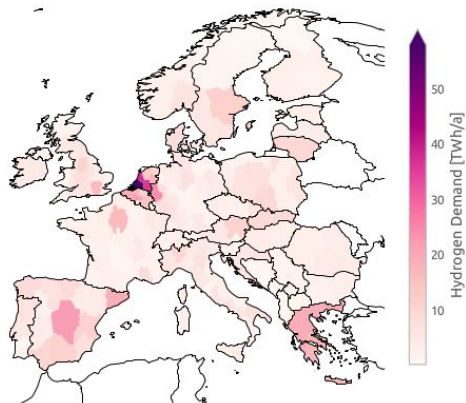
<https://tubcloud.tu-berlin.de/s/E7tx3BagXsKXLre>

Spatial distribution of energy demands

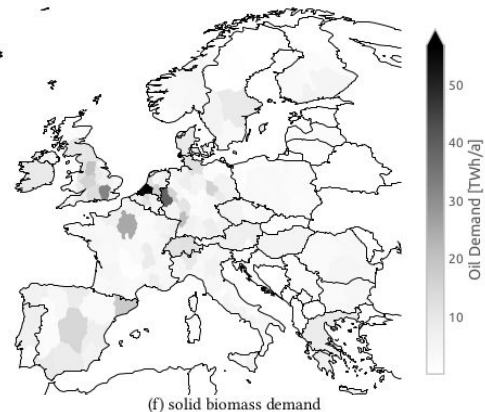
(a) electricity demand



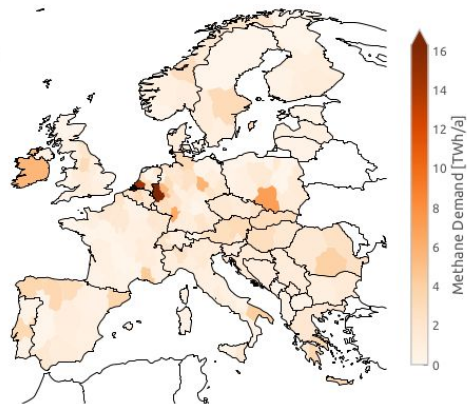
(b) hydrogen demand



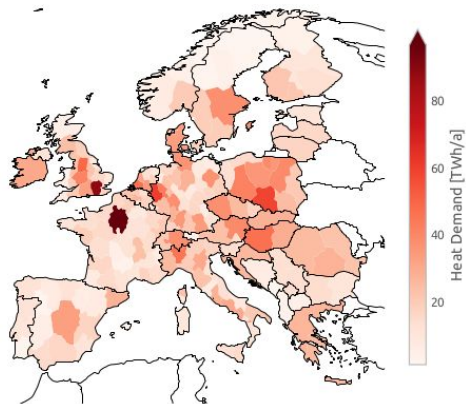
(e) oil-based product demand



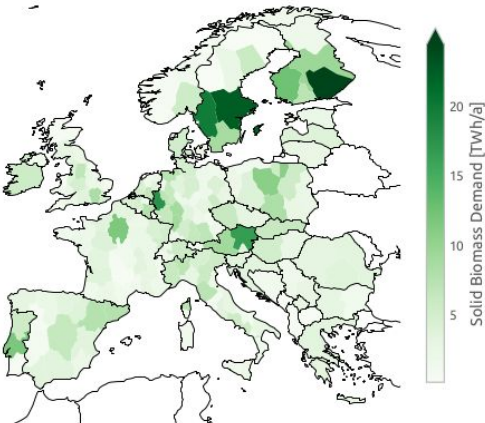
(c) methane demand



(d) heat demand



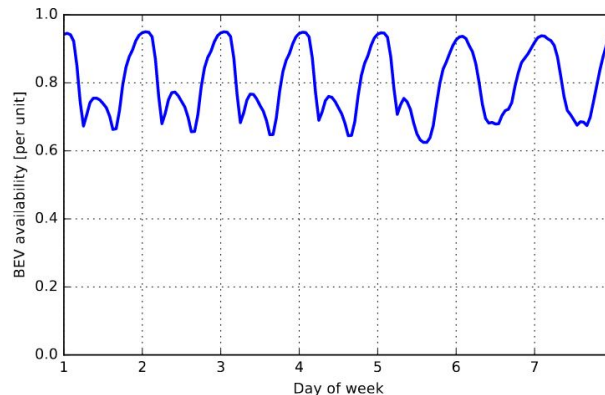
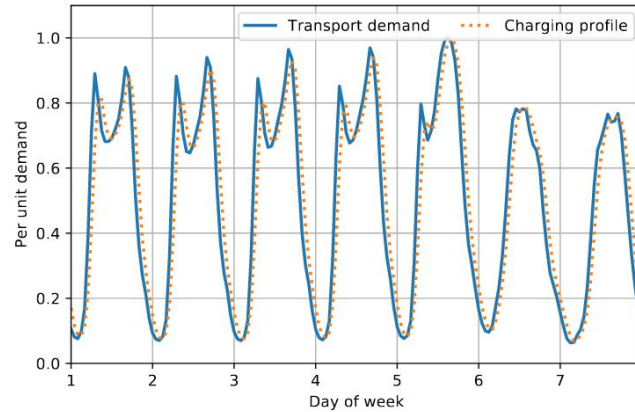
(f) solid biomass demand



Use of PyPSA components

- **Buses** impose energy balances for different energy carriers (heat, hydrogen, methane), and account for CO₂ emitted/absorbed from the atmosphere.
- **Links** represent **energy converters**, e.g. a heat pump that withdraws energy from the electricity bus and supplies heat to the heating bus with a certain efficiency.
- **Multilinks** represent more **elaborated conversion processes**, e.g. a biomass-fired CHP that withdraws biomass and supplies electricity, heat and CO₂ emissions.
- **Stores** can represent available potential (if allowed to discharge throughout the modelled period) or relax the hourly energy balancing constraint to annual (when a free store is added).

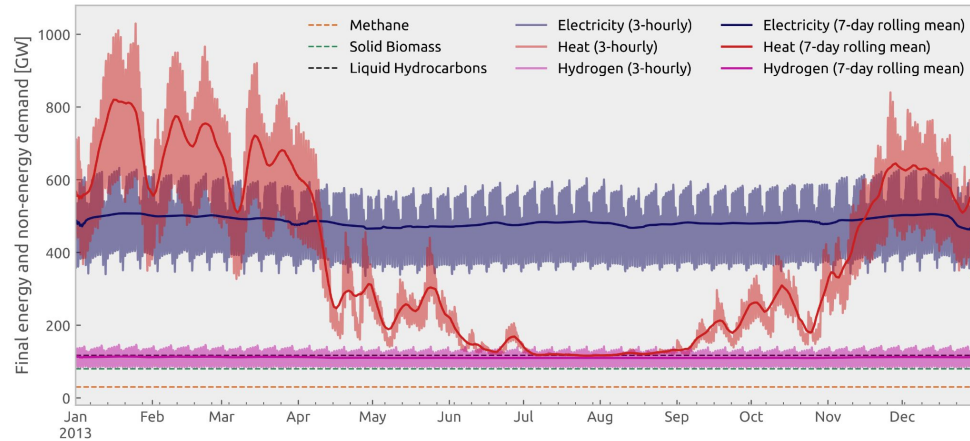
Transport - Electrification of land transport



- Road and rail are **largely electrified**, ICE only during pathway to net-zero
- Time series from traffic counting
- Lumped representation of electric vehicles: a configurable share of them participates in demand-side management (**DSM**) or vehicle-to-grid (**V2G**)
- Electric vehicles have **time-varying availability** for charging depending on driving profiles
- EV battery almost fully charged in the

Modelling **land transport transformation** endogenously is currently under development!

Heating - Demand



Heating demand time series created by :

- space heating based on degree-day and daily profiles, weighted by population
 - constant water heating demand
 - scaled to historical annual demand
 - split into individual and central systems
- (cooling demand already electrified)

Heating - Tech for individual & district heating

Decentral individual heating

can be supplied by:

- air- or ground-sourced heat pumps
- resistive heaters
- gas / oil / biomass / hydrogen boilers
- solar thermal
- small water tanks

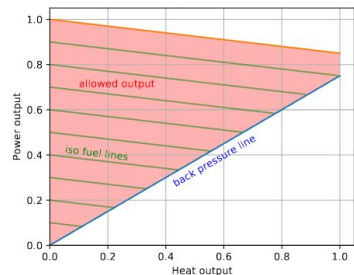
Building renovations can be co-optimized to reduce space heating demand.

District heating systems

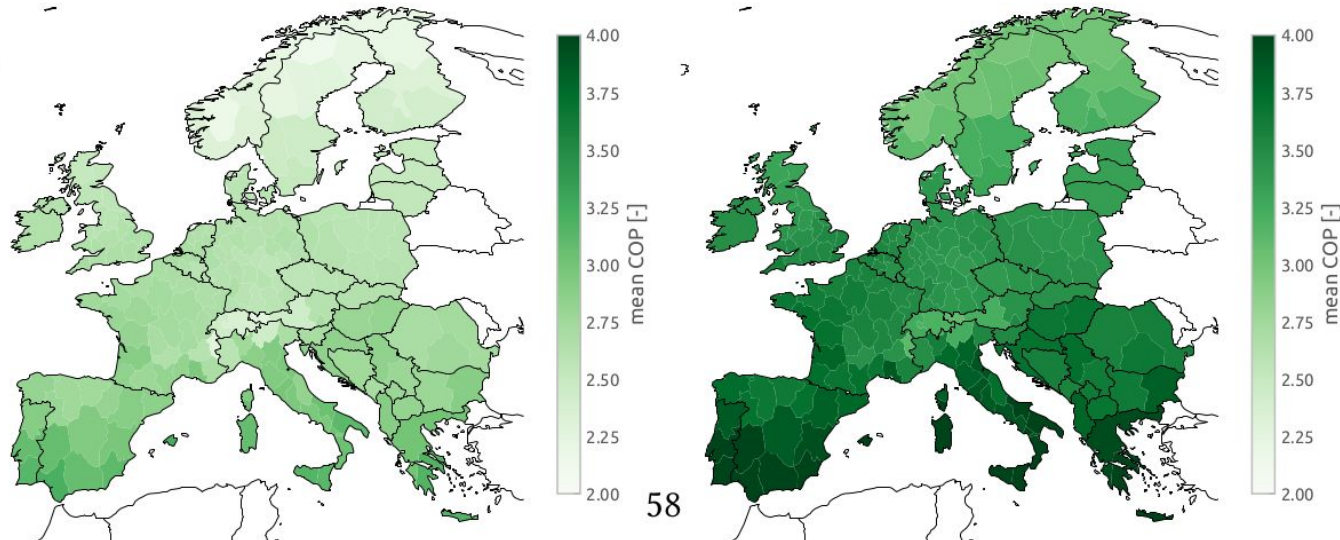
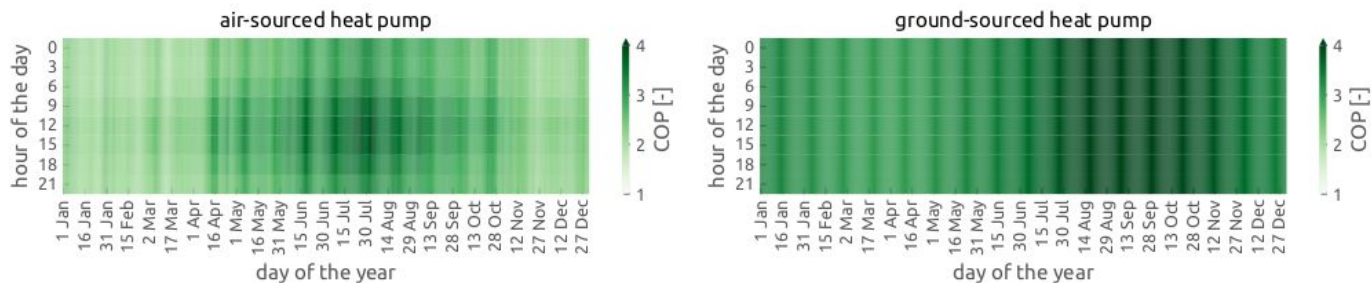
can be supplied in urban areas by:

- air-sourced heat pumps + other
- resistive heaters
- gas / hydrogen / biomass / waste CHPs
- gas / oil / biomass / hydrogen boilers
- solar thermal
- long-duration hot water storage
- waste heat from industrial processes

CHP feasible dispatch:

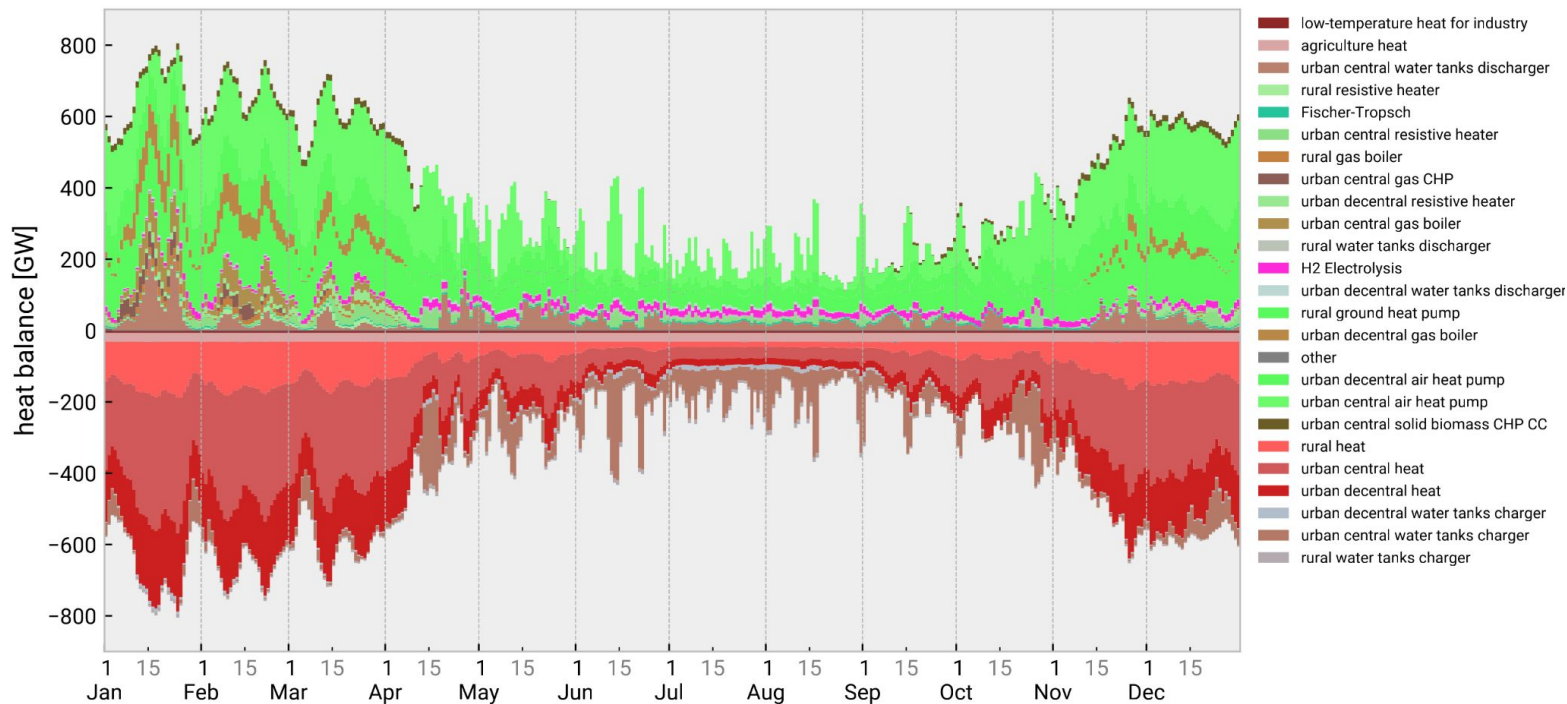


Heating - Heat pumps as new variable supply tech



Geothermal heat
sources have
been integrated
very recently!

Heating - Example daily heat system balance

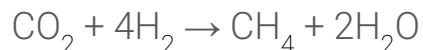


There are difficult periods in winter with **low** wind and solar, **high** space heating demand and **low** air temperatures, which are bad for air-sourced heat pump performance. In this case **gas boilers** and **CHP plants** jump in as backup.

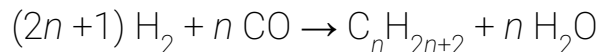
Synthetic fuels - Power-to-X

Carbonaceous fuels produced synthetically:

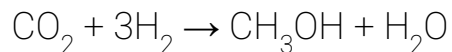
- Synthetic methane using Sabatier reaction



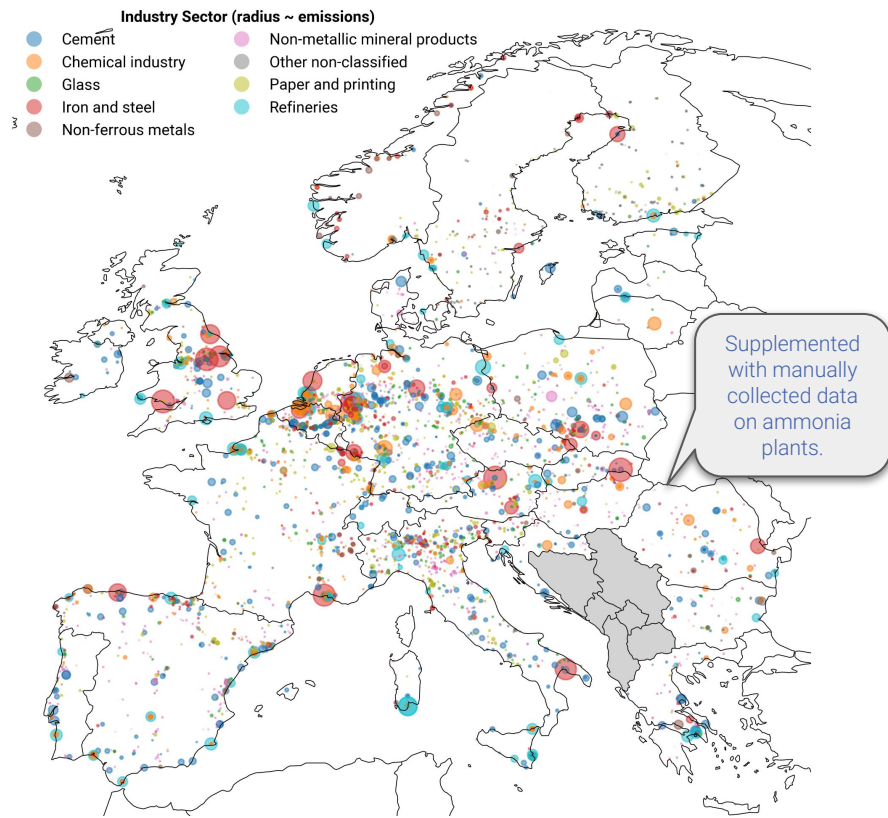
- Synthetic oil via Fischer-Tropsch



- Synthetic methanol in methanolisation plants



Industry - Regionalisation based on Hotmaps



https://gitlab.com/hotmaps/industrial_sites/industrial_sites_Industrial_Database

Iron & Steel	Phase-out integrated steelworks; increased recycling; rest from H ₂ -DRI + EAF
Aluminium	Methane for high-enthalpy heat; increased recycling
Cement	Solid biomass; capture of CO ₂ emissions
Ceramics	Electrification
Ammonia	Gray, blue, green hydrogen
Plastics	Synthetic naphtha; MtO/MtA, increased recycling
Other industry	Electrification; process heat from biomass
Shipping	Methanol, (oil), (liquid hydrogen), (LNG)
Aviation	Kerosene from Fischer-Tropsch or methanol

Modelling **industry relocation, high-temperature heat source & shipping fuels** endogenously is currently under development!

Infrastructure - Gas network with H₂ retrofitting

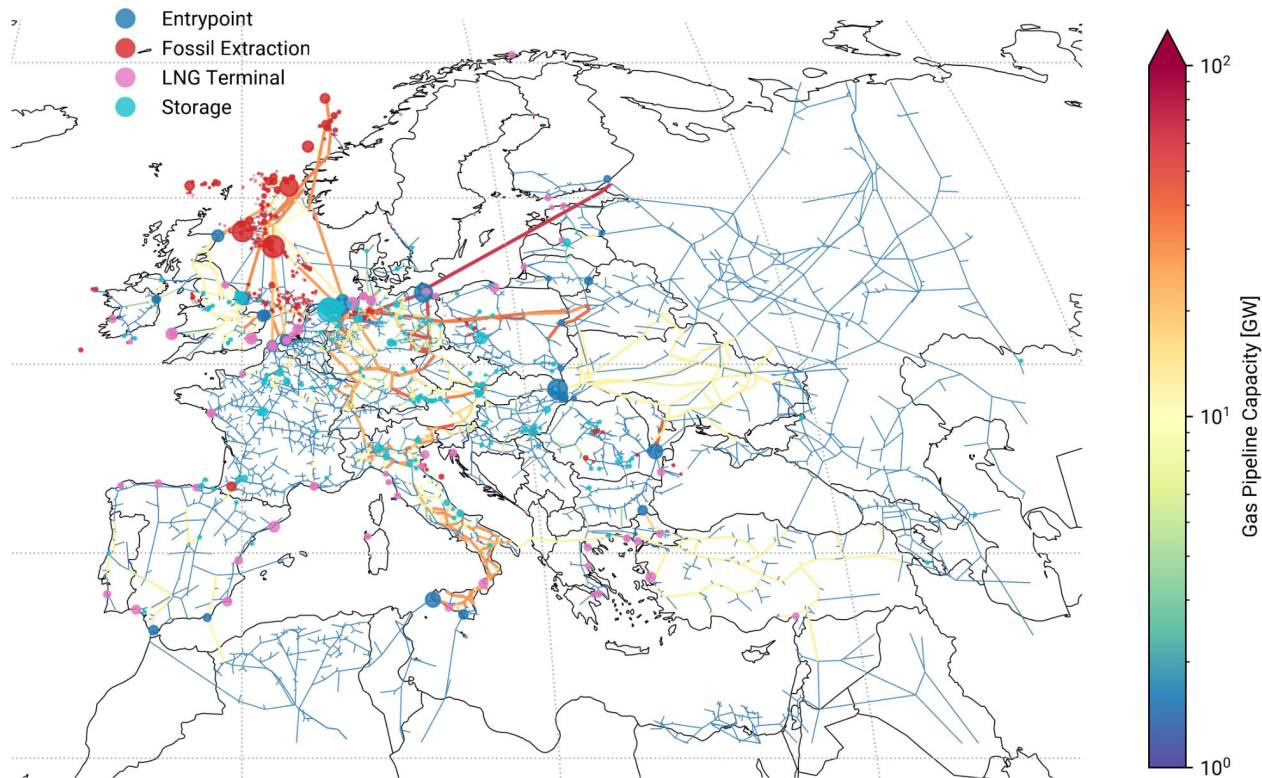
Compiled from open
SciGRID_gas dataset.

Fossil gas enters at **LNG terminals** or **gas fields**.

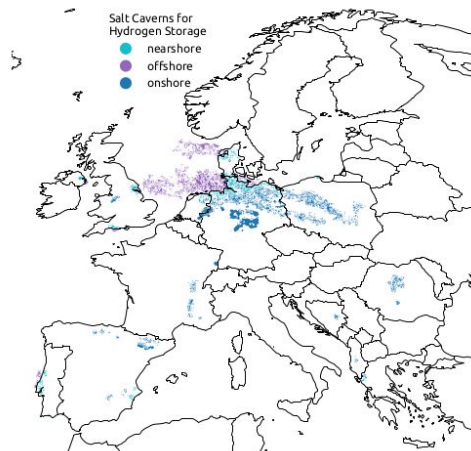
Gas flow **physics** and **valve control** neglected transport model.

Electricity demand for **compression** and **leakage** configurable.

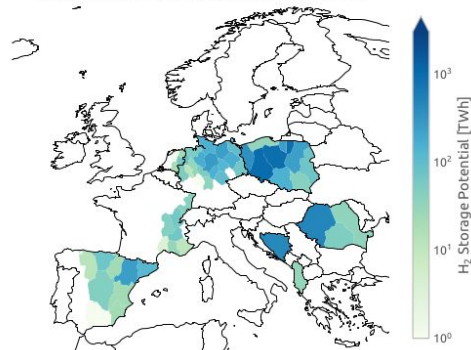
Pipelines can be **retrofitted** to H₂ with costs from **EHB**.



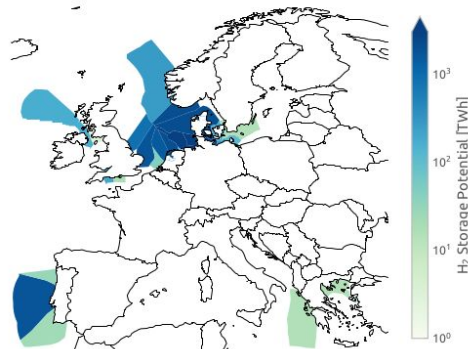
Infrastructure - Hydrogen storage potentials



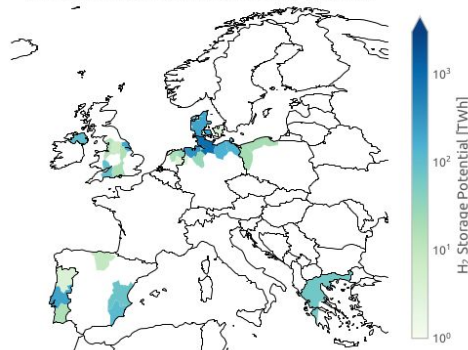
Onshore Salt Cavern H₂ Storage Potentials



Offshore Salt Cavern H₂ Storage Potentials



Nearshore Salt Cavern H₂ Storage Potentials

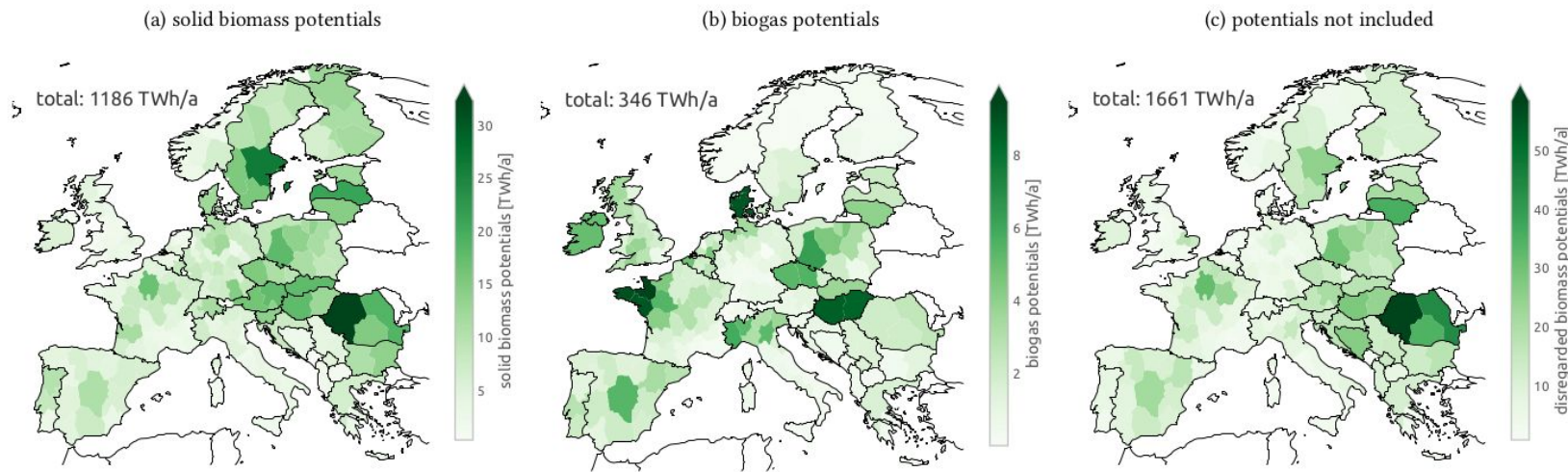


The regional distribution of **geological potential** to store hydrogen in **salt caverns** is considered.

The user can **configure** if onshore and/or offshore potentials can be used.

Dilara Gulcin Caglayan, Nikolaus Weber, Heidi U. Heinrichs, Jochen Linßen, Martin Robinius, Peter A. Kukla, Detlef Stolten, *Technical potential of salt caverns for hydrogen storage in Europe*, **International Journal of Hydrogen Energy**, Volume 45, Issue 11, 2020, 6793-6805, <https://doi.org/10.1016/j.ijhydene.2019.12.161>

Infrastructure - Biomass from JRC ENSPRESO

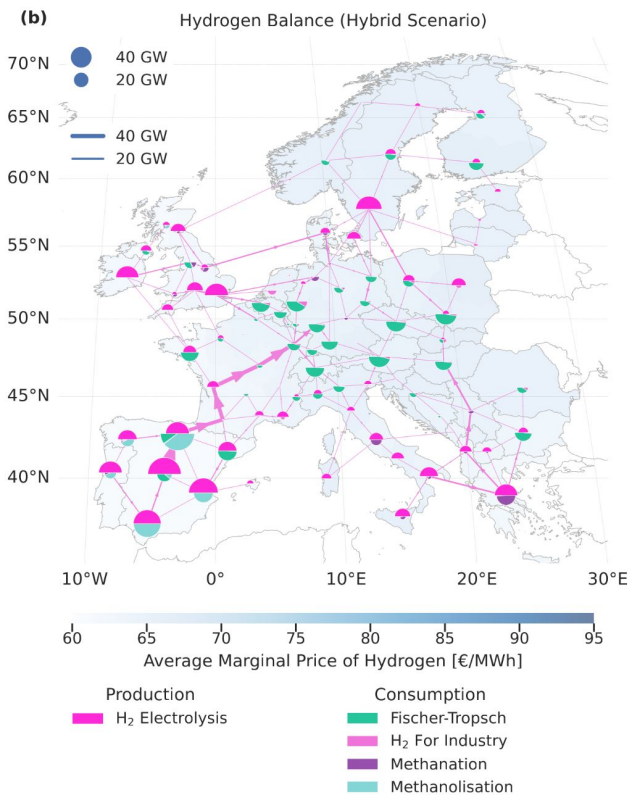
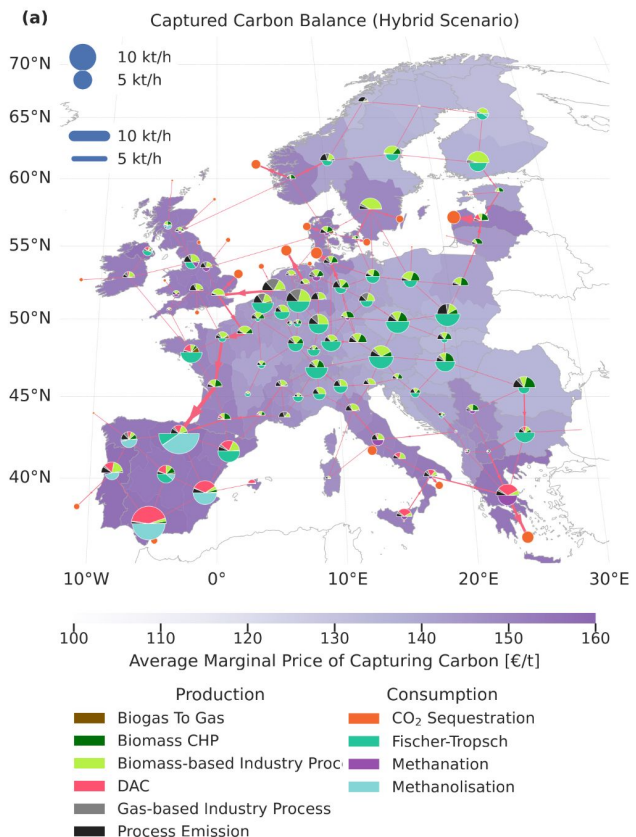


Biomass potentials are split between **solid biomass** and **biogas** (which can be, for instance, upgraded).

The user can configure low/medium/high potentials and what categories of biomass to consider (e.g. forest residues).

The default configuration only considers **residual biomass**, no energy crops.

Infrastructure - Carbon management

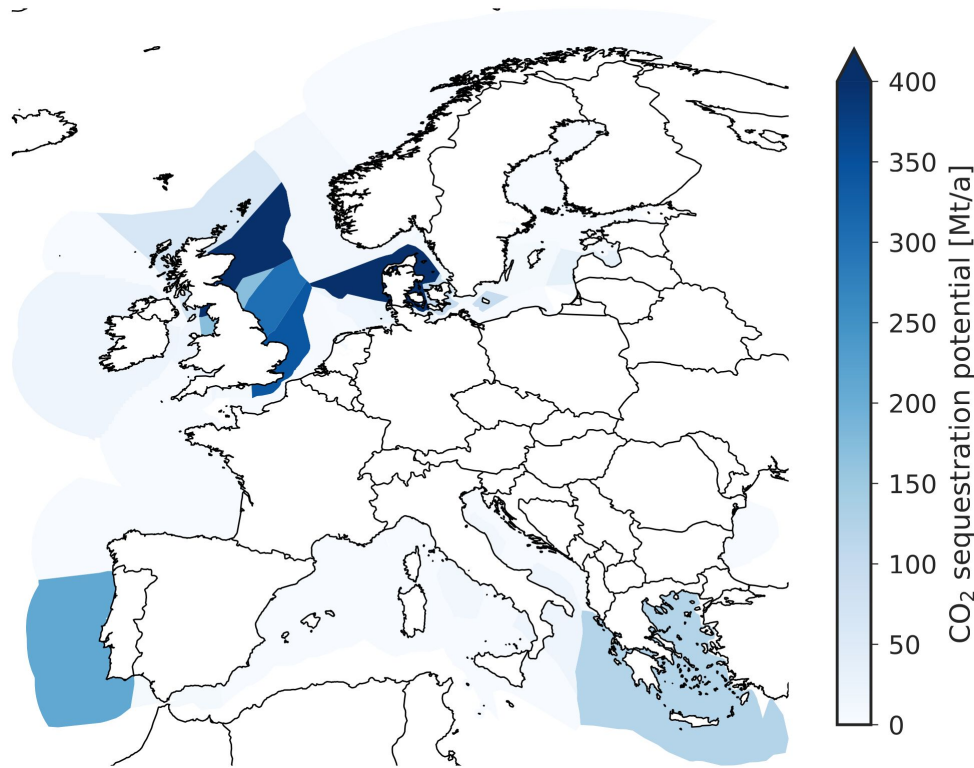


Built-in carbon flows:

- **Capture:**
DAC, process emissions, fossil / biomass CHP
- **Transport:**
CO₂ pipelines
- **Storage:**
intermediate storage and long-term geological sequestration
- **Utilization:**
for synthetic carbonaceous fuels

Infrastructure - Carbon sequestration potentials

Example: Offshore carbon sequestration potentials

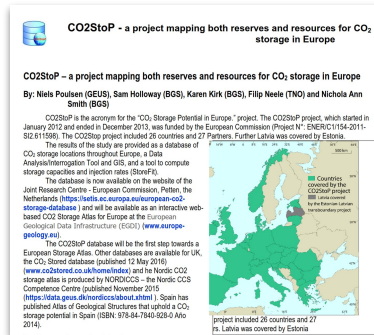


The user can **configure**

- onshore/offshore sequestration,
- gas fields/oil fields/aquifer, and
- low/medium/high potentials,

as well as a **total limit** on the annual sequestration, e.g. 250 Mt per year.

Data source:



https://energy.ec.europa.eu/publications/assessment-co2-storage-potential-europe-co2stop_en

Technology choices - endogenous vs. exogenous

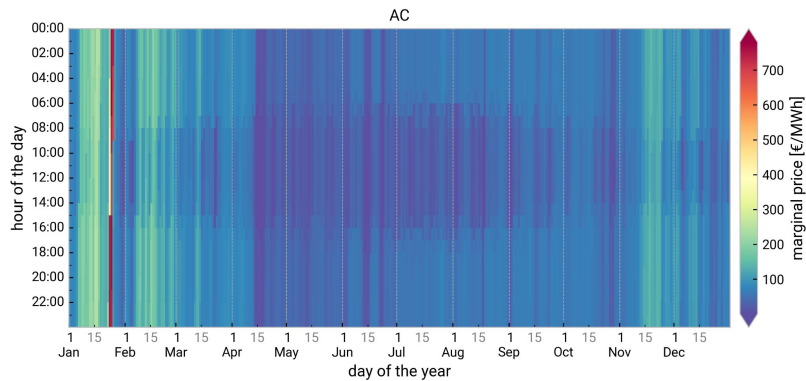
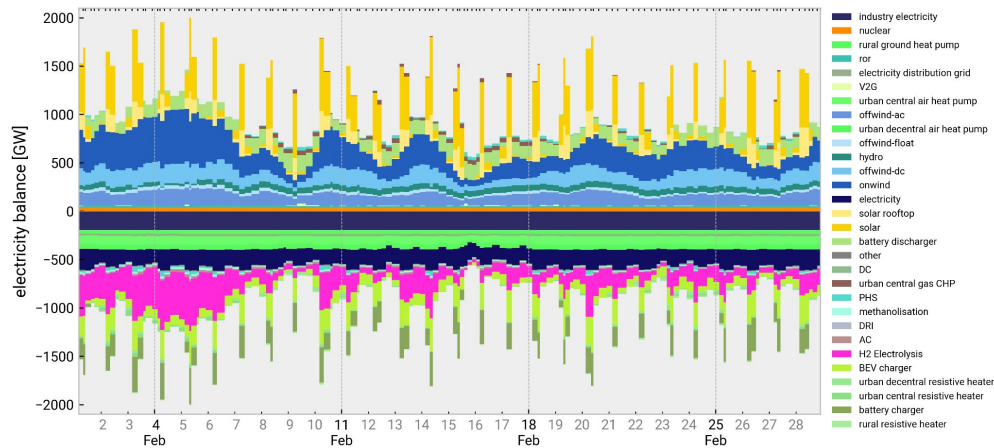
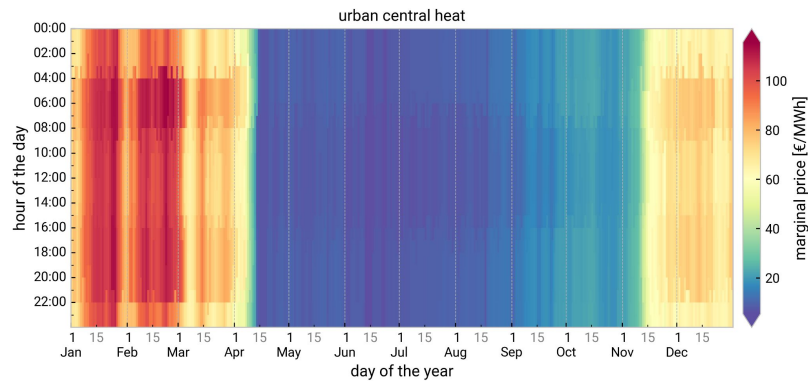
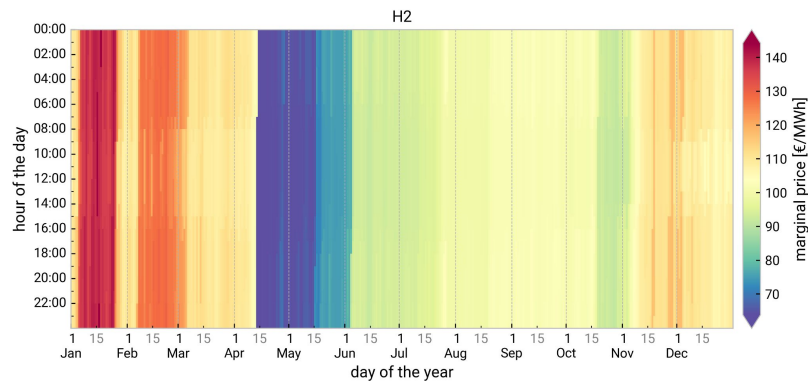
Exogenous assumptions (modeller chooses):

- energy services demand (e.g. heat)
- district heating shares
- energy carrier shares for road transport
- kerosene for aviation
- methanol for shipping
- electrification & recycling in industry
- steel production with DRI + EAF

Endogenous choices (model optimizes):

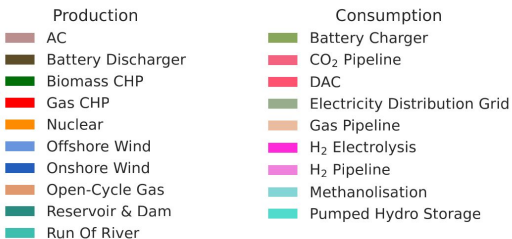
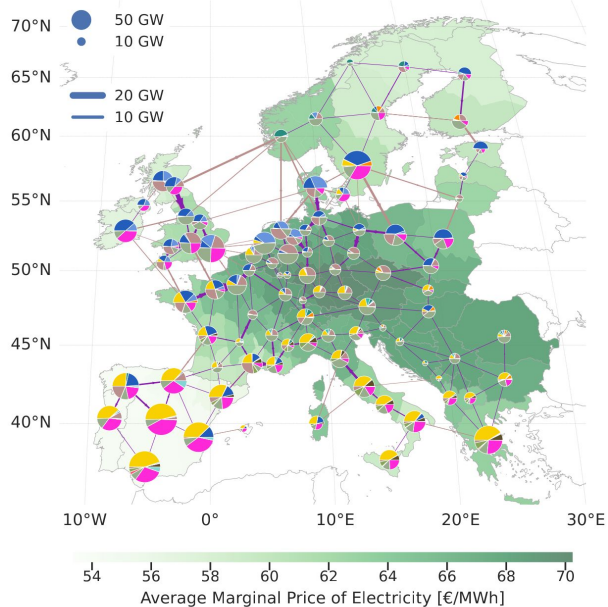
- change in electricity generation fleet
- transmission reinforcement
- capacities and locations of short and long-duration energy storage
- space and water heating technologies (including building renovations)
- all P2G/L/H/C
- supply of process heat for industry
- carbon capture (e.g. CHP, industry, DAC)

Prices in the model

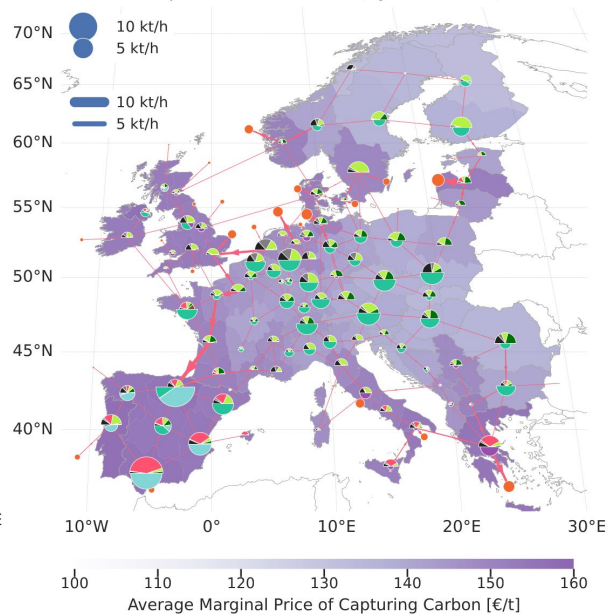


Prices in the model

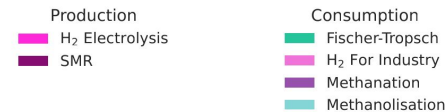
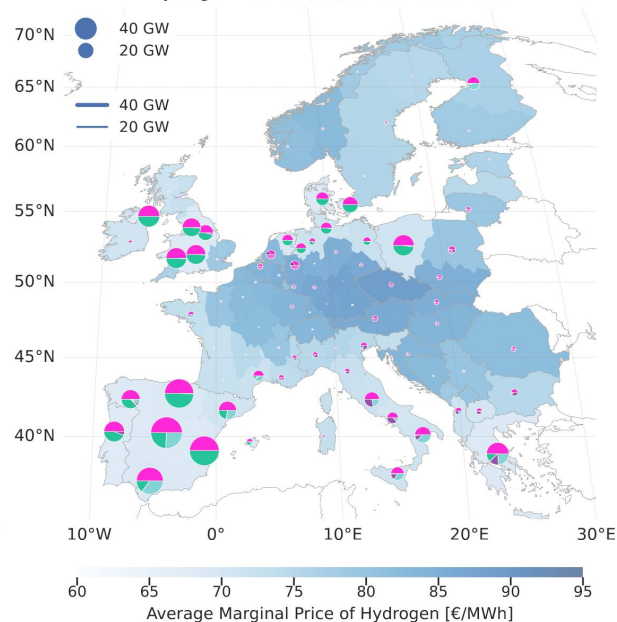
Electricity Balance (Hybrid Scenario)



Captured Carbon Balance (Hybrid Scenario)



Hydrogen Balance (CO₂-Grid (NN) Scenario)



Supply, consumption and storage options by carrier

Electricity (115 regions)

Supply	Withdrawal
rooftop solar	industry electricity
utility-scale solar	residential electricity
onshore wind	services electricity
offshore wind (fixed-pole/floating, AC/DC-connected)	agriculture electricity
nuclear	air-sourced heat pump
hydro reservoirs	ground-sourced heat pump
pumped-hydro	resistive heater
run-of-river	electric vehicle charger
import by HVDC link	battery charger
gas CHP (w/wo CC)	pumped-hydro
biomass CHP (w/wo CC)	hydrogen pipeline (compression)
gas turbine (OCGT)	direct air capture
methanol turbine (OCGT)	Haber-Bosch
hydrogen turbine (OCGT)	electric arc furnace
hydrogen fuel cell CHP	direct iron reduction
battery discharger	distribution grid losses
vehicle-to-grid	transmission grid losses
	methanolisation
	electrolysis
Grids & Storage	distribution grid
	transmission grid
	battery storage
	pumped-hydro storage
	electric vehicles

Hydrogen (115 regions)

Supply	Withdrawal
import by pipeline	Fischer-Tropsch
import by ship	methanolisation
electrolysis	electrobiofuels
chlor-alkali electrolysis (exogenous)	direct iron reduction
steam methane reforming (w/wo CC)	Haber-Bosch
ammonia cracker	hydrogen turbine (OCGT)
	hydrogen fuel cell CHP
	methanol-to-kerosene
	Sabatier
Grids & Storage	new pipelines
	retrofitted pipelines
	storage in salt caverns
	storage in steel tanks

Methane (not spatially resolved)

Supply	Withdrawal
import by ship	gas for high-T industry heat (w/wo CC)
fossil gas	steam methane reforming (w/wo CC)
biogas upgrading (w/wo CC)	gas boiler (rural/urban)
Sabatier	gas CHP
	gas turbine (OCGT)
Storage	hydrocarbon storage

Liquid Hydrocarbons (not spatially resolved)

Supply	Withdrawal
import by ship	kerosene for aviation
fossil oil refining	naphtha for industry
Fischer-Tropsch	diesel for agriculture
electrobiofuels	
Storage	hydrocarbon storage

Methanol (not spatially resolved)

Supply	Withdrawal
import by ship	methanol turbine (OCGT)
methanolisation	methanol for shipping
	methanol for industry
	methanol-to-kerosene
Storage	hydrocarbon storage

Ammonia (not spatially resolved)

Supply	Withdrawal
import by ship	ammonia cracker
Haber-Bosch	ammonia for fertilizer
Storage	ammonia tank

Supply, consumption and storage options by carrier

Heat (115 regions)

Supply	Withdrawal
air-sourced heat pump	residential heat
ground-sourced heat pump (only rural)	services heat
resistive heater	agriculture heat
gas boiler	low-T industry heat
biomass boiler	direct air capture
solar thermal	water tank charger
water tank discharger	
biomass CHP (w/wo CC, only DH)	
gas CHP (w/wo CC, only DH)	
hydrogen fuel cell CHP (only DH)	
electrolysis (only DH)	
Haber-Bosch (only DH)	
Sabatier (only DH)	
Fischer-Tropsch (only DH)	
methanolisation (only DH)	
Storage	long-duration thermal storage (only DH) hot water tank

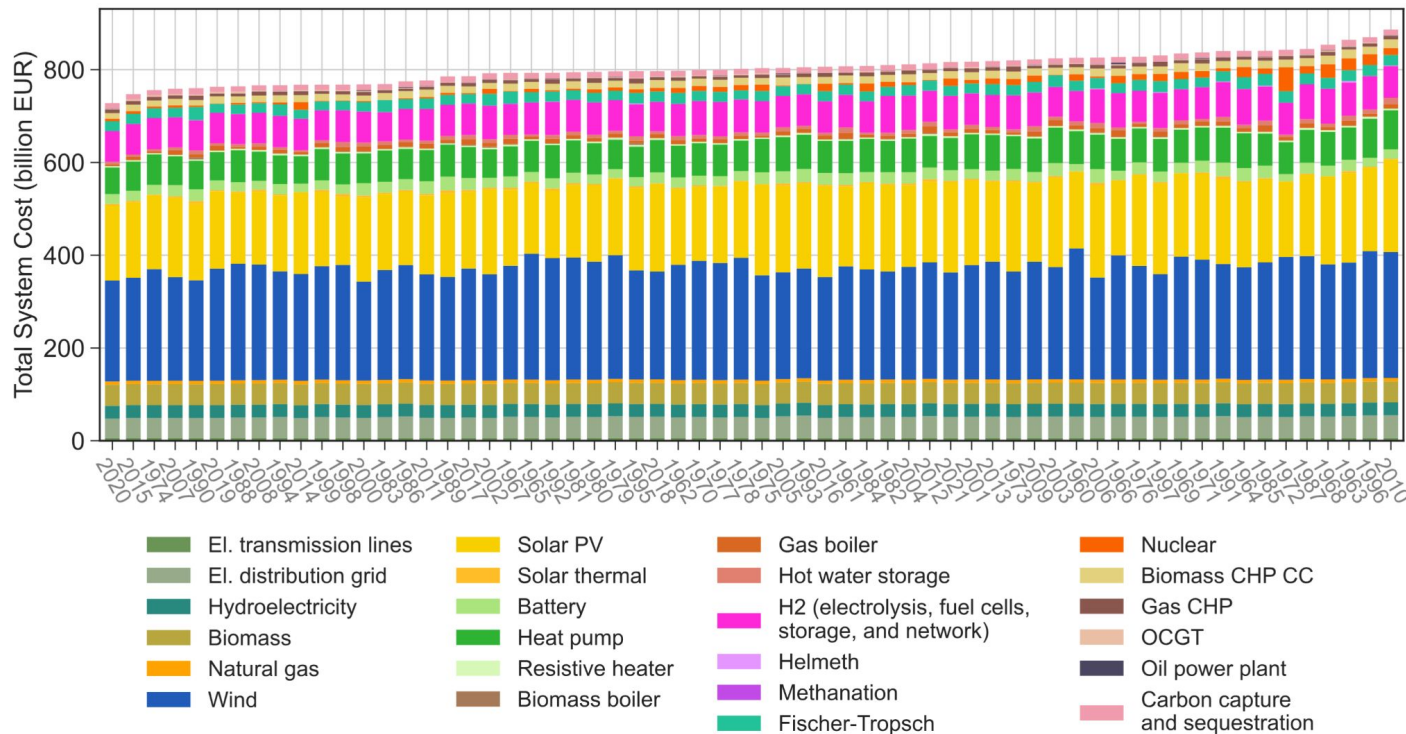
CO2 atmosphere (not spatially resolved)

Supply	Withdrawal
kerosene for aviation	solid biomass for industry (w CC)
diesel for agriculture	solid biomass CHP (w CC)
methanol for shipping	biogas upgrading (w CC)
methanol for industry	direct air capture
naphtha for industry	electrobiofuels
gas boiler	
gas CHP (w/wo CC)	
gas turbine (OCGT)	
methanol turbine (OCGT)	
process emissions (w/wo CC)	
fossil oil refining	
gas for high-T industry heat (w/wo CC)	
steam methane reforming (w/wo CC)	

CO2 commodity (not spatially resolved)

Supply	Withdrawal
direct air capture	Fischer-Tropsch
biogas upgrading (w CC)	methanolisation
gas CHP (w CC)	sequestration
biomass CHP (w CC)	Sabatier
steam methane reforming (w CC)	
process emissions (w CC)	
solid biomass for industry (w CC)	
gas for high-T industry heat (w CC)	
Storage	intermediate storage in steel tank long-term geological sequestration

PyPSA-Eur can be run on different **weather years**!



The years **2010, 2013, 2019** and **2023** are currently available “out of the box”.

Other years **1940-2024** require a few more steps.

We are planning to expand the number of “plug-and-play” years.

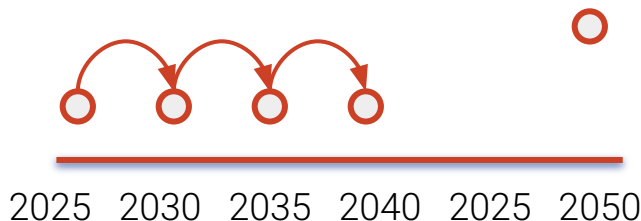
Overnight and pathway optimization

- **Overnight** optimization builds the complete system in one go. Already existing capacities can be included (**brownfield**) or the system can be built from scratch (**greenfield**)
- The choice of **investment years** is arbitrary.
- **Pathway optimization with myopic foresight** (next slide). The CO₂ emission **reduction path** is exogenous.
- **Perfect foresight pathway planning** is currently experimental (i.e. use the CO₂ budget endogenously).

Overnight

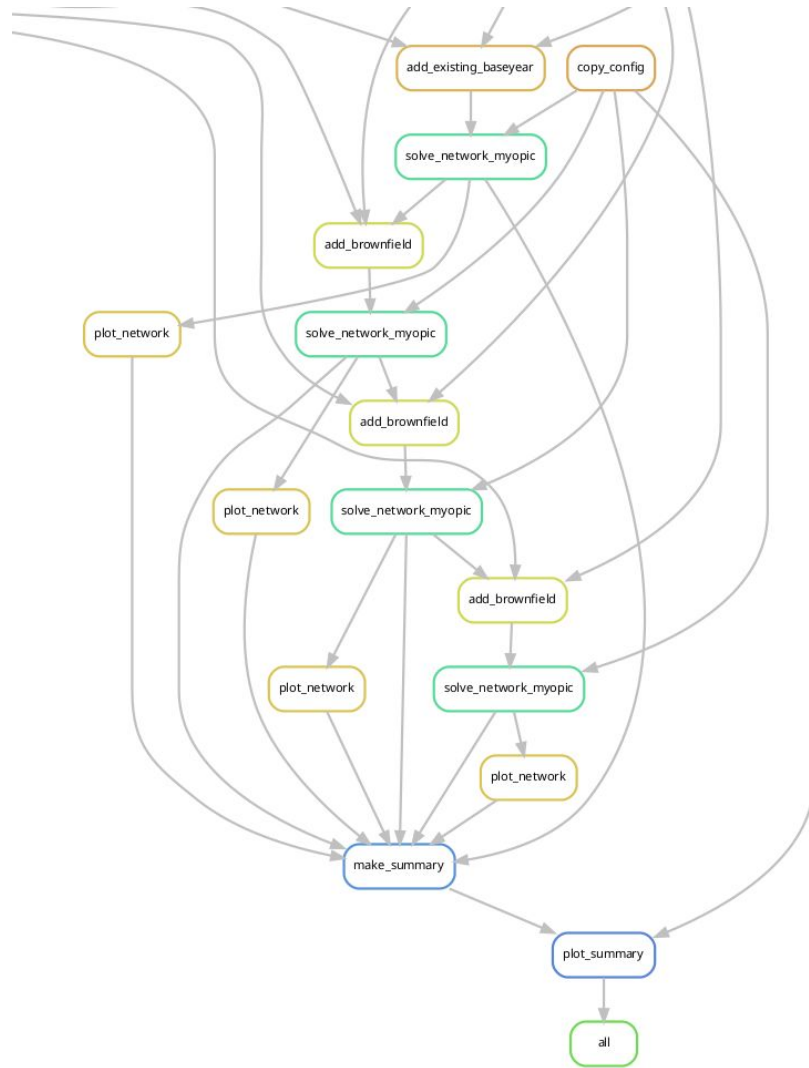
Myopic pathway

Perfect-foresight pathway



Myopic pathway optimization

- The CO₂ emission **reduction path** is exogenous.
- Optimise **start network** for e.g. 2025, starting with existing energy infrastructure.
- Take results from **2025 as input** for 2030 infrastructure optimisation, take 2030 results for next iteration, etc.
- Infrastructure decommissioned when reaching end of lifetime
- Configurable social discount rate and number of planning horizons



Scenario management

PyPSA-Eur has integrated & scalable scenario management.

config/config.yaml

```
run:
  name: all
  scenarios:
    enable: true

scenario:
  clusters: [90]

sector:
  H2_network: true
  gas_network: true
  H2_retrofit: true

electricity:
  transmission_limit: vopt
```

With these two files
configured, run:

```
$ snakemake all -n
```

and

```
$ snakemake all
```

config/scenarios.yaml

```
no-h2-network:
  sector:
    H2_network: false

no-grid-expansion:
  electricity:
    transmission_limit: v1.0

no-to-both:
  sector:
    H2_network: false
  electricity:
    transmission_limit: v1.0

yes-to-both:
  sector:
    H2_network: true
  electricity:
    transmission_limit: vopt
```

Live Demo – very similar to electricity-only case

Start with a dry-run:

```
$ snakemake all --configfile config/test/config.overnight.yaml -n
```

Then execute the same command “for real” by dropping “-n” flag:

```
$ snakemake -call all --configfile config/test/config.overnight.yaml
```

And for myopic pathway optimisation:

```
$ snakemake -call all --configfile config/test/config.myopic.yaml
```

To explore results, start a Jupyter notebook:

```
$ jupyter notebook
```


Practical Phase

(sector-coupled)

1) Run PyPSA-Eur sector-coupling tutorial with **snakemake**

Guide:

https://pypsa-eur.readthedocs.io/en/latest/tutorial_sector.html

Users of Windows, add two lines to YAML:

```
run:  
  use_shadow_directory: false
```

```
snakemake -call all --configfile config/test/config.overnight.yaml
```

2) Explore CSV files and images in **results** directory.

Go to <https://pypsa-eur.readthedocs.io/en/latest/configuration.html> and try to find out how to configure some of the settings for **sector-coupled models** listed below:

1. Disable vehicle-to-grid discharging.
2. Disable hydrogen network expansion.
3. Increase the carbon sequestration potential to 800 Mt/a.
4. Allow hydrogen underground storage also onshore.
5. Reduce the primary production of plastics by increasing recycling rates.