

# TulipaEnergy

**Faster Solving and  
Higher Detailed  
Large-Scale Energy  
System Models**

Diego Tejada | JuMP dev 2024



# TNO innovation for life

*Netherlands Organisation for Applied Scientific Research*

Creating impactful innovations for the sustainable wellbeing and prosperity of society.



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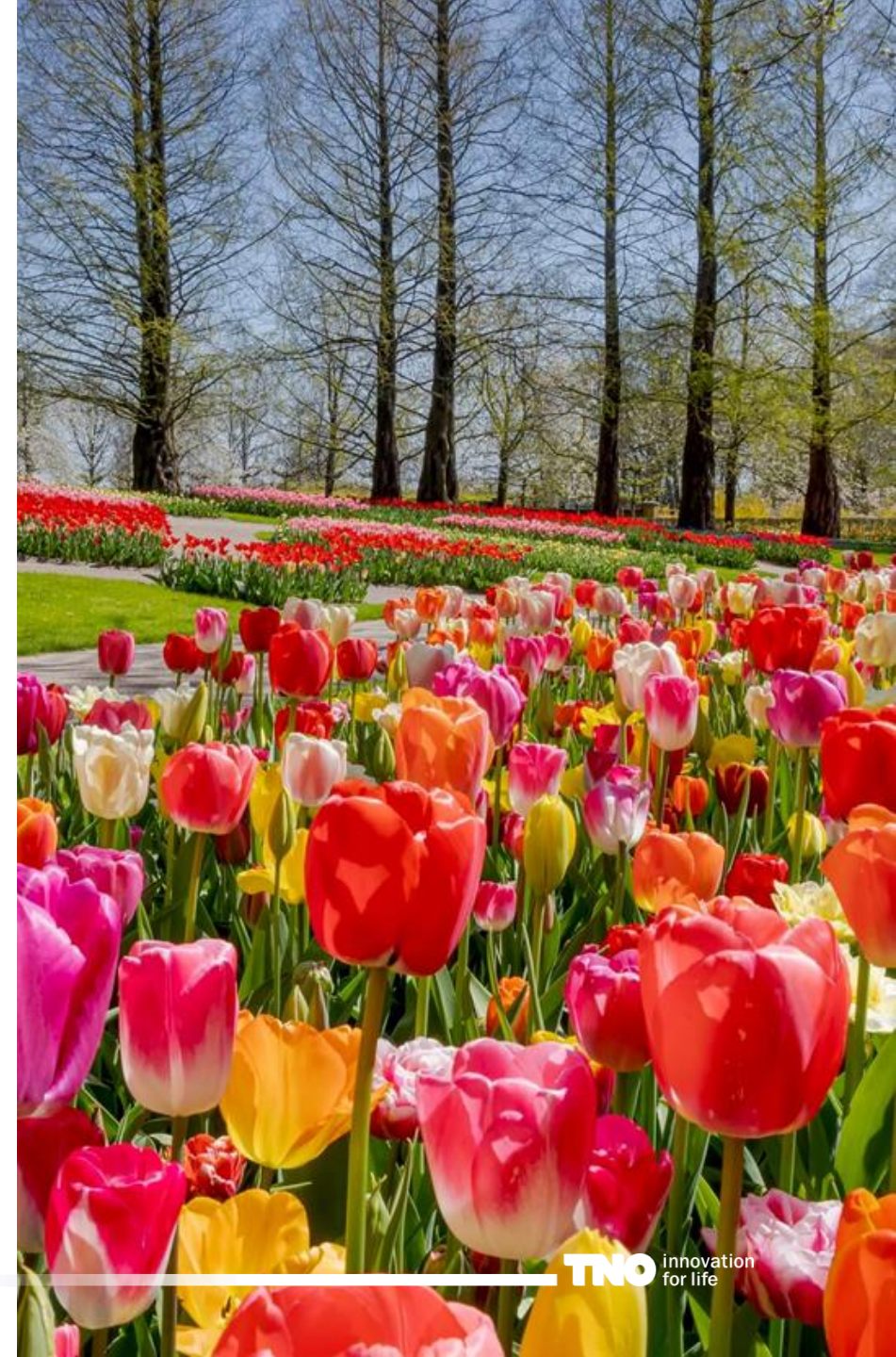
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# Yet Another Energy System Model in Julia..



# The Challenge

- Models aid in integrating renewable energy and coupling energy carriers.
- Optimizing investments helps stakeholders understand system dynamics for energy transition.
- Existing models excel in either technological, operational, or spatial-temporal detail, but never all three simultaneously.
- Key challenge: including enough details while remaining computationally tractable.

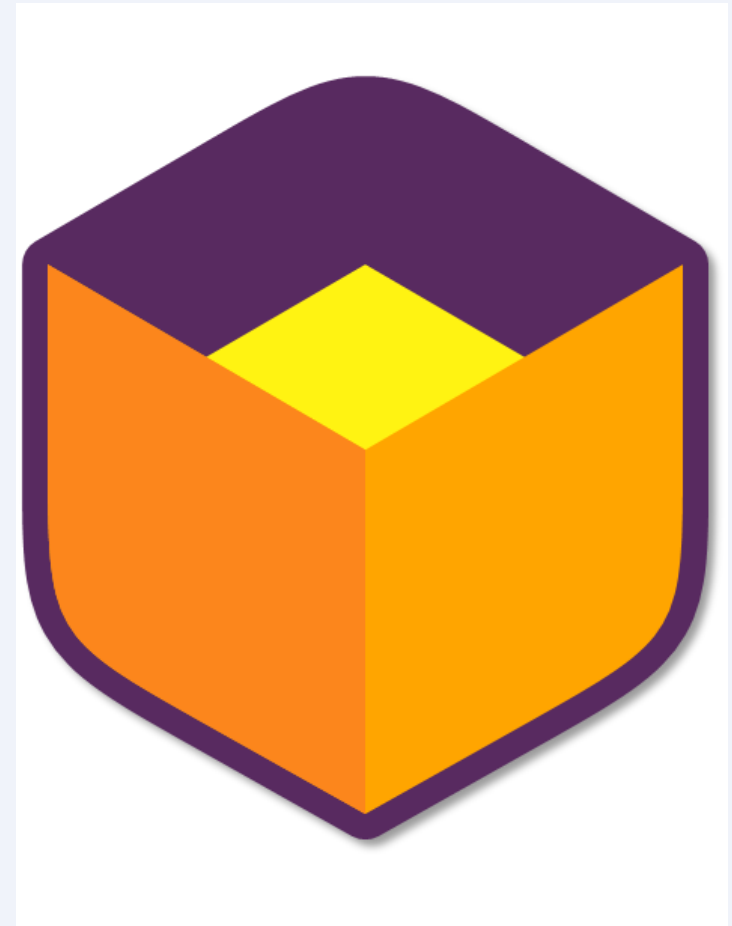


# Project Description

- New energy model from [scratch](#) (model design and coding)
- [Sector coupling](#): e.g., electricity, H2 and heat
- The main objective is to determine the optimal [investment](#) and [operation](#) decisions
- Representation of different types of [energy assets](#) (e.g., producers, consumers, conversions, storages, and transports)



Focus on [compact and efficient formulations](#) and implementations suitable for expert energy system researchers



# The TulipaEnergy Team

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 **TU**Delft



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Rep. Periods*



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*PhD: Accurate &  
Efficient Formulations*



**Utrecht  
University**



Zhi Gao  
*PhD: Fully Flexible  
Temporal Resolution*

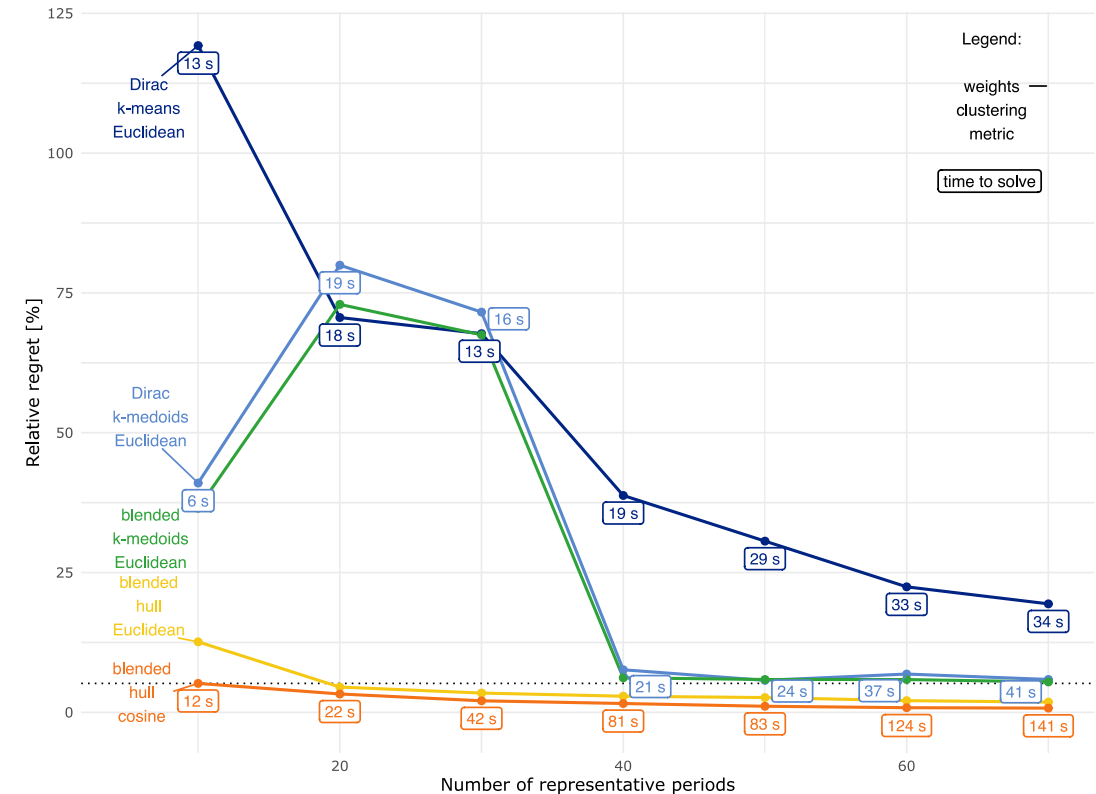
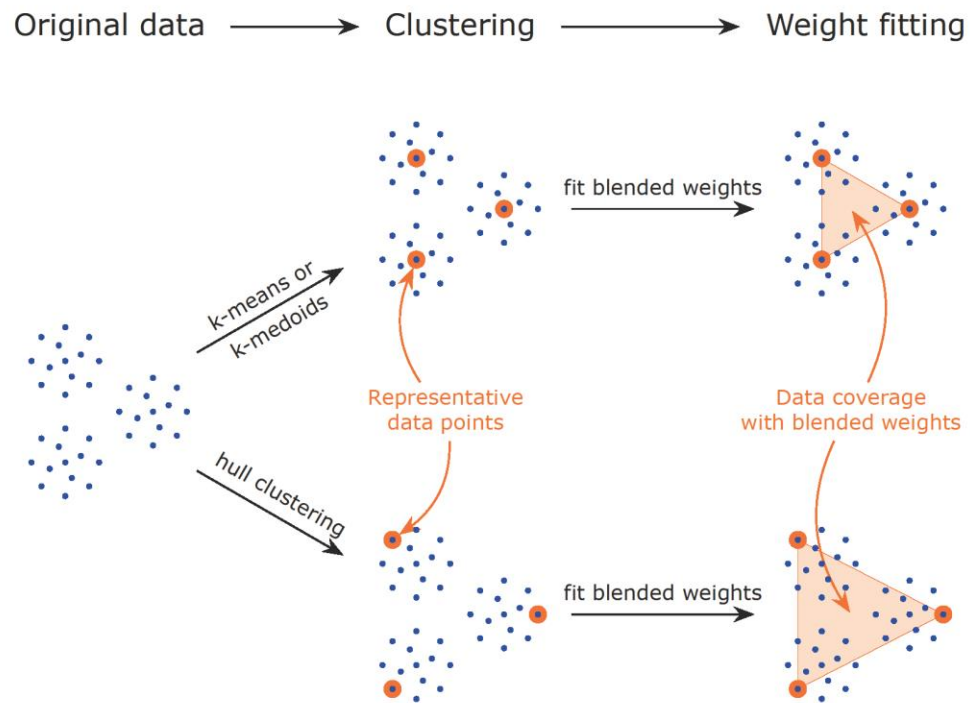


Matthijs Arnoldus  
*MSc: Modelling to  
Generate Alternatives*

# Innovations in Energy Modelling

# Hull Clustering with Blended Representative Periods

- Method of hull clustering with blended representative periods (RPs)
- Advantages over k-means/medoids in data representation
- Faster performance with lower relative regret using fewer RPs

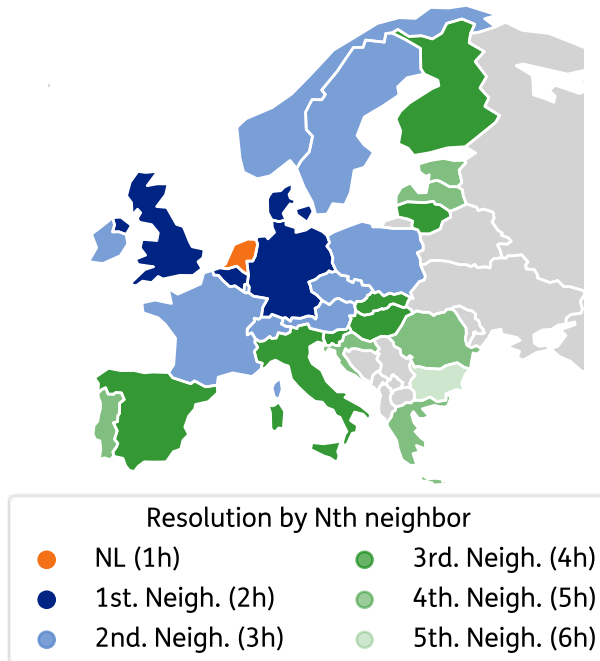




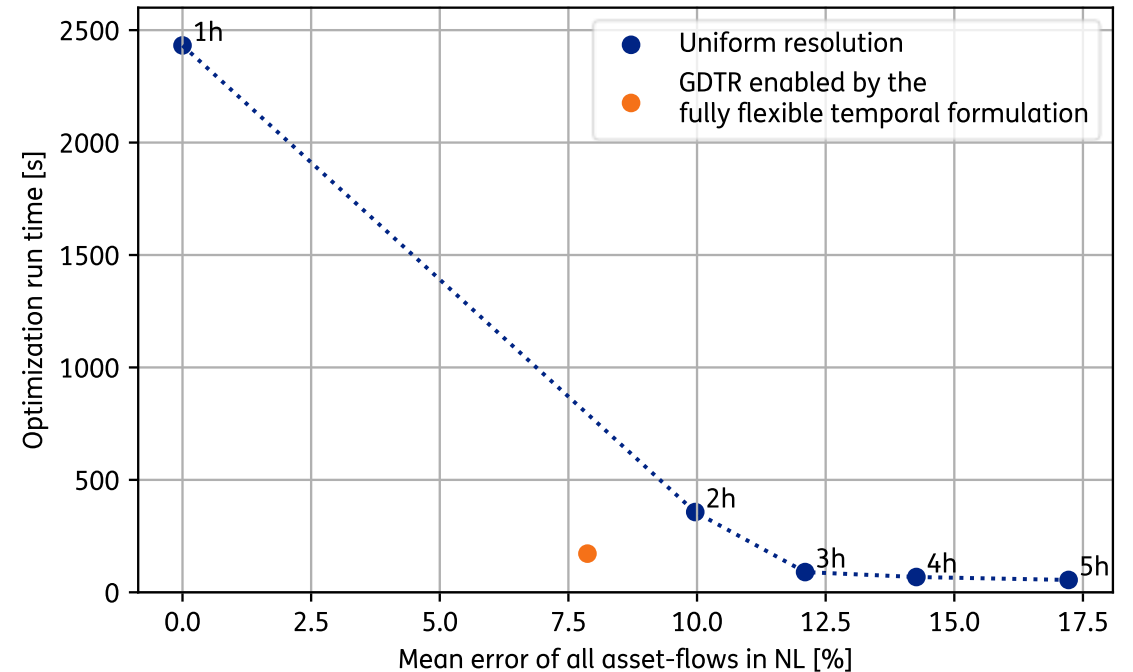
# Fully Flexible Temporal Resolution

- Flexible formulation for temporal resolution
- Capability to mix independent resolutions across carriers, regions, and time horizons
- Example of geographical application in the Netherlands

Geographically Decreasing Temporal Resolution (GDTR)

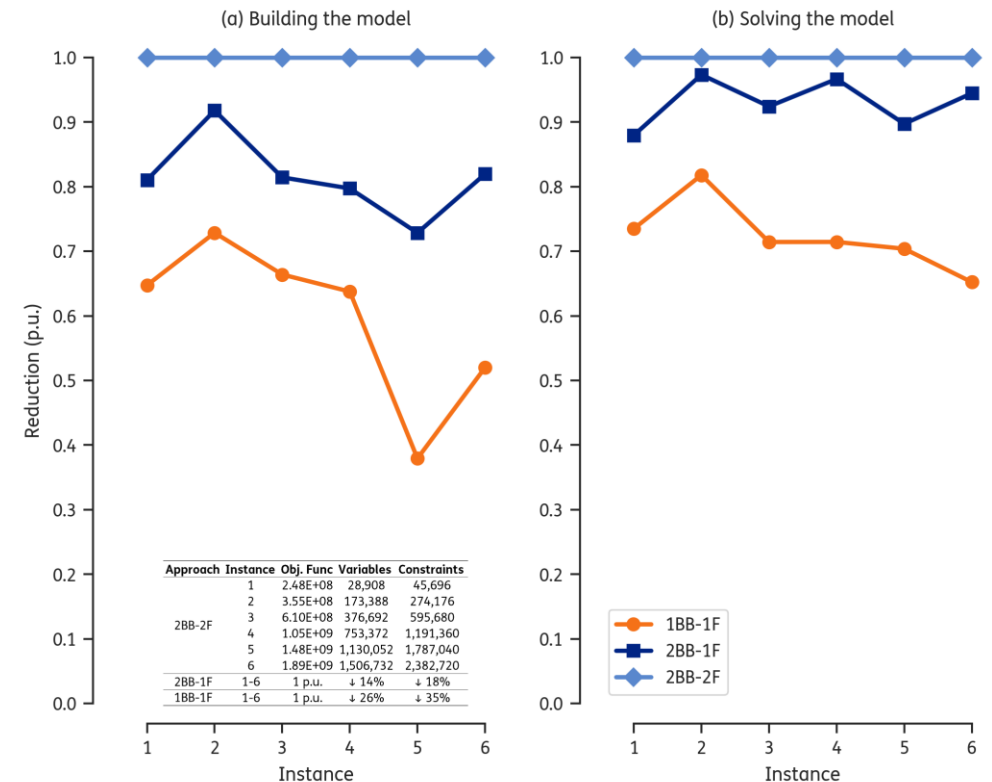
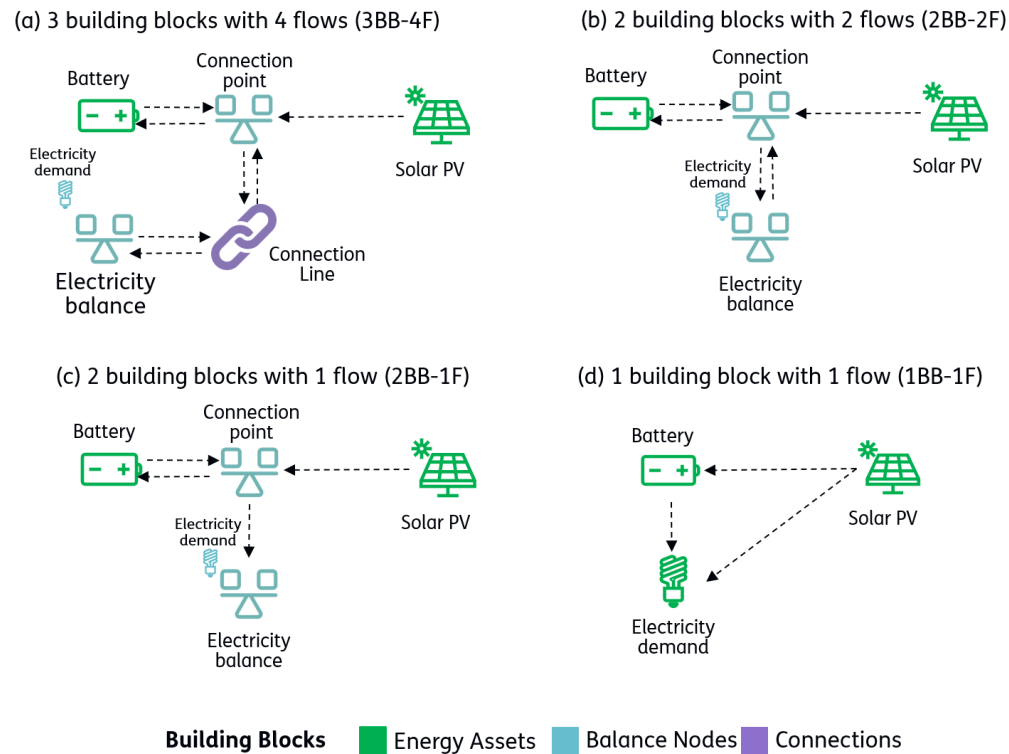


Computational cost vs solution accuracy



# Graph-Based System Representation

- Breaking the misconception of LP as the simplest representation
- Graph theory approach reducing problem size without losing accuracy
- Faster model building and solving with increasing model size



TulipaEnergy

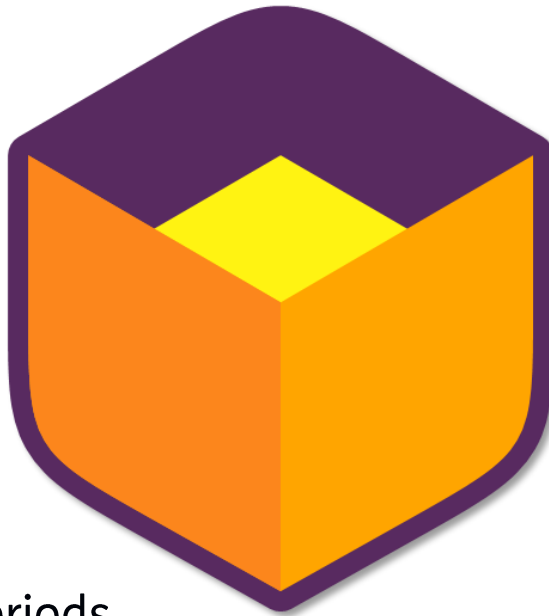
# TulipaEnergy and Julia/JuMP

TulipaEnergy

# TulipaEnergy Packages

## TulipaEnergyModel.jl

Builds and runs the optimisation model



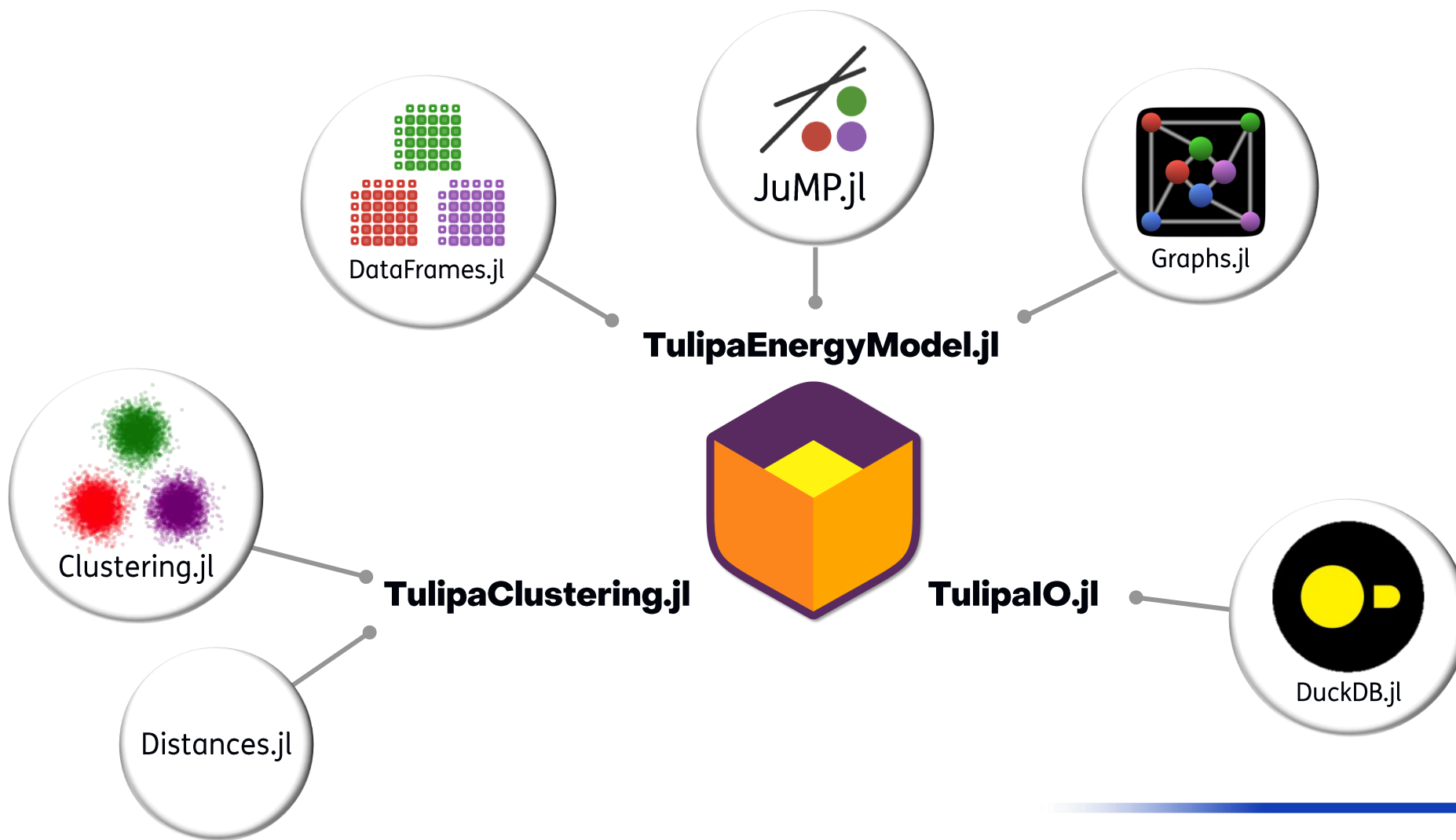
## TulipaClustering.jl

Selects the Blended Representative Periods

## TulipaIO.jl

Script-based IO for data manipulation

# TulipaEnergy Main Dependencies




Tulipa

# TulipaEnergy and Useful Tool for Development



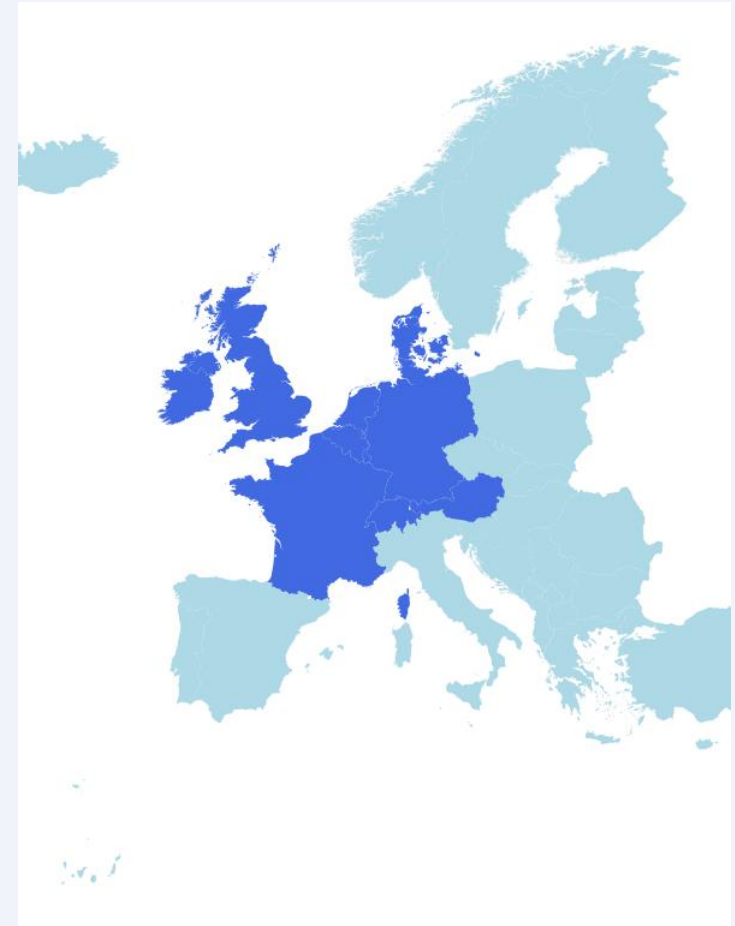
# TulipaEnergyModel.jl

- [Open-source](#) Julia/JuMP package available on GitHub 
- Timeline:
  - **2023** → Core features development and innovations
  - **2024** → Multi-year investment and power system operation constraints
  - **2025** → Operation constraints in other sectors (e.g., gas) and uncertainty
- [Applying best practices for software development](#) (e.g., atomic commits, semantic versioning, code review, tests, documentation)



# Performance Results - Western European Countries

- 10 European Countries with an hourly resolution
- Minimize operating costs for one year
- Optimization problem size:
  - # variables:  $\approx$  1.2 million
  - # constraints:  $\approx$  2.5 million
- **TulipaEnergyModel.jl** building time<sup>†</sup> and memory usage:
  - Initial code: 314s and 32GB
  - Optimized code: 86s (**↓73%**) and 18GB (**↓44%**)



<sup>†</sup>First draft of the code: 8min for 2 EU countries



# How did we achieve it?

## Basic JuMP

```

1 @constraint(
2     model,
3     max_transport_flow_limit[f ∈ Ft, rp ∈ RP, B_flow ∈ graph[f...].partitions[rp]],
4     duration(B_flow, rp) * flow[f, rp, B_flow] ≤ upper_bound_transport_flow[f, rp, B_flow]
5 )

```

## Using DataFrames to linearise indices

```

1 df = filter(row -> (row.from, row.to) ∈ Ft, df_flows)
2 model[:max_transport_flow_limit] = [
3     @constraint(
4         model,
5         duration(row.time_block, row.rp) * flow[row.index] ≤ upper_bound_transport_flow[row.index],
6         base_name = "max_transport_flow_limit[(row.from),(row.to),(row.rp),(row.time_block)]"
7     ) for row in eachrow(df)
8 ]

```

```
julia> energy_problem.dataframes[:flows]
```

```
648x9 DataFrame
```

Row	from	to	rep_period	timesteps_block	efficiency	index	flow
	Symbol	Symbol	Int64	UnitRange...	Float64	Int64	GenericV...
1	ocgt	demand	1	1:1	0.0	1	flow[(ocgt, demand), 1, 1:1]
2	ocgt	demand	1	2:2	0.0	2	flow[(ocgt, demand), 1, 2:2]
3	ocgt	demand	1	3:3	0.0	3	flow[(ocgt, demand), 1, 3:3]

# Some Final Thoughts

## Good Things

- Speed & efficiency!
- Straight-forward syntax
- Great user community support!
- Others: DuckDB, Graphs, Clustering...
- Friendly to both researchers and software engineers

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## Room for Improvement

- Skill required for best speed/memory
- Add more tips for speed/efficiency improvement
- New Extensions?
  - Gather Update Solve Scatter (GUSS)
  - NearOptimalAlternatives.jl\* for modelling to generate alternatives – MGA

# Check out our GitHub!





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