

Addressing the challenge of modelling energy storage in a whole energy system

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Future energy targets are driving the deployment of renewable technologies



London Array – the world's largest offshore wind farm



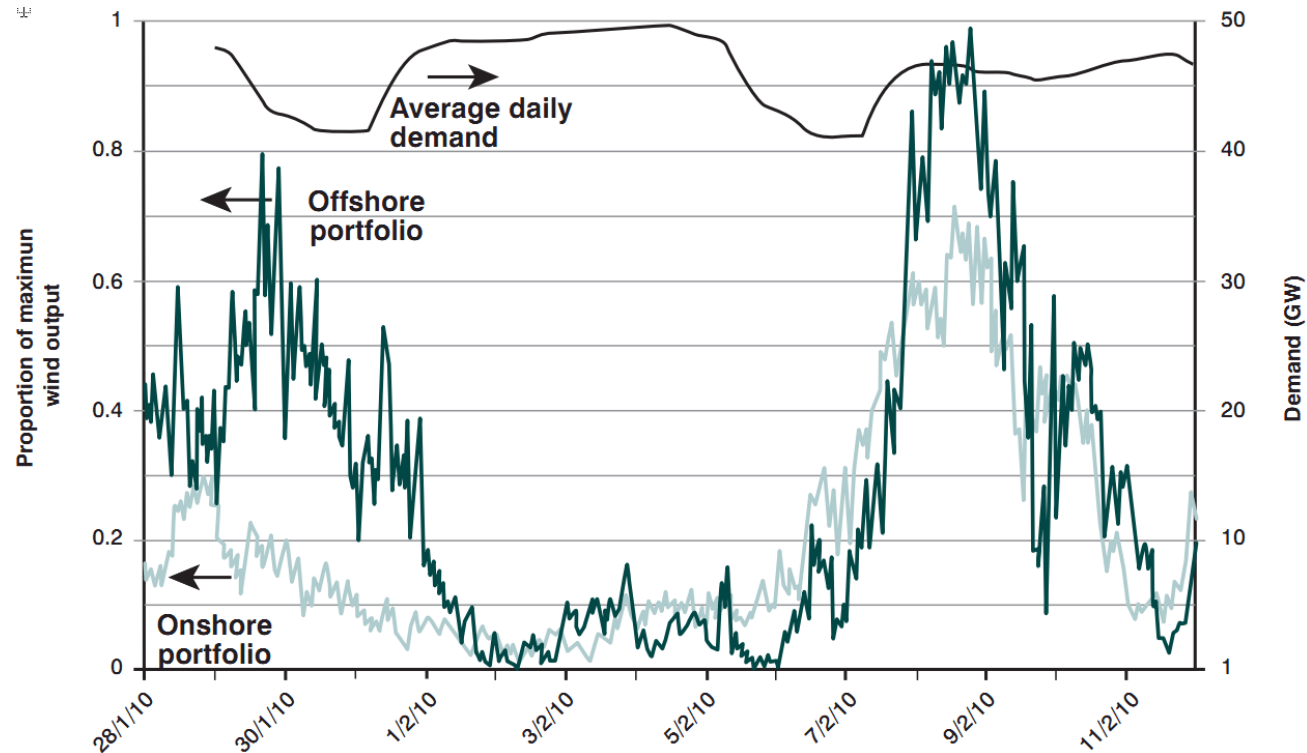
A solar farm



Aquamarine Power's Oyster Wave Energy

The intermittency challenge

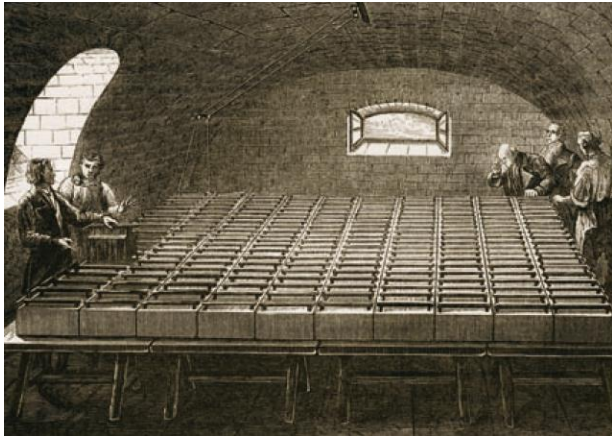
Source: The future role for energy storage in the UK, Energy Research Partnership, 2011



- Renewable energy is generated when it's not needed
- Large dip in generation during high demand
- Peaking generators, e.g. gas turbines, used to balance supply and demand are expensive and produce GHG emissions

Energy storage solution

Photo credit: The future role for energy storage in the UK, Energy Research Partnership, 2011



Royal Institution Battery 1807

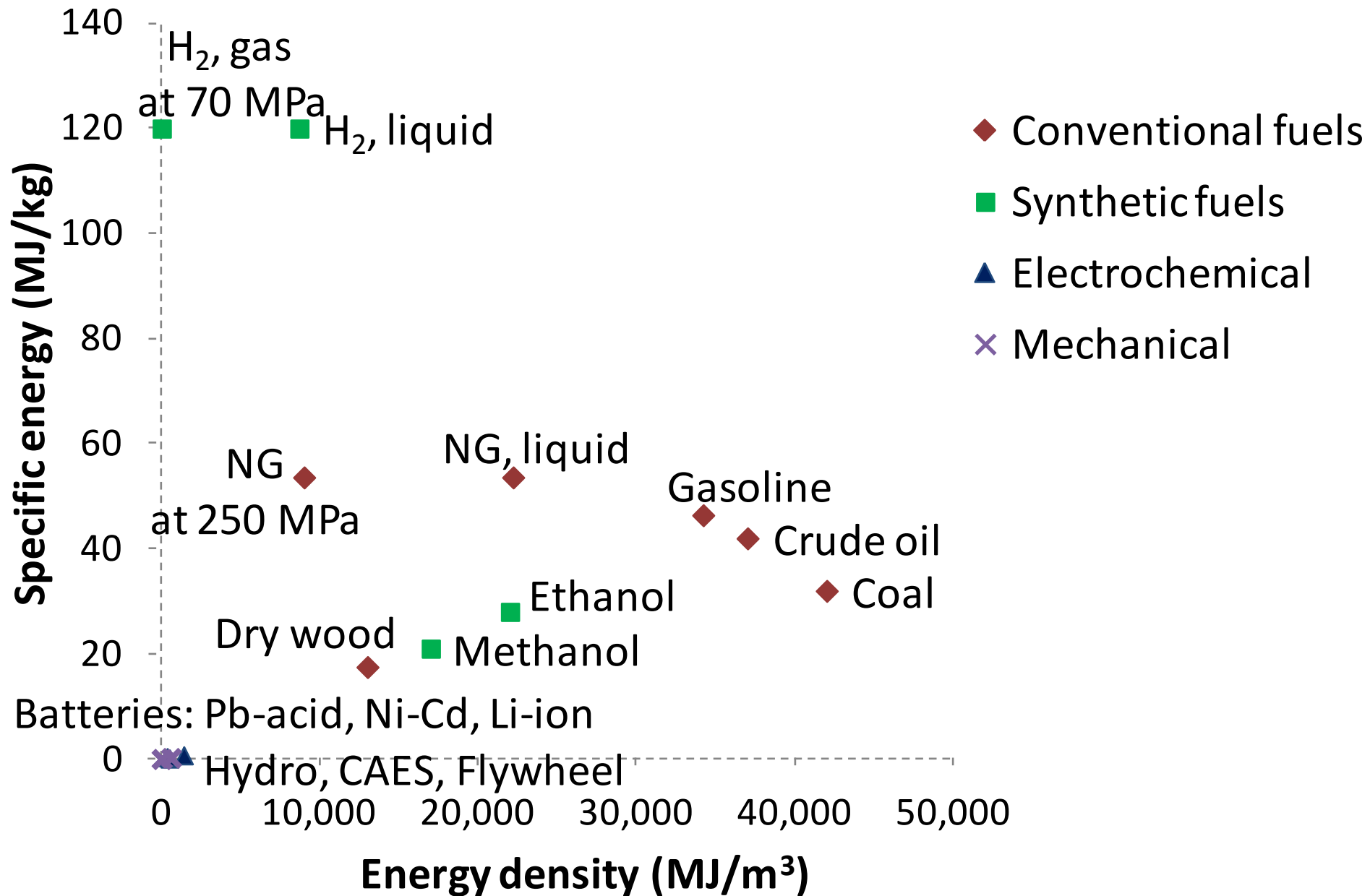


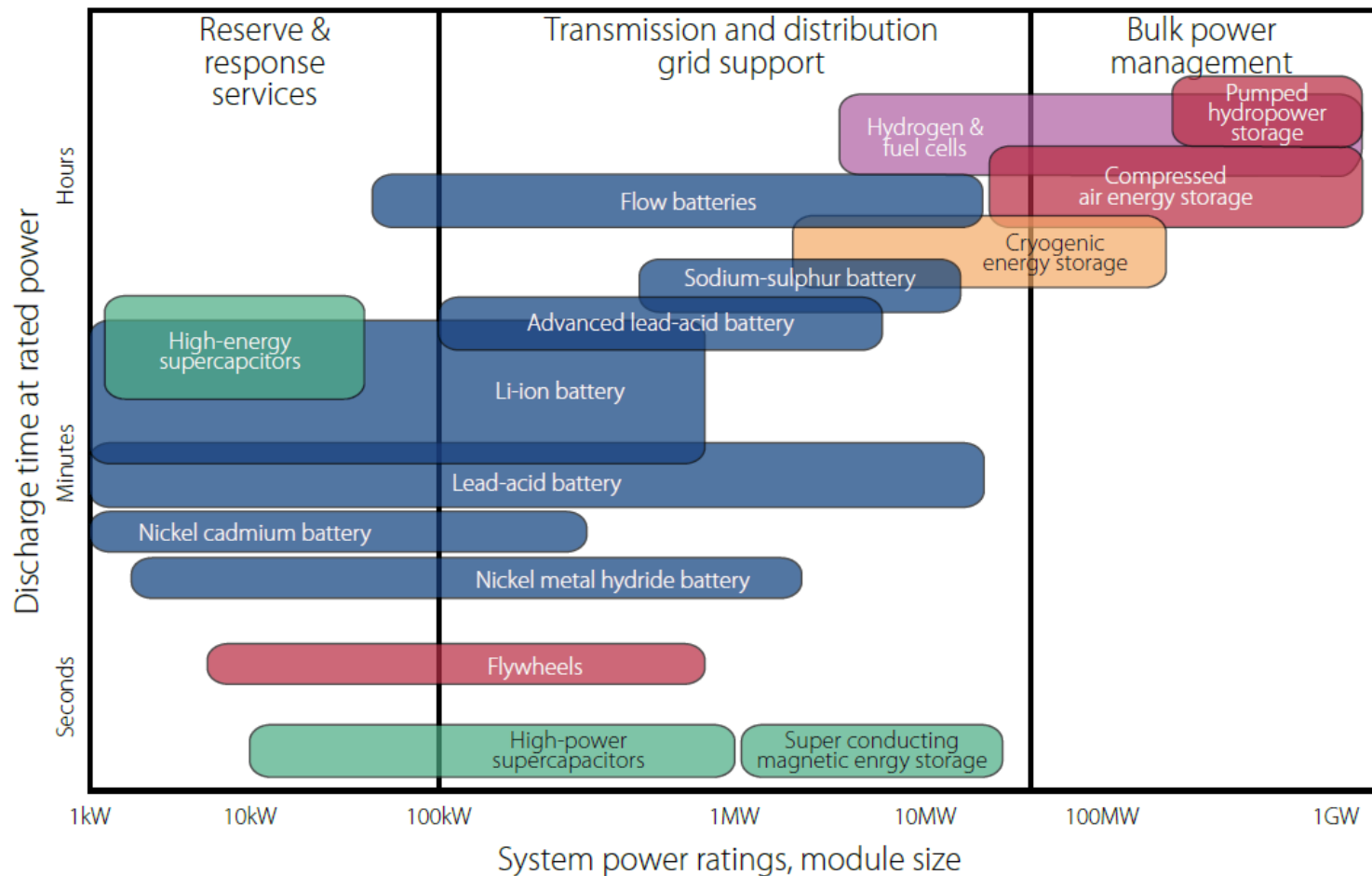
Llyn Stwlan reservoir



Hot water storage tank in the basement of a smart house

- Enables “wrong-time” energy generation from intermittent renewables
- Reduces need for peaking generators
- Improves energy use efficiency





Key: Types of storage Hydrogen-related Mechanical Electrochemical Electrical Thermal

Source: Williams, R. Is liquid air the missing link in energy storage. Focus April 2013.

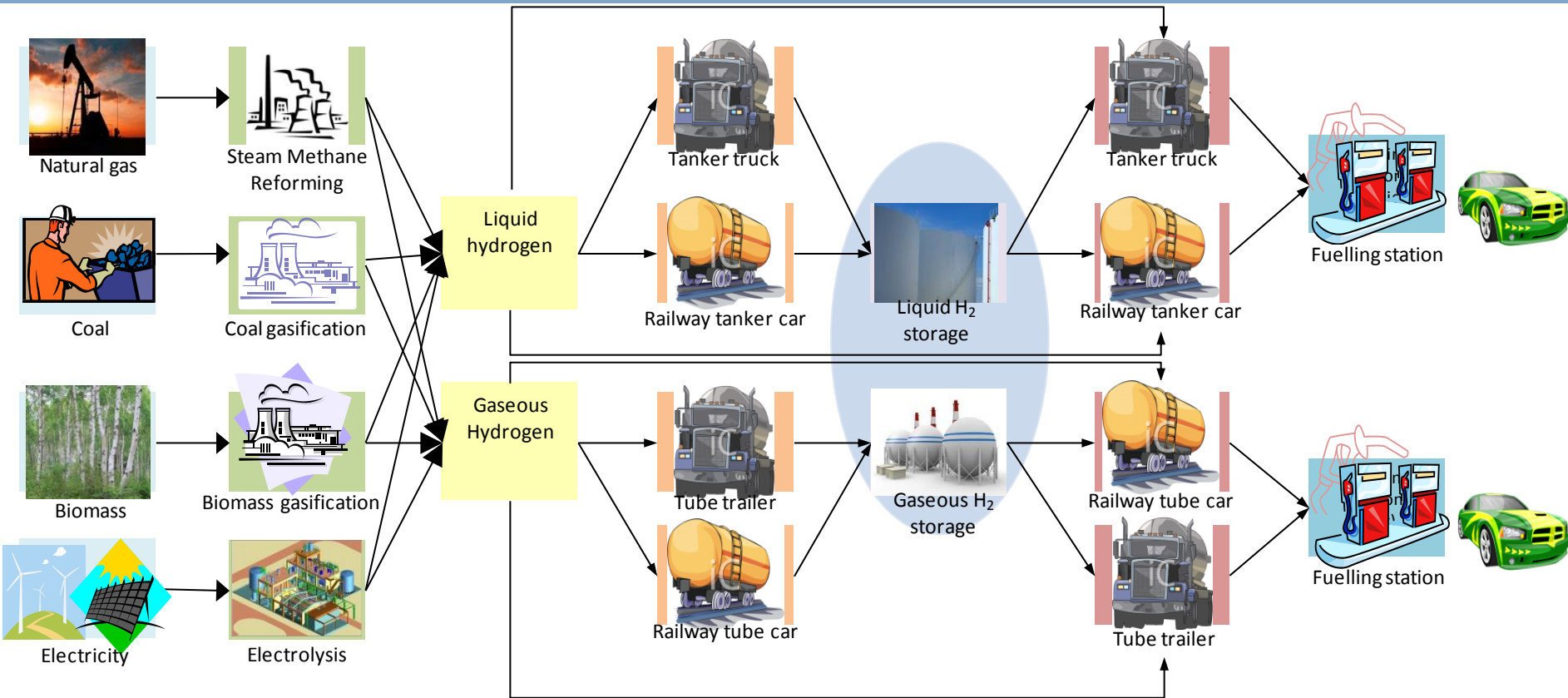
Modelling challenge: the dynamics of storage technologies occur over short time scales (<hourly), very different from the time interval in energy system planning models (>yearly)

Tractability is an issue!

Challenges

- We need a dynamic energy system model with a very wide range of time scales
 - Planning: years or decades
 - Seasonal: variations in demands and availability
 - Hourly (or shorter):
 - Dynamics of storage technologies
 - Variations in demands, intermittency of renewable resources
- Still need to model spatial aspects
 - Demands and availability depend on location
 - Determine location/size of technologies and storage facilities
 - Requires integer variables
 - Transport of resources (centralised vs. distributed)
- Very large scale model

Hydrogen Supply Chain (HSC) model

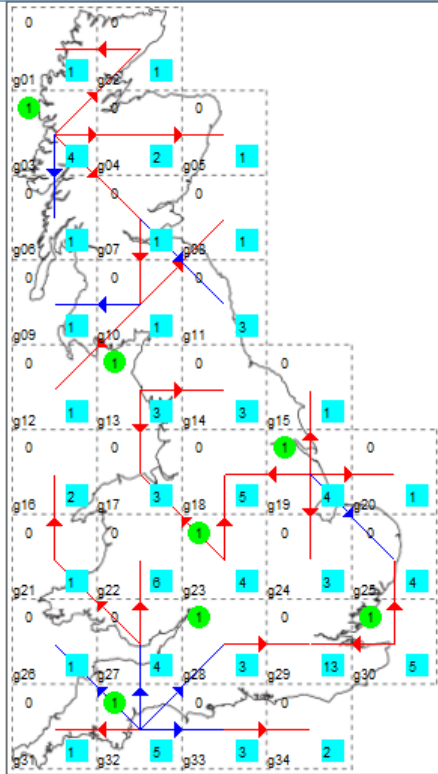


Spatial element: Great Britain represented by 34 108×108 km² square cells

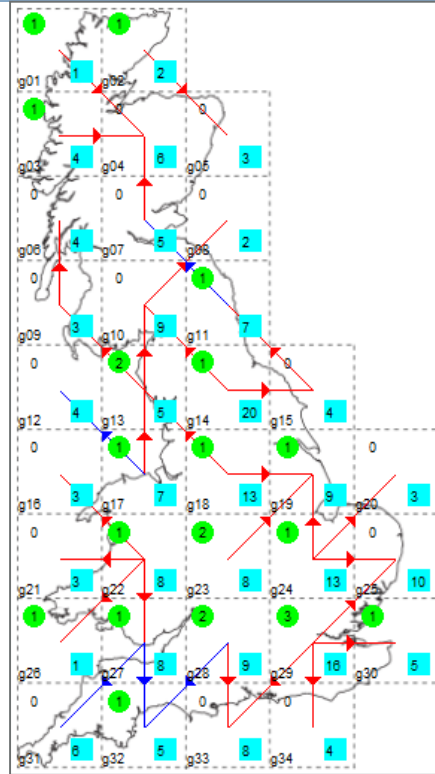
Temporal element: 2015-2044 divided into 5 6-year periods

Last upgrade: made it a dynamic model with time intervals of 4 seasons in a year and 4 6-hr periods in a day

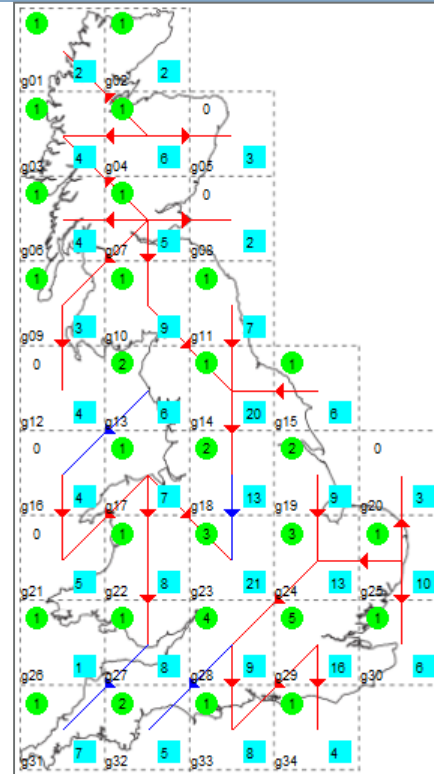
Example case study



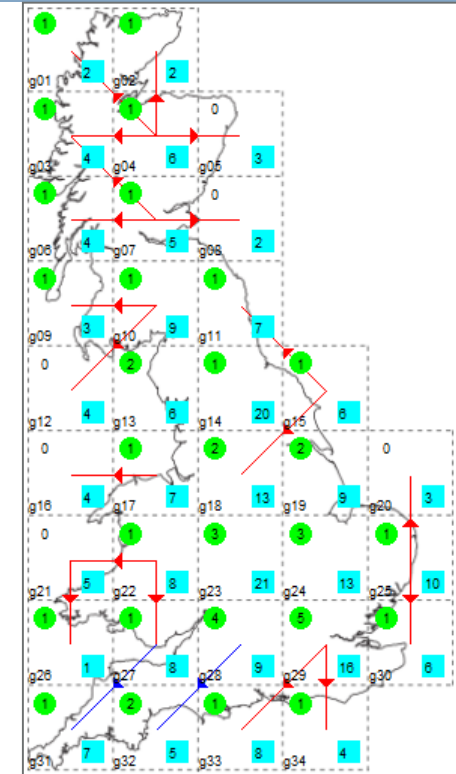
Summer 2015-2020



Summer 2027-2032



Summer 2039-2044



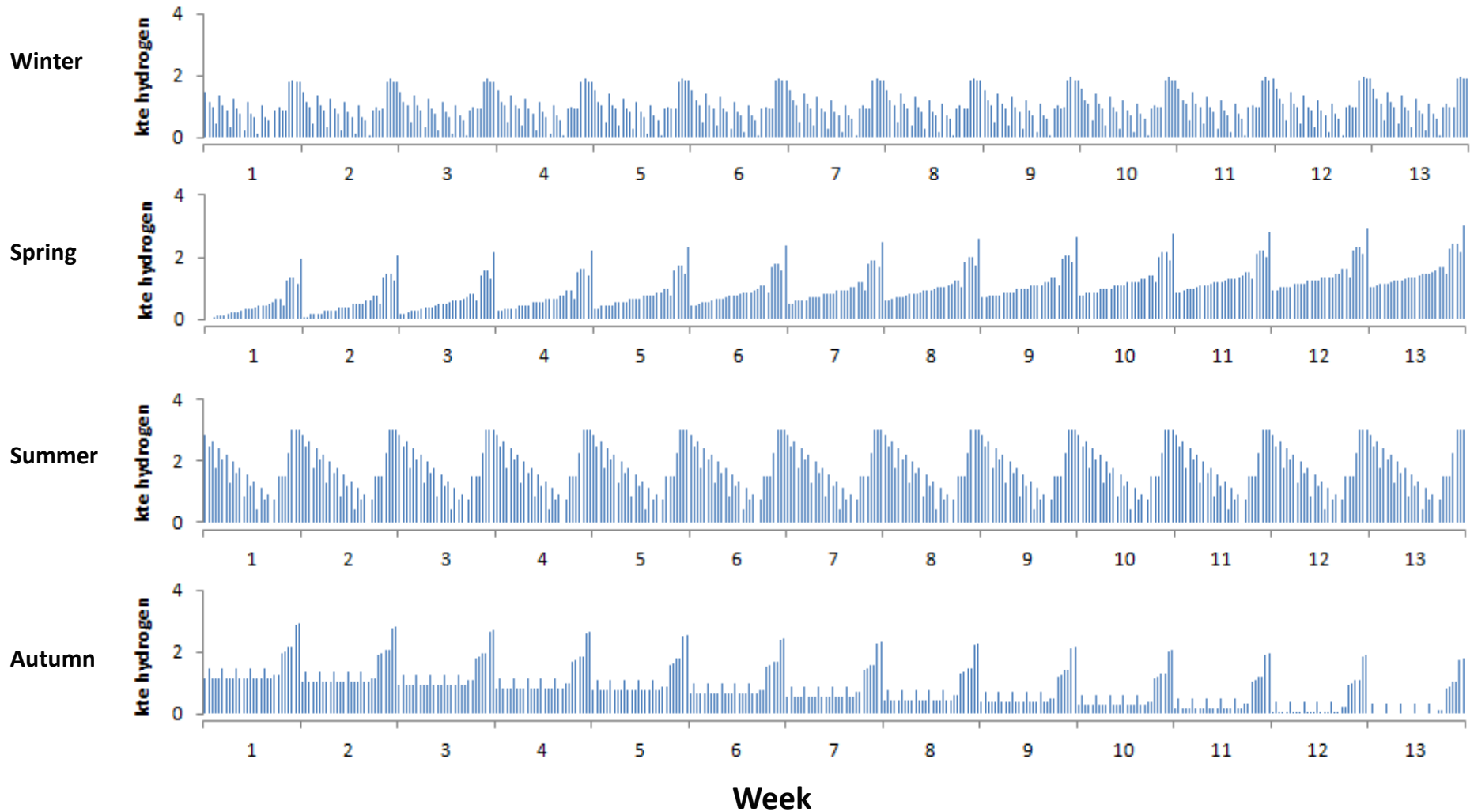
Winter 2039-2044

→ Rail transport ● No. of production plants ■ No. of storage facilities
→ Truck transport

No. of variables > 1M, No. of constraints > 0.5M, Integers = 15k
Took ~3 days to solve full MIP!

Inventory profile for a whole year

London and the South East (cell 29) in 2039-2044



Limitations of the HSC model

- Multi-echelon model
 - Pathways inflexible
 - Distribution within cells too complex
 - Too many binary variables
 - Big M formulation
- Too large to be extended
 - Adding a pipeline transport mode resulted in intractable problems
 - Difficult to add new technologies and resources
 - Still not enough time intervals

Back to basics:

A very simple MILP model with storage

Resource balance: $I_{rch} + U_{rch} + \sum_p \alpha_{rp} P_{pch} + \sum_{c'} (Q_{rc'ch} - Q_{rcc'h}) - D_{rch} = S_{rch} - S_{rc,h-1} \quad \forall r, c, h$

Production capacity constraint: $P_{pch} \leq NP_{pc} p_p^{\max} \quad \forall p, c, h$

Resource availability constraint: $U_{rch} \leq u_{rch}^{\max} \quad \forall r, c, h$

Storage capacity constraint: $S_{rch} \leq s_{rc}^{\max} \quad \forall r, c, h$

Objective function definition

A simple model is intractable for the time horizon needed for a planning model!

$|r| = 3$ $|p| = 4$ $|c| = 14$ No. of integer variables = 56

h - contiguous hourly interval

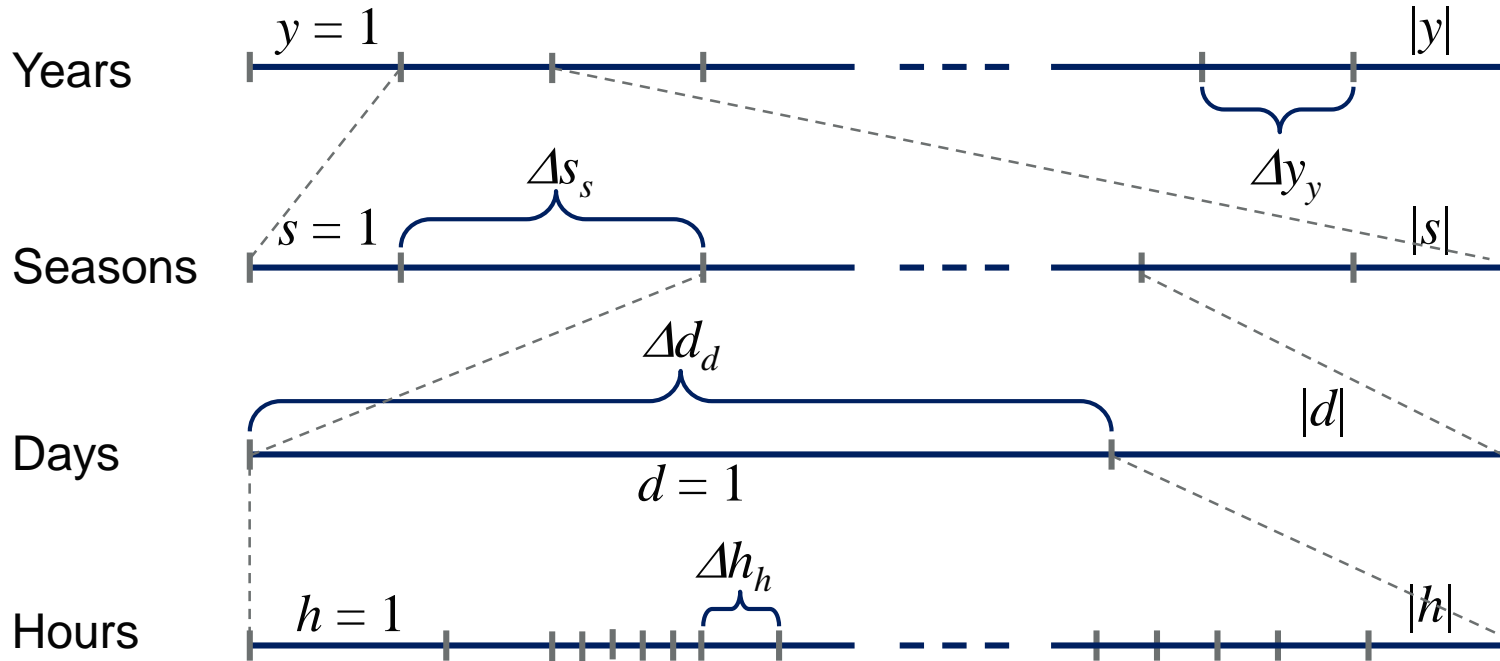
$ h $	No. of variables	No. of constraints	Solution time (s)
24	6,139	3,703	7
168	42,427	25,879	744
720	181,531	110,887	22,248
2,160	544,411	332,647	>155,520
8,760	2,207,611	1,349,047	??!



STeMES

- **Spatio Temporal Model for Energy Systems**
- RTN representation of energy pathways
 - same framework as the BVCM and TURN model in SynCity toolkit
- MILP formulation
- Efficient representation of time
- Detailed storage formulation
- Transport losses modelled in detail

Hierarchical non-uniform time discretisation

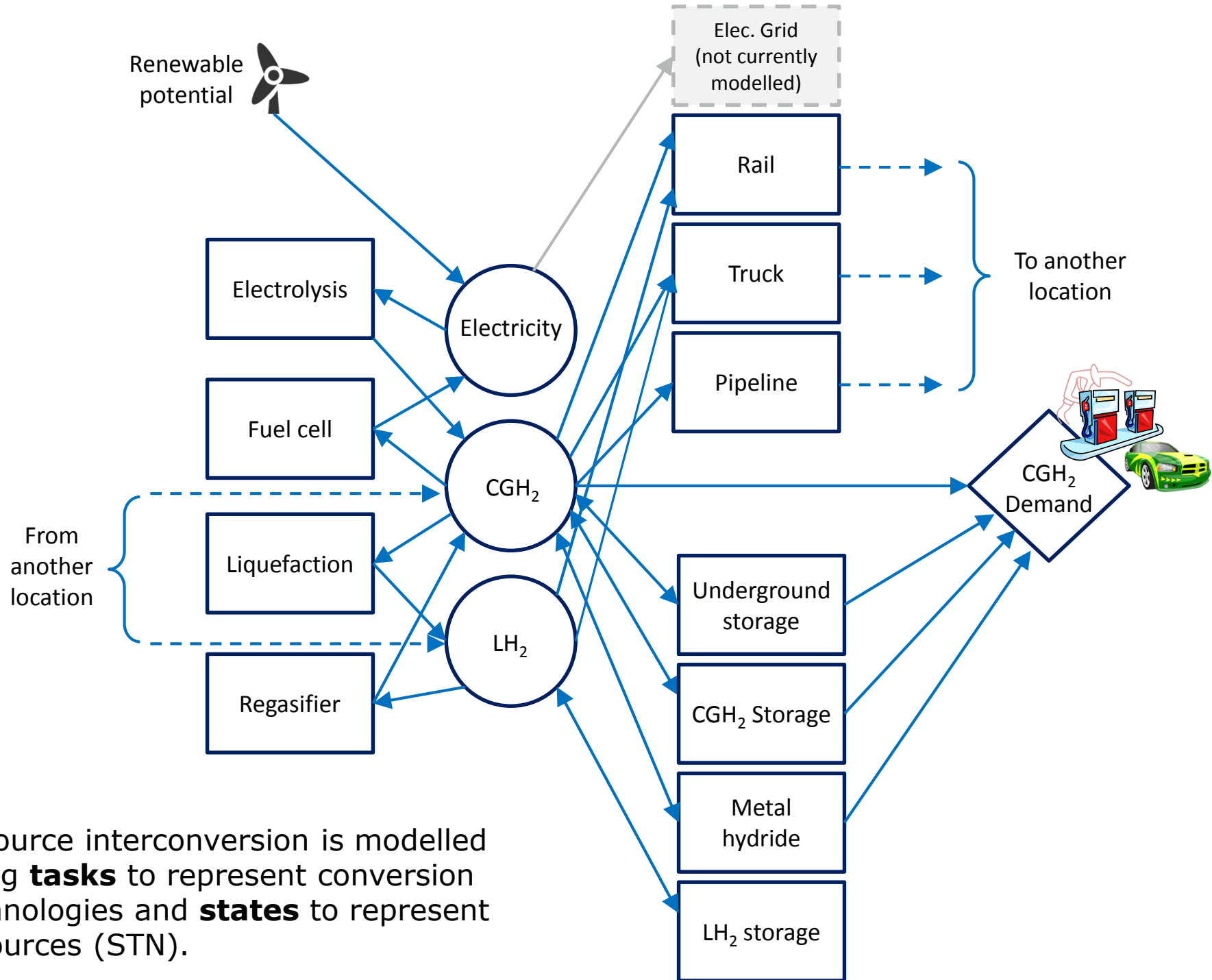


Total number of time intervals $T = |y| \times |s| \times |d| \times |h|$

e.g. for one year, $T = 1 \times 4 \times 2 \times 24 = 192 \ll 8760$

Without storage – very easy!

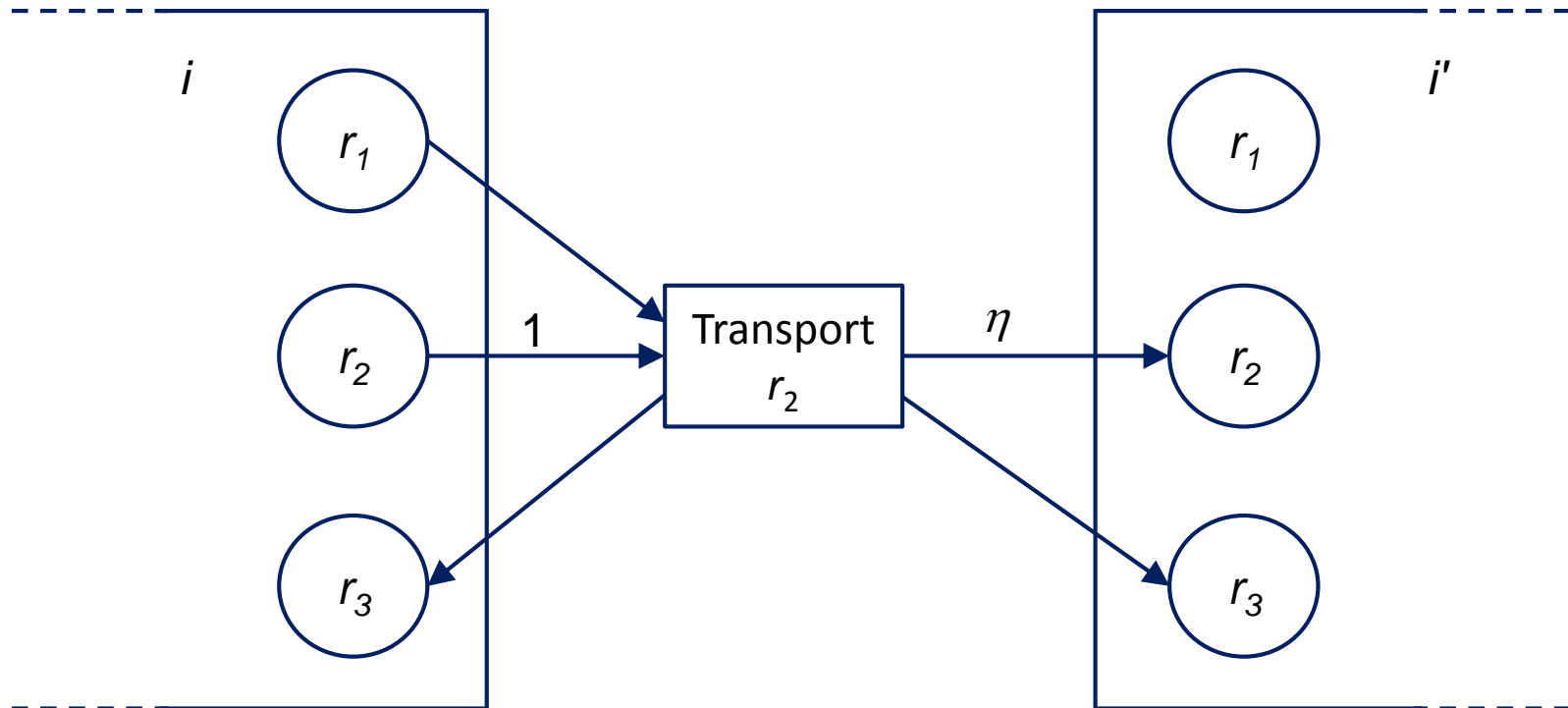
With storage – extra variables for initial inventories; extra constraints to link inventories within and between time levels



Resource interconversion is modelled using **tasks** to represent conversion technologies and **states** to represent resources (STN).

Transport

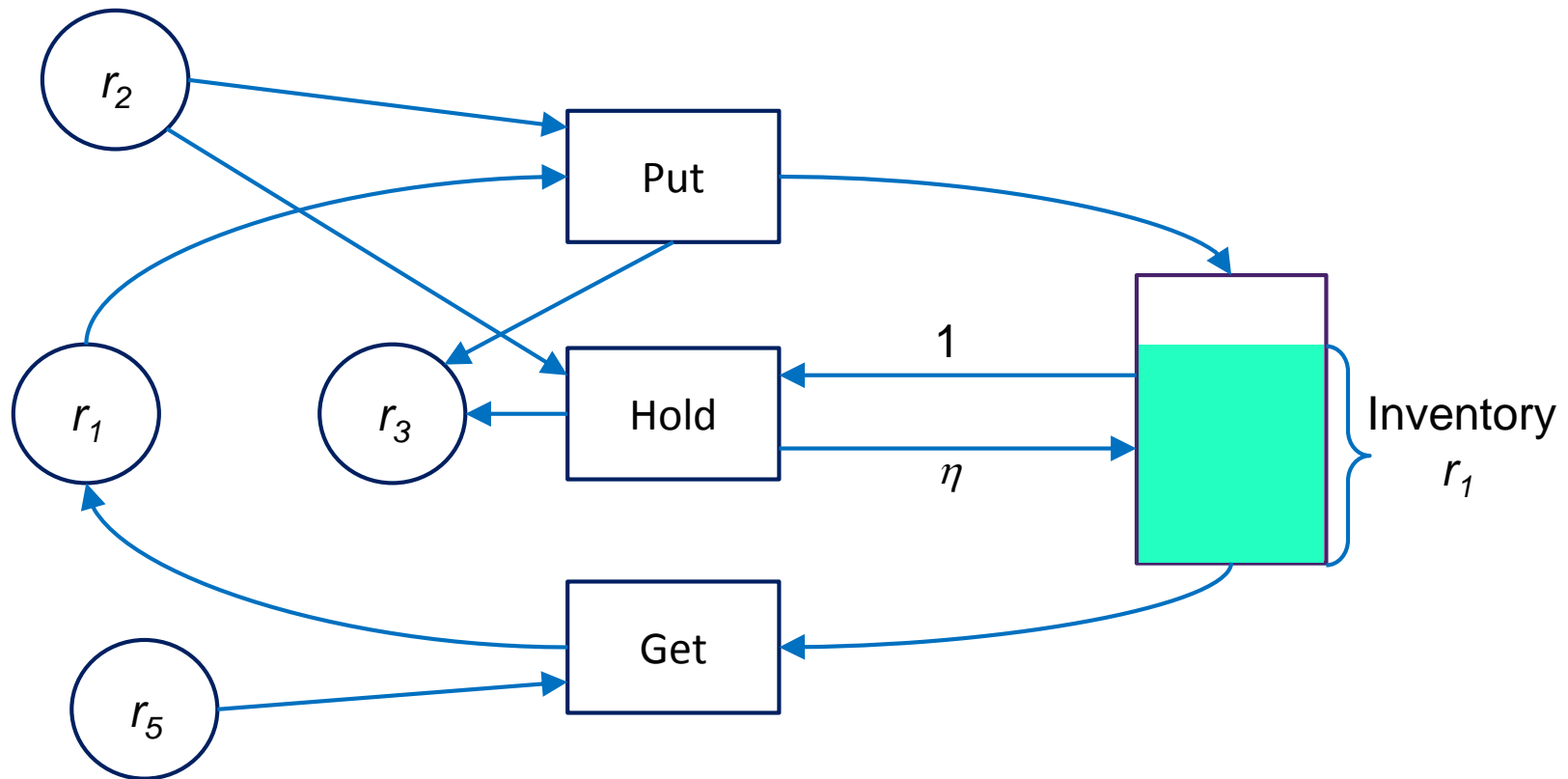
Transport task – used to model connections between cells



Resource r_2 is transported from cell i to cell i' , which requires r_1 from cell i and results in waste r_3 being generated in both cells

Storage

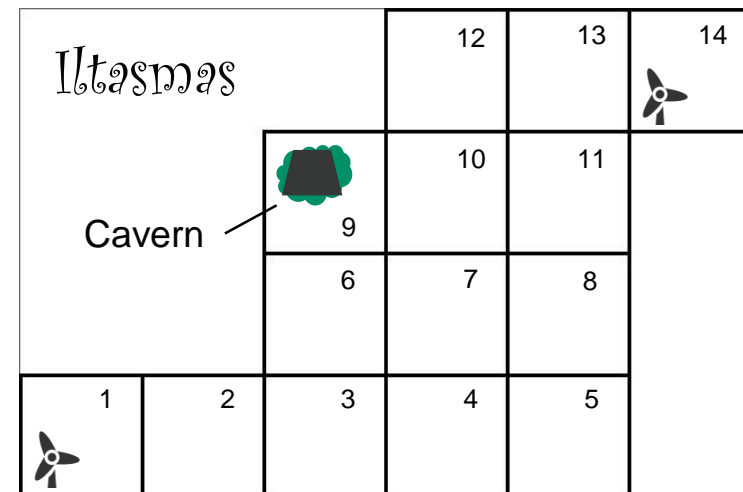
An example set of **storage tasks** to store resource r_1 .



The “put” task transfers r_1 from the cell to the store, requiring some r_2 and producing some wastes r_3 (e.g. CO_2). The “hold” task maintains r_1 in storage, which also requires some r_2 but at less than 100% efficiency, the losses being converted to r_3 . Finally, the “get” task retrieves r_1 from storage and delivers it to the cell, requiring some r_5 .

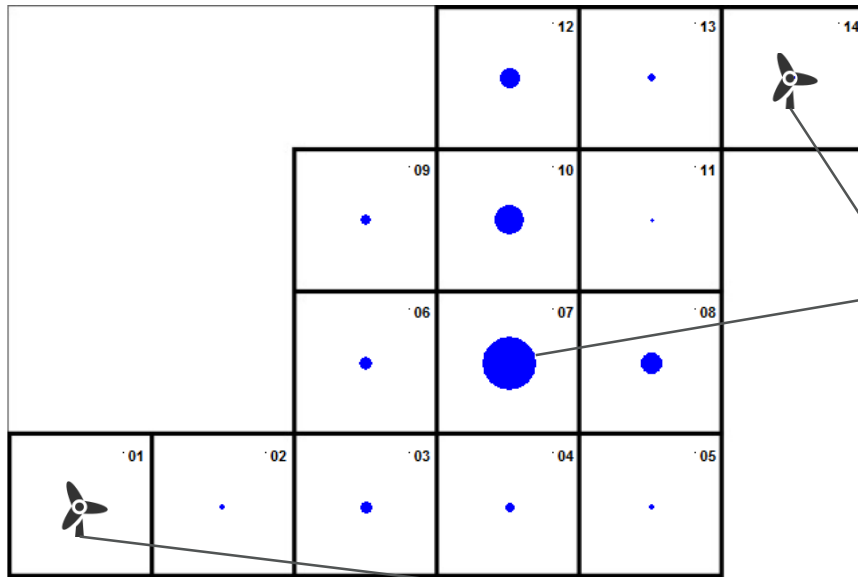
STeMES prototype

- Developed and tested for a hypothetical island of 14 50×50km cells
- Wind generation installed at two locations
- Choice of storage technologies
 - Salt cavern available for use as hydrogen storage facility
 - Other H₂ storage technologies: gaseous (tank), liquid, metal hydride
- Target: transport demand to be met by hydrogen (CGH₂)
- Objective: Minimum cost
- Decisions
 - Location and size/number of hydrogen production and storage facilities
 - Operation of production facilities
 - Operation of storage facilities
 - when to charge and discharge
 - Transportation of hydrogen

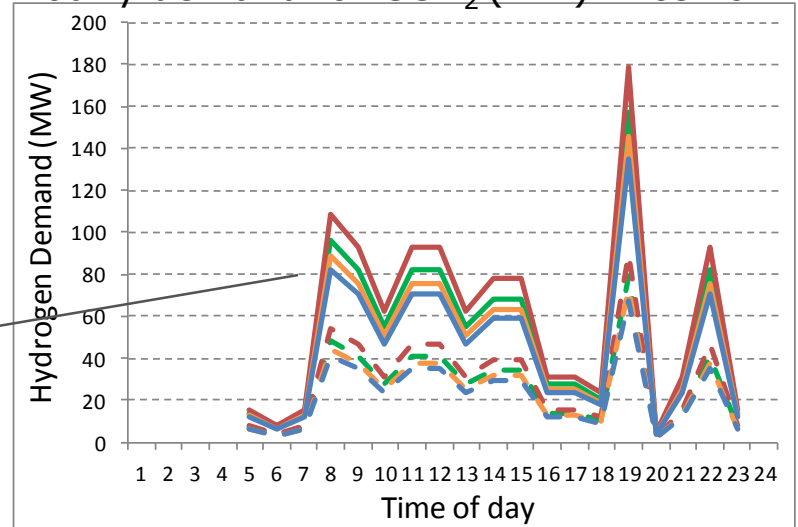


Spatio-temporal input data

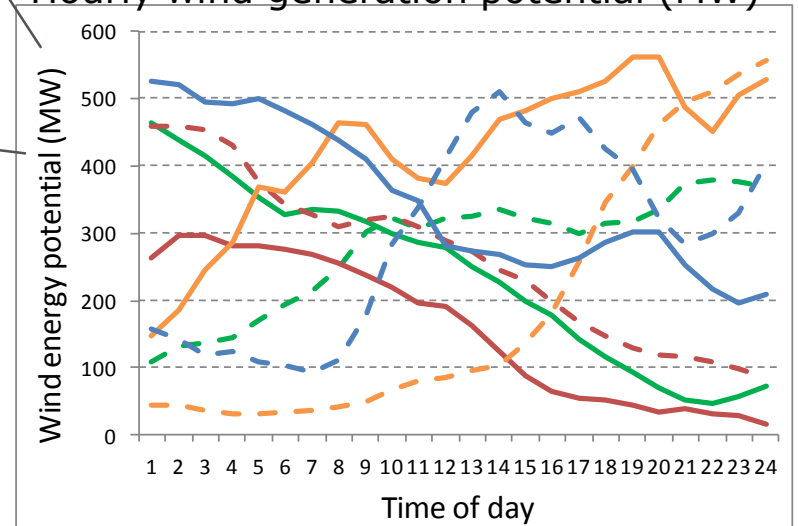
Ave. daily demand for CGH₂ (MW)



Hourly demand for CGH₂ (MW) in cell 07



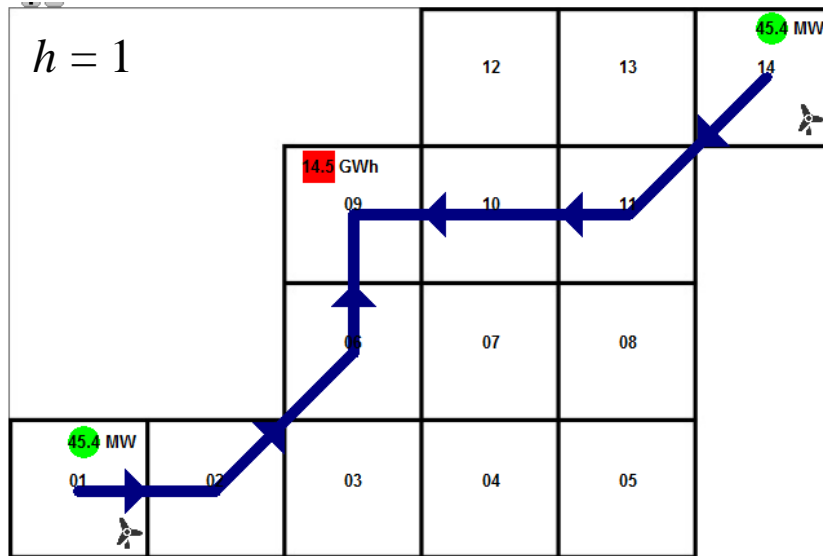
Hourly wind generation potential (MW)



- Spring WD - - Spring WE
- Summer WD - - Summer WE
- Autumn WD - - Autumn WE
- Winter WD - - Winter WE

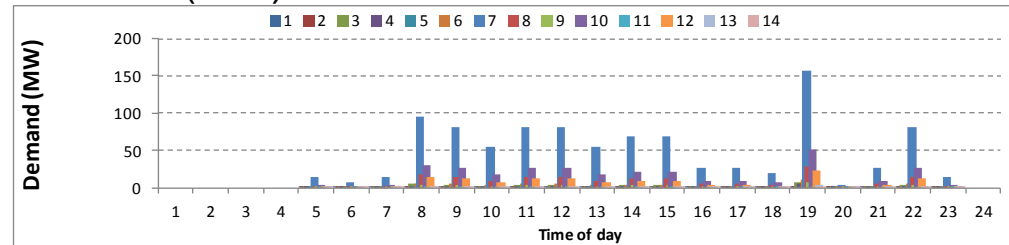
Results $|y|=1, |s|=4, |d|=2, |h|=24$

Snapshot of the network during weekday ($d=1$) in spring ($s=1$)

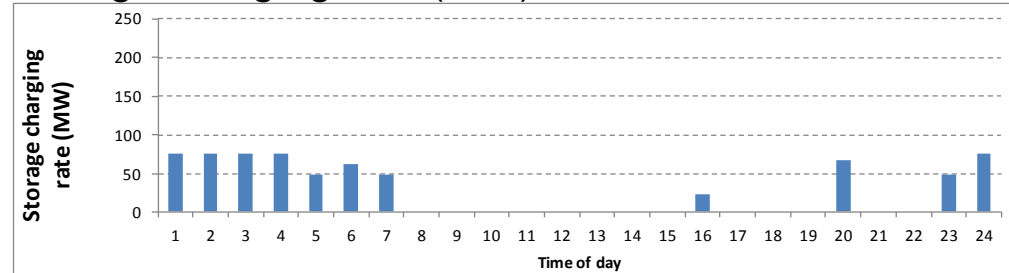


- Hourly transport of CGH2 by pipeline
- Installed electrolyser capacity (3 small units)
- Installed underground storage capacity

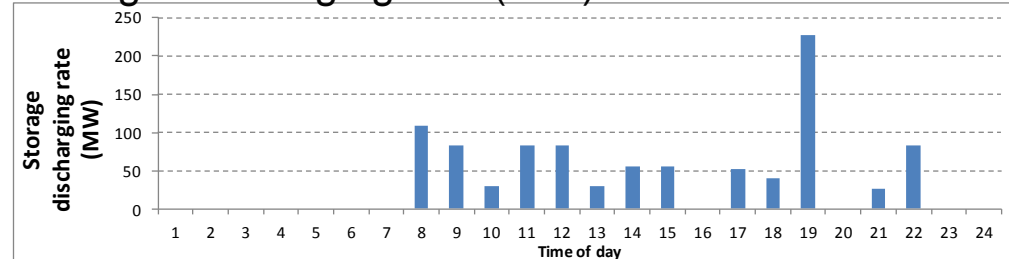
Demand (MW)



Storage charging rate (MW)

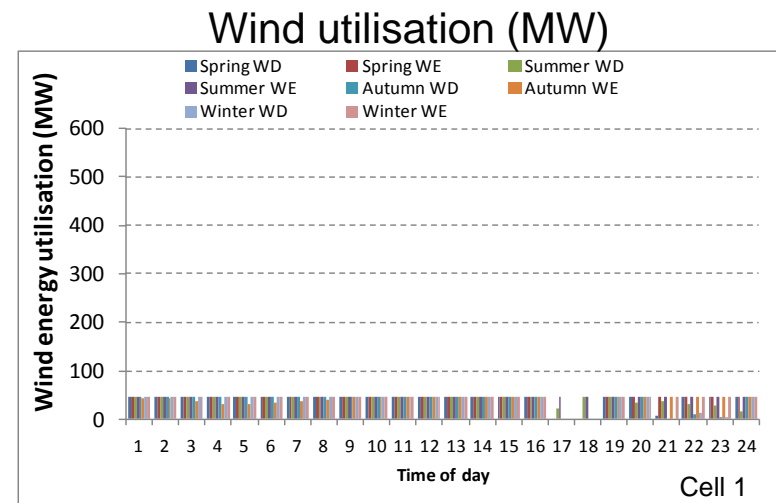
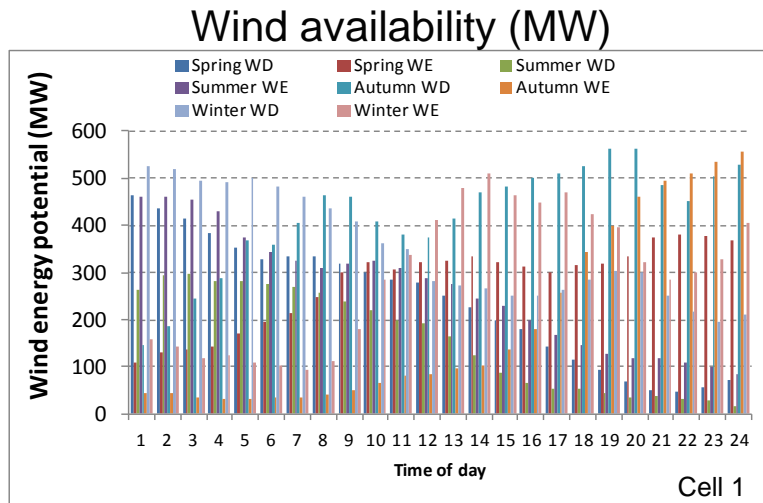


Storage discharging rate (MW)

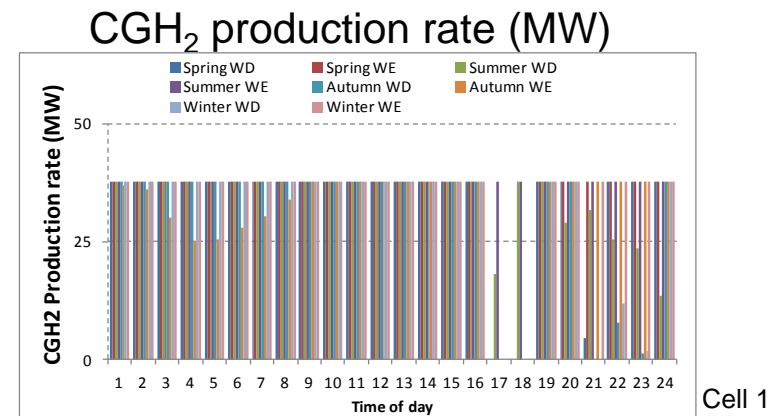


Resource Utilisation

- Without storage, the scenario is infeasible
- BUT with storage, only a fraction of the available wind energy is needed!

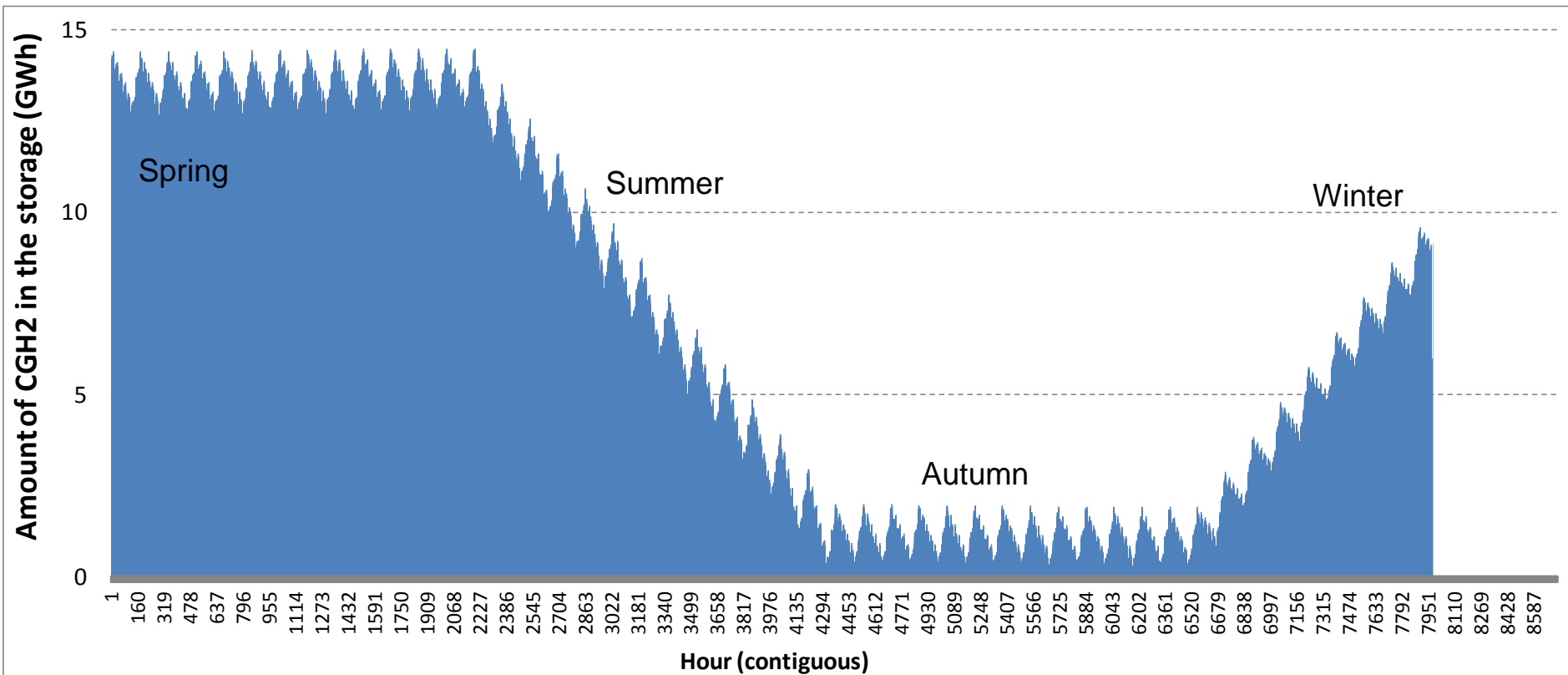


- The rate of operation of electrolyser is effectively constant



Results $|y| = 1$, $|s| = 4$, $|d| = 2$, $|h| = 24$

Hourly inventory of CGH₂ in the storage for a whole year

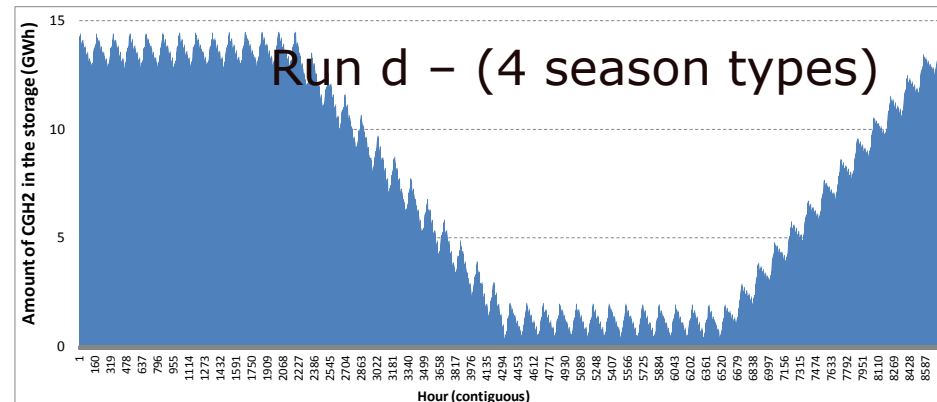
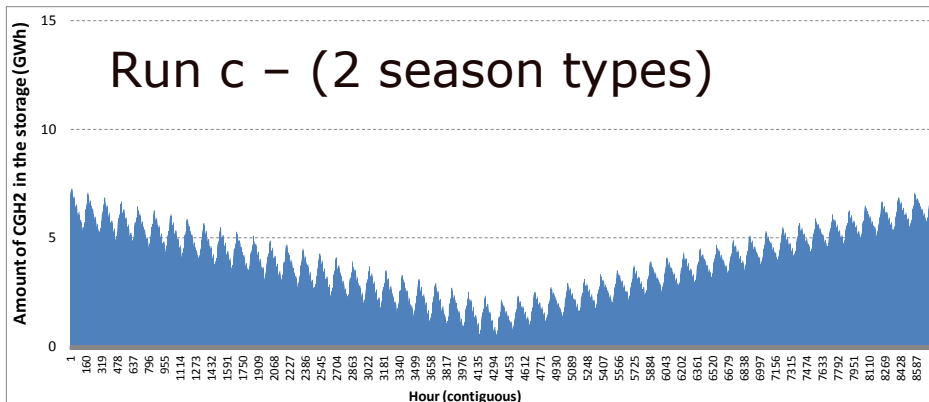


Benchmarking

No. of integers = 336, relative tolerance = 0.1%

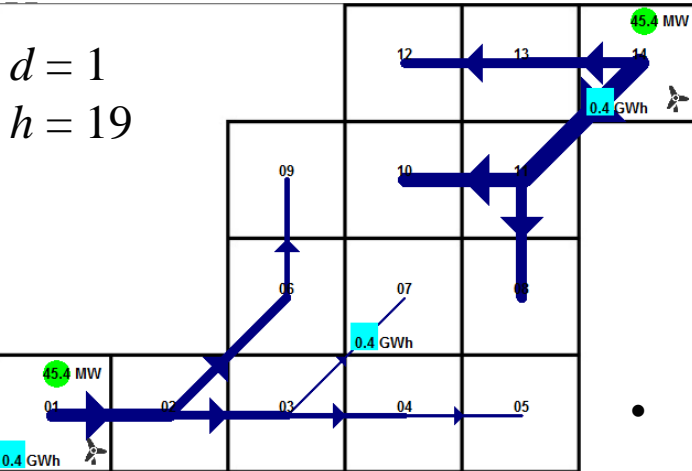
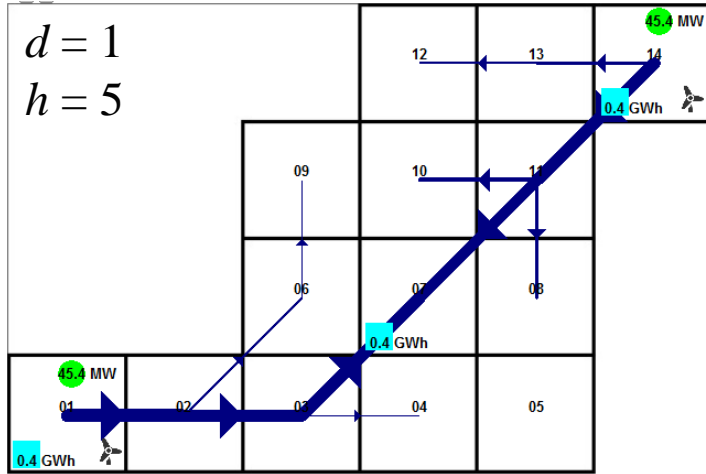
Run ID	$ y $	$ s $	$ d $	$ h $	No. of variables	No. of constraints	Solution time (s)
a	1	1	1	24	32,823	98,631	7
b	1	1	2	24	64,959	196,911	272
c	1	2	2	24	129,399	393,639	2,543
d	1	4	2	24	258,279	787,095	69,480

- All runs determined 45.4 MW of electrolysis capacity installed in cells 1 and 14, H₂ transport by pipeline and underground storage.
- However, the runs with fewer time intervals underestimated the storage capacity.



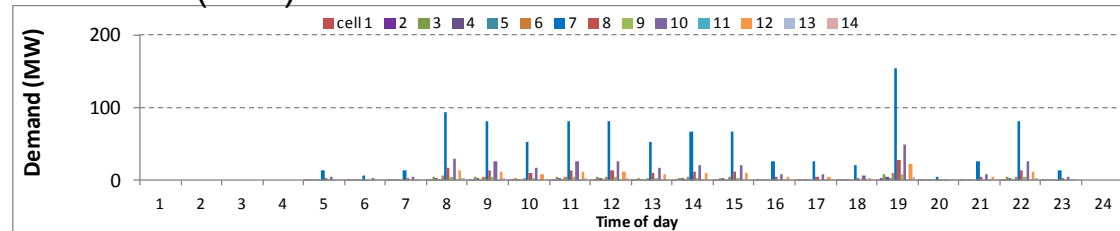
If underground storage is not an option

$|y|=1, |s|=1, |d|=2, |h|=24$

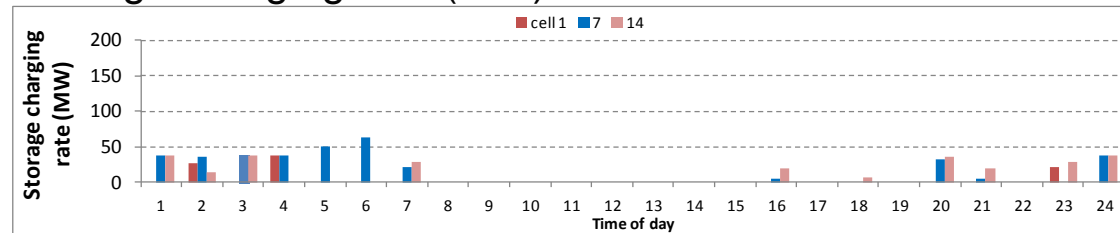


- ➔ Hourly transport of CGH₂ by pipeline
- Installed electrolyser capacity (3 small units)
- Installed CGH₂S capacity (1 small unit each)

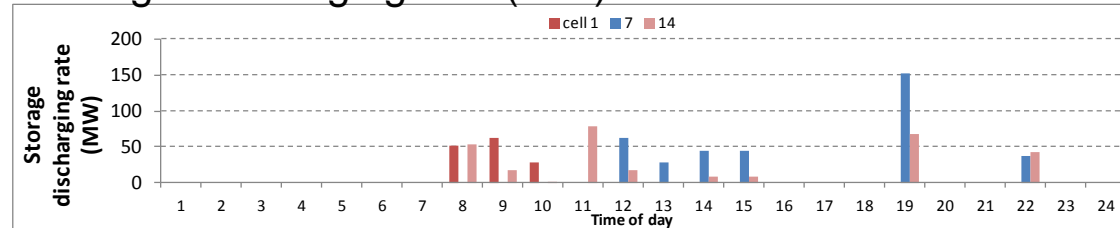
Demand (MW)



Storage charging rate (MW)



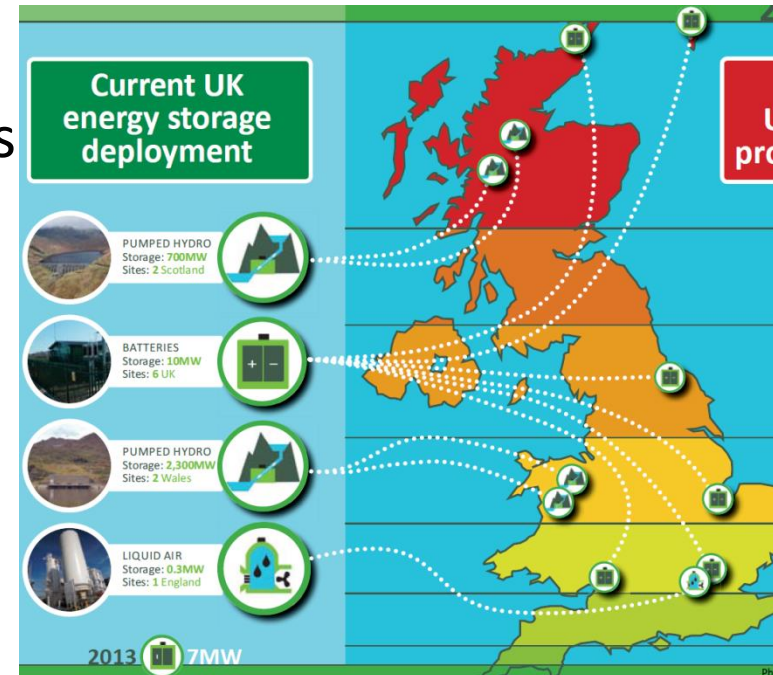
Storage discharging rate (MW)



- Network is more distributed with small compressed gaseous H₂ storage technologies installed in cells where generation are located and cell the with highest demand.

Next steps

- Real case studies (e.g. UK scenarios)
- Add more resources and technologies
- Exploit the full potential of the non-uniform hierarchical discretisation method
 - E.g. Use fewer *non-uniform* hourly intervals
- Additional decomposition methods
 - Benders decomposition did not work
 - Test in-house approaches



Source: The Electricity Storage Network. Development of electricity in the national interest. May 2014

Conclusions

- Storage is a key-enabling technology for meeting the energy demands using renewable resources
 - Without storage the example problem is infeasible
 - With storage, only a small fraction of available primary resource is used and the generation technology operates effectively at a constant rate
- To model storage accurately, hourly or shorter intervals are needed
 - In the example, four seasons are also needed
- Model tractability is a big challenge
 - Even the simplest model cannot handle a whole year at an hourly level
 - Hierarchical time decomposition allows a whole year (and longer planning horizon) to be considered by exploiting periodicity in the data