

# How does technology learn ?

Marta Victoria

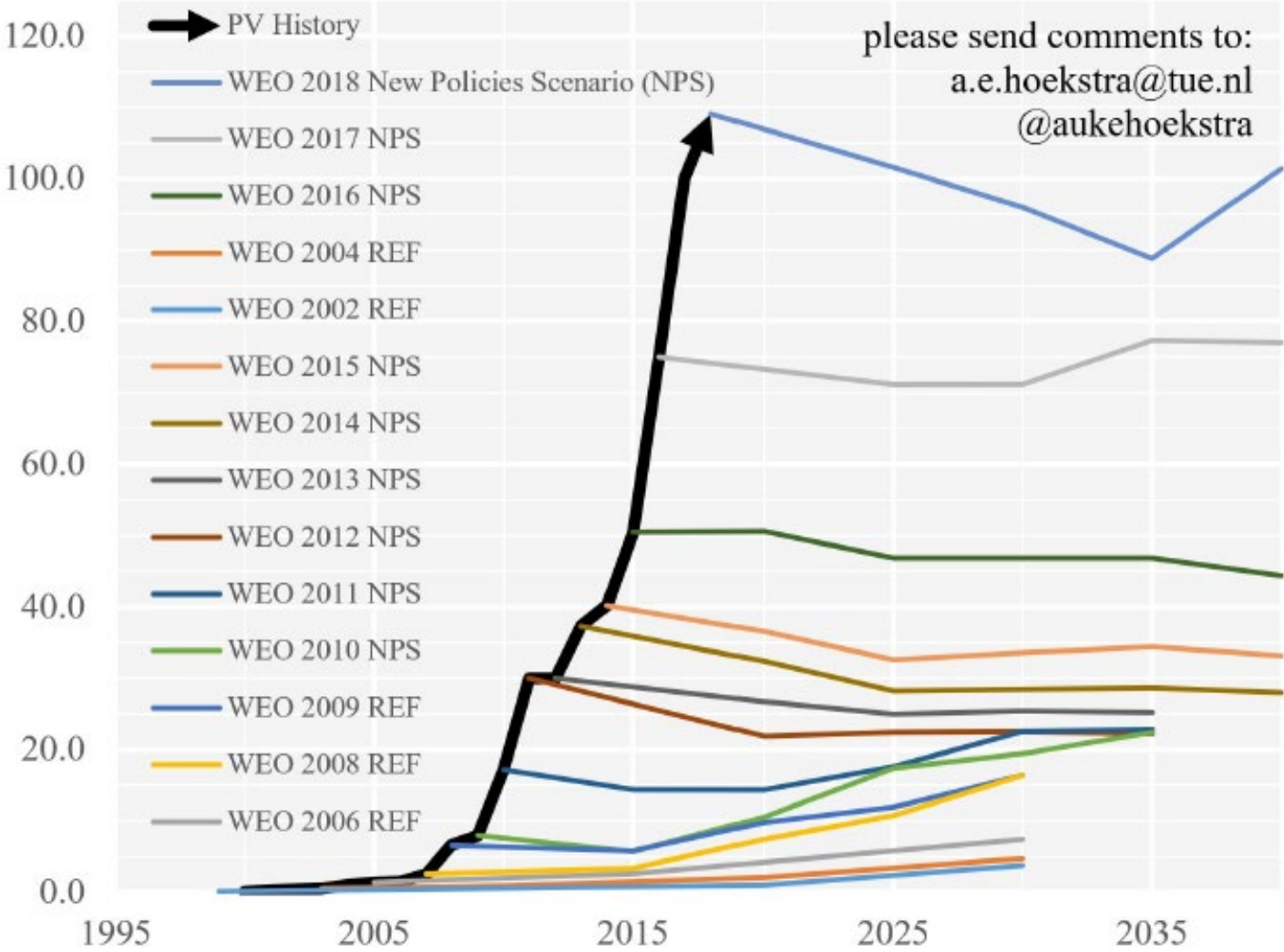
# Outline

- Technology learning: the example of solar photovoltaics
- How do we include learning in systems modelling?
- How could learning impact CO<sub>2</sub> capture and conversion?

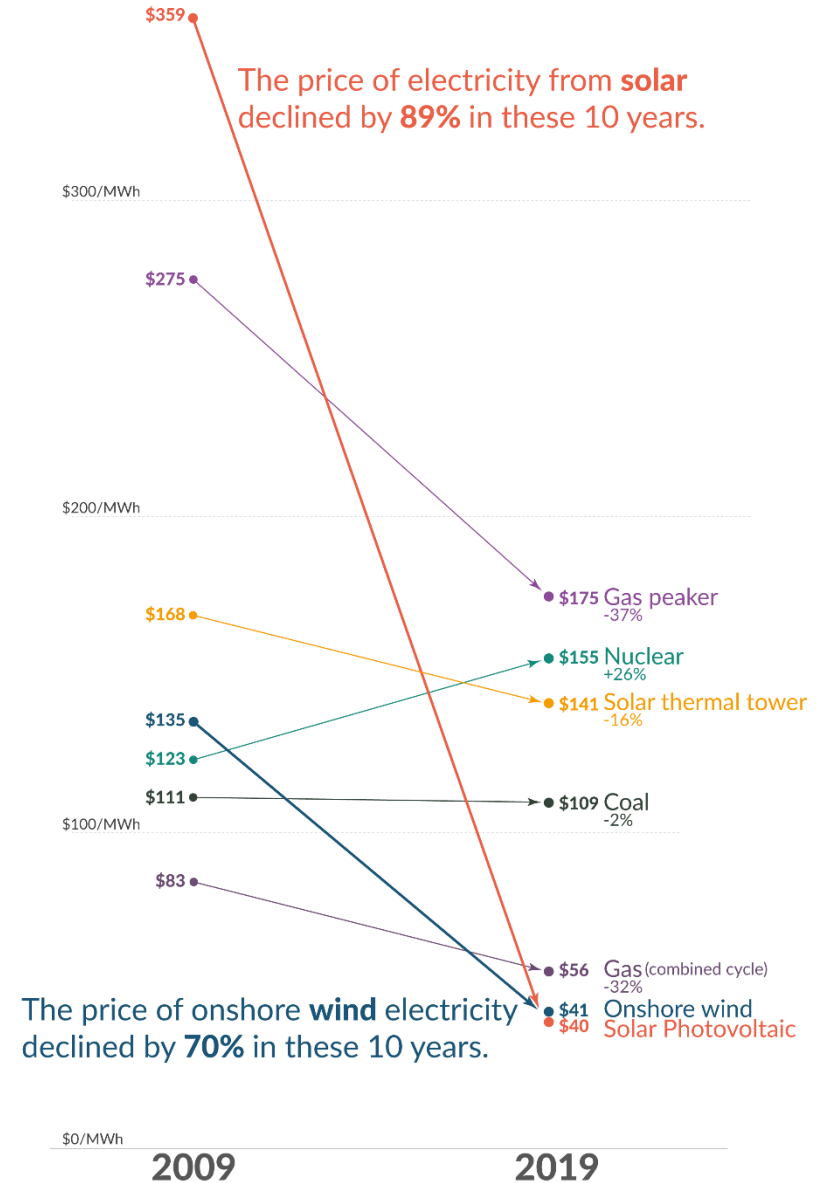
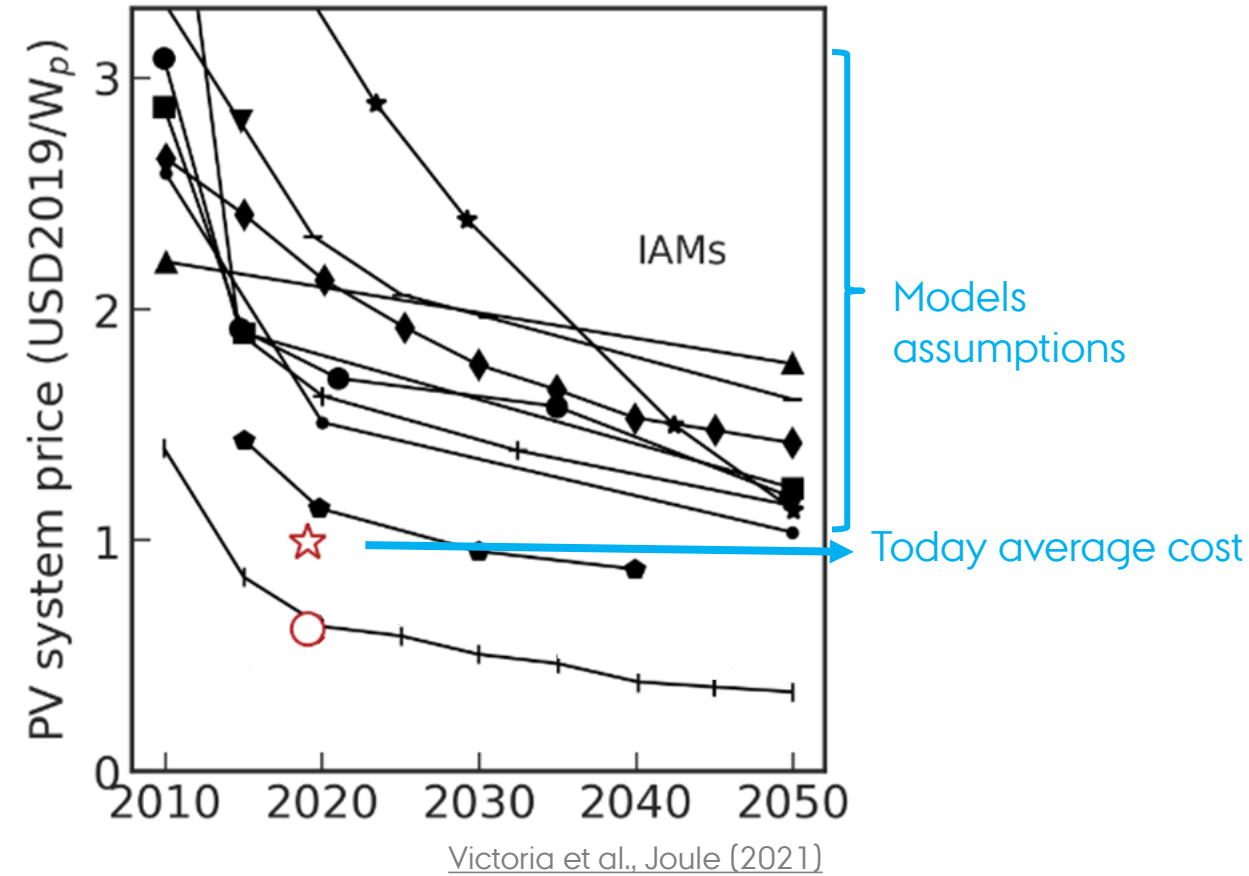
# Almost everybody underestimated the potential of solar photovoltaics (PV)

## Annual PV additions: historic data vs IEA WEO predictions

In GW of added capacity per year - source International Energy Agency - World Energy Outlook



# Many models assumed too high future cost for solar PV



# How did solar become so cheap so fast?

**Learning by doing** cost decreases as experience of production increases.

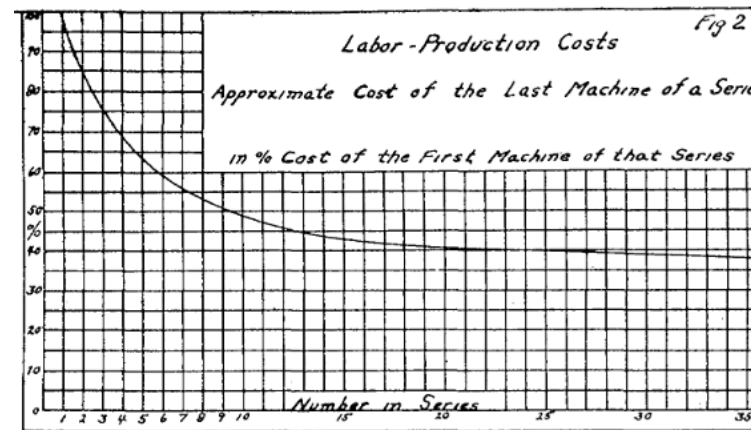
$$\frac{Cost_y}{Cost_{y_0}} = \left( \frac{Cumulative\ capacity_y}{Cumulative\ capacity_{y_0}} \right)^{-b}$$

Learning rate = relative price reduction every time cumulative capacity doubles

$$LR = 1 - 2^{-b}$$

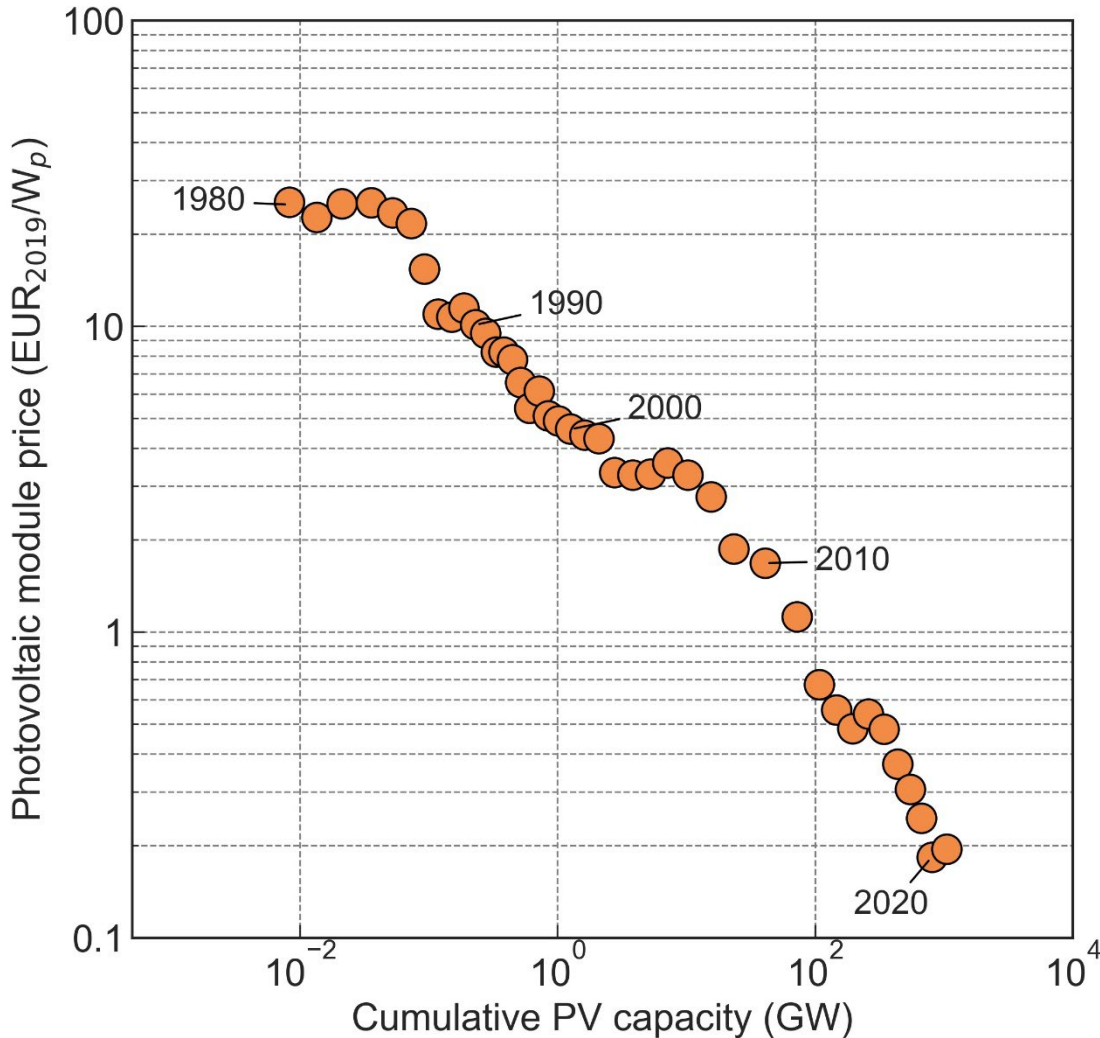
PV module show LR=23% since 1980

First-time observed in airplanes



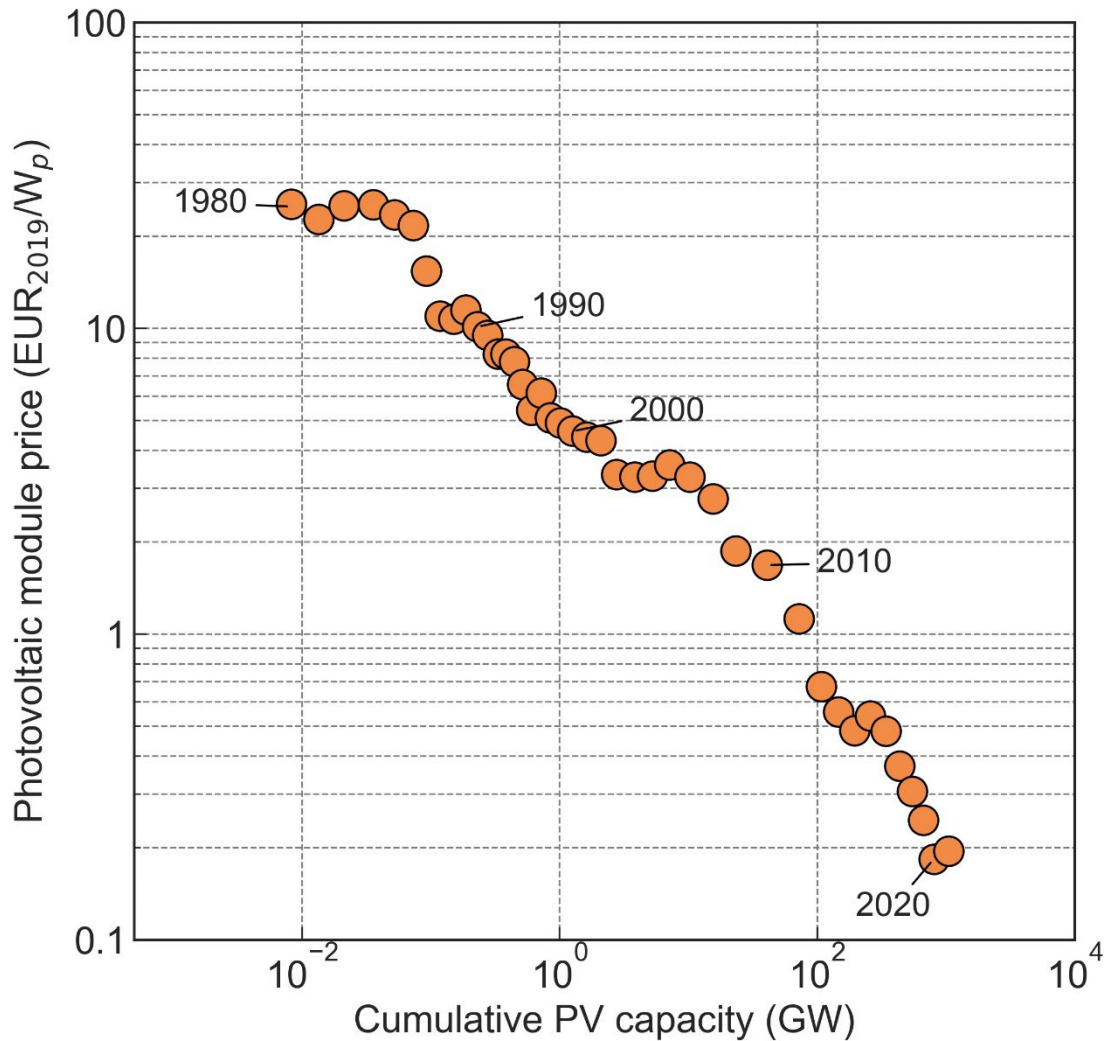
Wright, 1934

**Learning by research**



# How did solar become so cheap so fast?

Solar PV is a modular technology



Drivers:

- efficiency increase
- massive scaling and automation of manufacturing
- harsh competition reduces economic margins
- cost reduction in other system components
- reliability improvements.

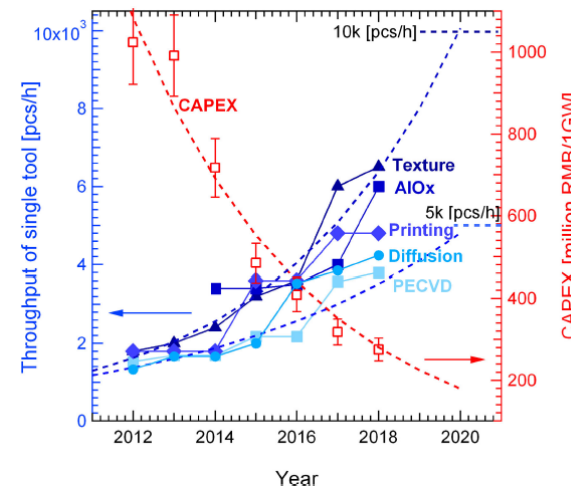
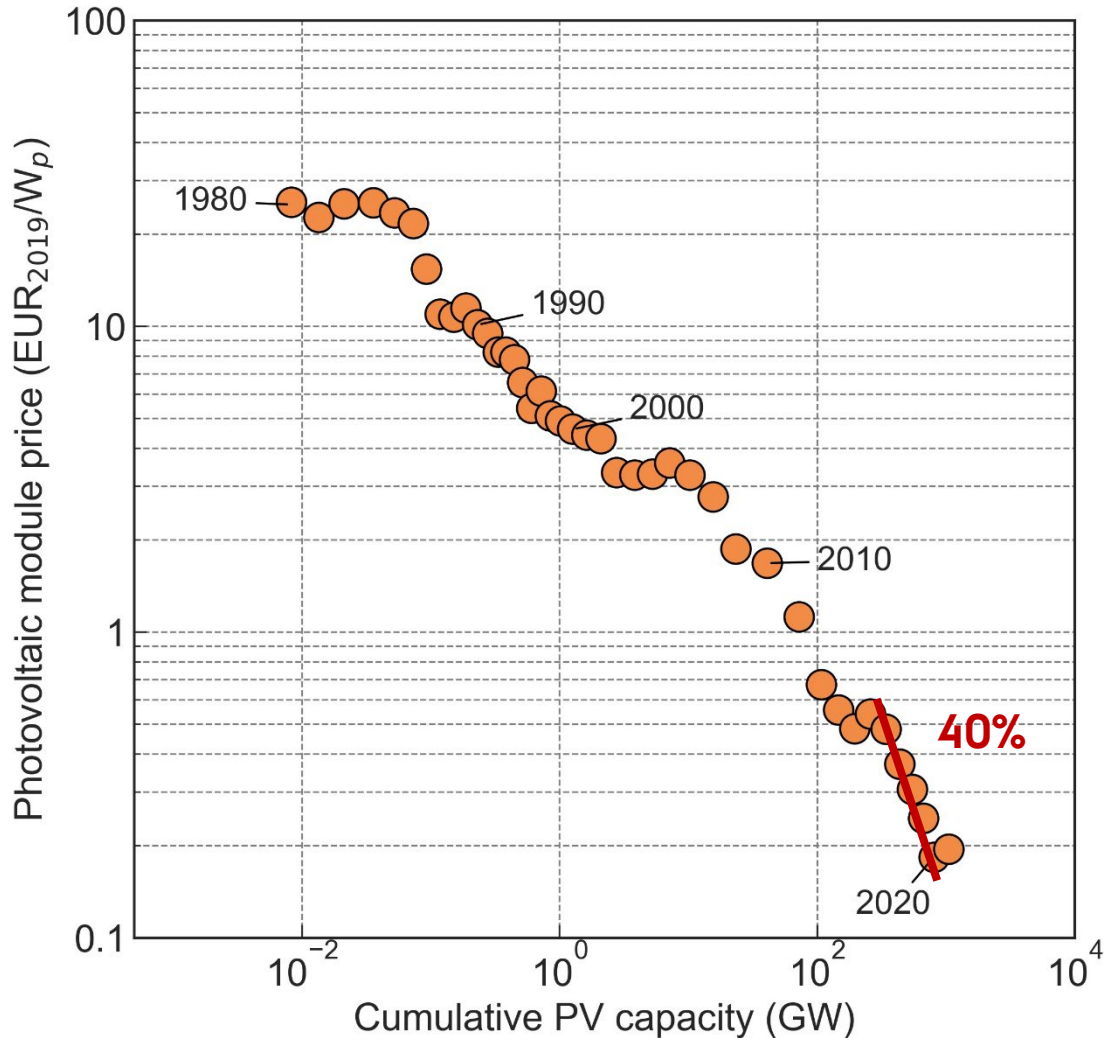
# How did solar become so cheap so fast?

## Solar PV is a modular technology

Drivers (after 2007):

- Supply chain and cell design standardized
- Manufacturing tools standardized (higher throughput)
- Historical lab-research available transferred to industry

- Learning on soft costs (labour, regulation, permitting...)
- Local or global learning
- Spill-overs from other industries
- Policies that stimulated market-growth



Chen et al., From laboratory to production, IEEE-PV Journal 2018

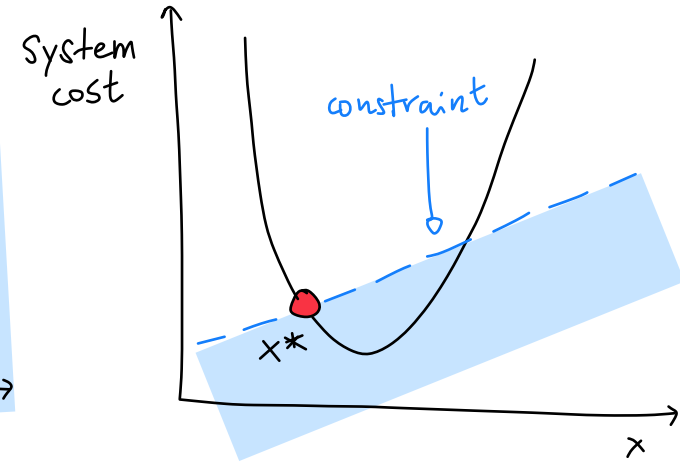
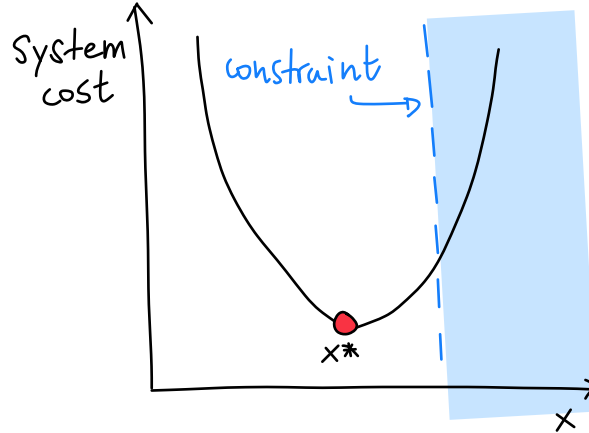
# Large-scale systems modelling

We look for cost-optimal system designs and define constraints to represent physical or societal limitations.

Detailed accounting of carbon capture, use and storage.

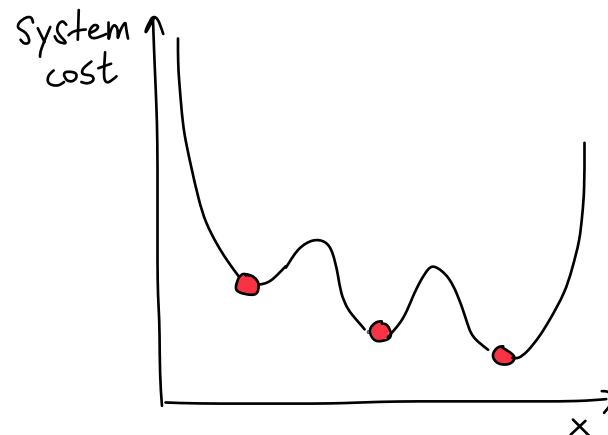
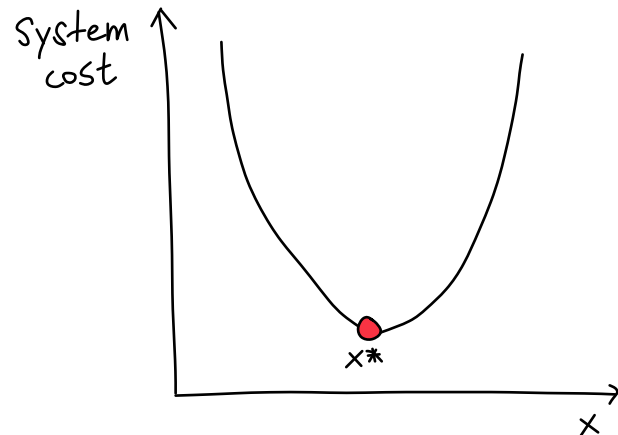
$$\left[ \begin{array}{l} \min \left( \sum_{n,s} \text{generation costs} + \text{storage costs} + \text{transmission costs} + \sum_{n,s,t} \text{variable costs} \right) \\ \text{subject to:} \\ \sum_s \text{generation}_{s,t,n} + \text{balance}_{t,n} = \text{demand}_{t,n} \leftrightarrow \lambda_{t,n} \quad \forall t,n \\ \sum_{s,t} \text{CO}_2 \text{ emissions} \leq \text{CO}_2 \text{ limit} \leftrightarrow \mu_{\text{CO}_2} \end{array} \right.$$

Cost and efficiency assumptions for the DEA Technology Data catalogue



We keep to problem linear to ensure a unique solution.

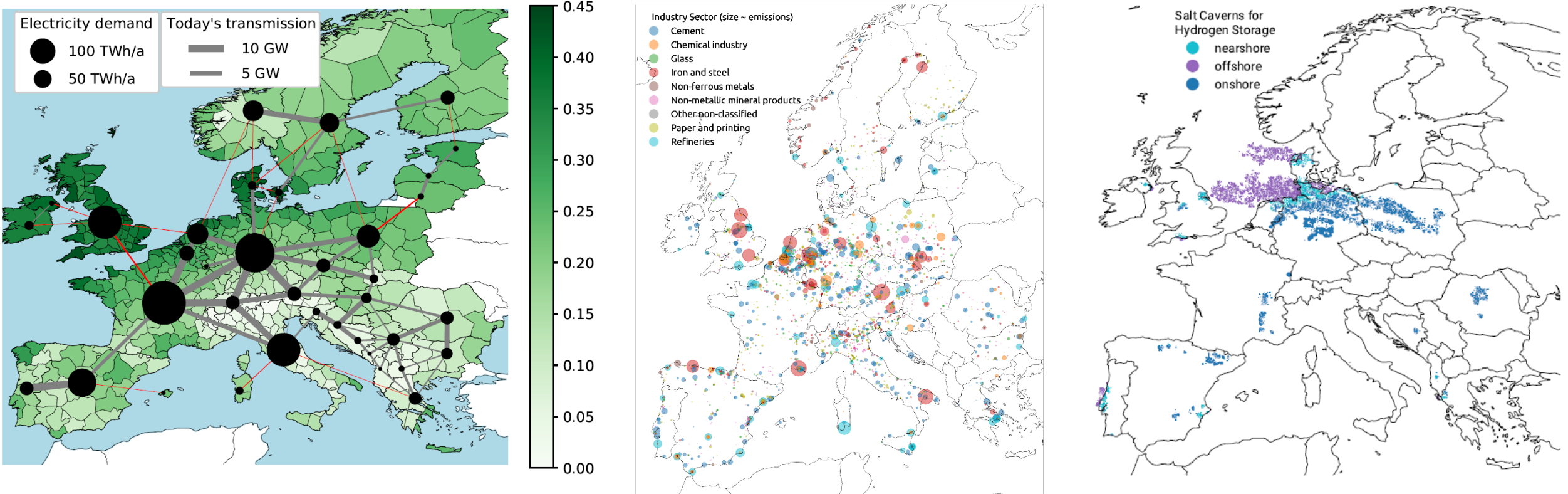
Learning is not linear, including it in the model makes it non-linear, non-convex and multiple solution exist





# Large-scale systems modelling

Our open model included detailed representation of networked sector-coupled Europe:  
Electricity, Heating, Transport, Industry and feedstock, carbon cycle



# Including learning in large-scale systems modelling

nature communications



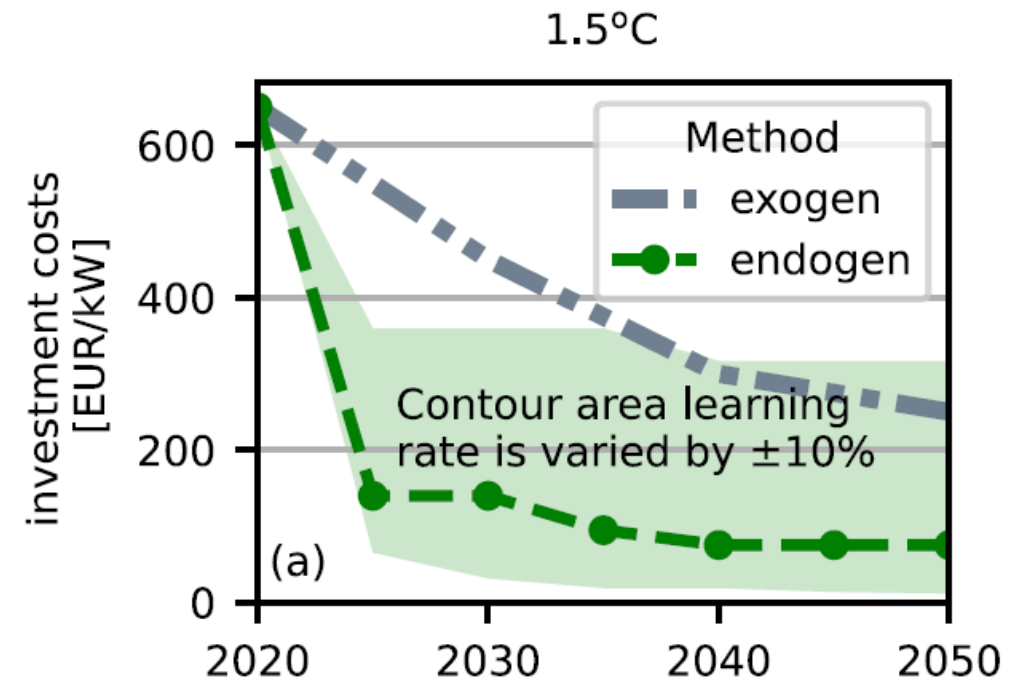
Article

<https://doi.org/10.1038/s41467-023-39397-2>

## Endogenous learning for green hydrogen in a sector-coupled energy model for Europe

Technology	Global capacity [GW]	European share [%]	Learning rate [%]
Solar PV	707 <sup>58</sup>	22 <sup>58</sup>	24 <sup>31</sup>
Onshore wind	699 <sup>58</sup>	26 <sup>58</sup>	10 <sup>31</sup>
Offshore wind	34 <sup>58</sup>	73 <sup>58</sup>	10 <sup>31</sup>
Electrolysis	1 <sup>159</sup>	Local learning	16 <sup>9</sup>

Zeyen, Victoria, and Brown, Nature Communications, 2023

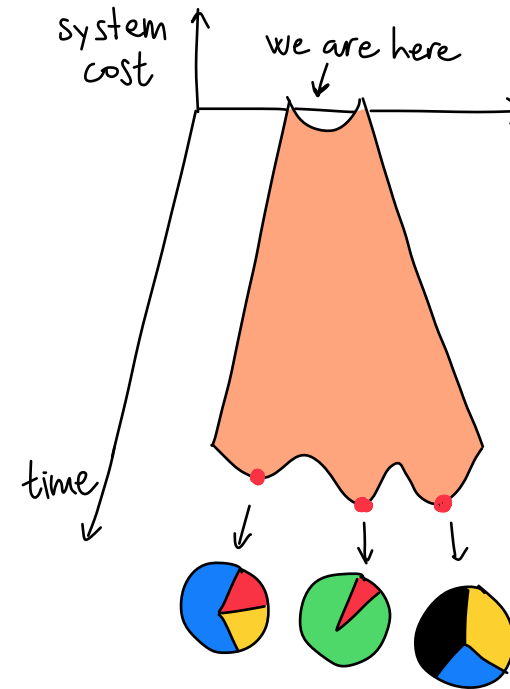
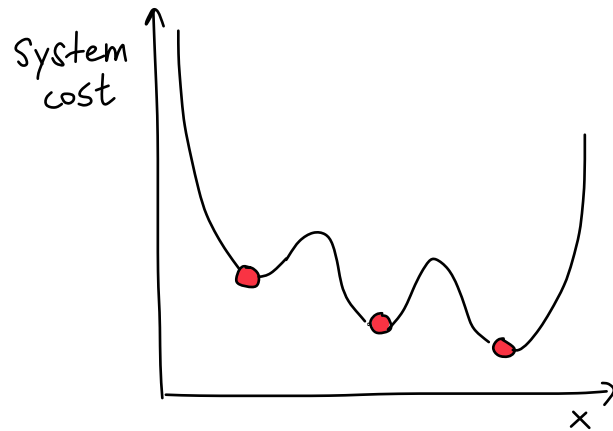


Electrolyzers deployment occurs earlier, and costs are reduced further if we consider endogenous learning

# Including learning in large-scale systems modelling

We keep to problem linear to ensure a unique solution.

Learning is not linear, including it in the model makes it non-linear, non-convex and multiple solution exist



Grub, Planetary Economics, 2014

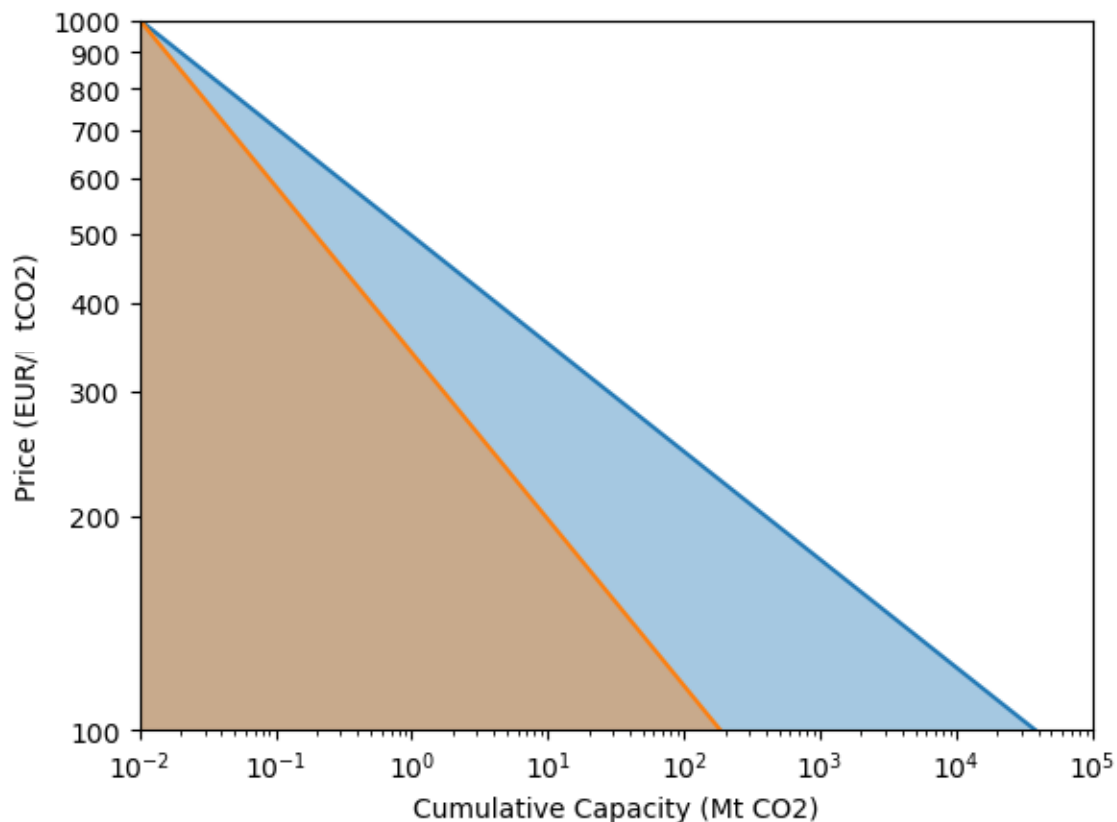
# How could learning impact CO<sub>2</sub> capture and conversion?

## A back of the envelope calculation.

Cumulative capacity for DAC today is 0.01 MtCO<sub>2</sub>/year at 1000 EUR/tCO<sub>2</sub>.

[IEA](#)  
[The state of Carbon Removal, 2023](#)

When do we achieve 100 EUR/tCO<sub>2</sub> ?



Analysis by Sina Kalwait

- With 15% learning rate, when we reach 185 MtCO<sub>2</sub>/year it costs us 24,000 million EUR it takes us 36 years
- With 10% learning rate, when we reach 37,000 MtCO<sub>2</sub>/year it costs us million 4,500,000 million EUR it takes us 60 years



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