

THE UNIVERSITY of NORTH CAROLINA at CHAPEL HILL

Background & Motivation Results Key Results of the Analysis: Hydrogen has been one of the technologies to receive the most Starting price is less important in the long-term, as the learning rate is a larger factor. Upfront capital costs could play a significant role in determining future competitive cost options for electrolyzers over the next 2-3 decades. • Projects with larger capacities can have the benefit of a noticeably decreased CAPEX.³ Cost differences between most expensive and least expensive scenarios will converge to smaller intervals as time progresses towards 2050. clean technology and is often referred to as "green hydrogen." This CAPEX is expected to decrease the most between 2022 and 2030, with a smaller decreases projected between 2030 and 2050. (PEM) electrolyzers, as there is early evidence to suggest these will Based on the CAPEX ranges of 2050, PEM electrolyzer costs are likely to decrease by a noticeable margin, even if not to the same learning rates as described in the more aggressive scenarios. 1400 1200()()(—Schmidt et al. R&D 1X electrolyzers being installed, as well as infrastructure that would be 800 ····· Schmidt et al. R&D 2XSchmidt et al. R&D 10X 600 —Schmidt et al. RD&D 1X —Schmidt et al. RD&D 2X 400 -Schmidt et al. RD&D 10X Methodology 200 2022 2026 2030 provide potential outcomes for electrolysis to meet feasible cost levels Figure: Trends of CAPEX (\$/kW) decreases from 2022 through 2050 from original Schmidt et al. scenarios⁴. Extrapolated the change in costs predicted for each Research and Development type between 2020 and 2030 to the year 2032. Next, used IEA hydrogen database to predict a doubling time for a new learning rate between 2032 and 2050. This time to double was then applied to a aggressive scenario and there are lower starting costs. electrolyzers, learning rates are estimated through the principles of **SCENARIO TYPE: LEARNING NUMBER OI SCENARIOS: RATE RANGE** • Wright's Law best fits the nature of the analysis as it determines a (%): decrease in cost of production based upon a doubling in installed **SCHMIDT** 31.93-27.8 6 **IEA NET ZERO** 73.55-5.9 IEA 12 29.56-5.09 • Given that most available electrolyzer data contained either a price CONSERVATIVE or expected capacity, Wright's Law was the best fit to determine a 31.93-25 NREL 9 **INDUSTRY** 12 25

recent attention as of late, specifically within the U.S. Inflation Reduction Act (IRA) production tax credit. The hope is that hydrogen can provide great support in goals of decarbonization. The prevalent questions are relating to what role the technology will play, and how hydrogen will reach regions of costcompetitiveness when compared to other green technologies. Hydrogen, if produced with clean electricity, can be considered a analysis focuses specifically on Polymer Electrolyte Membrane be the most cost effective electrolyzers, with support from government, industry, and academia in this claim.

Variations of measurement for the cost of hydrogen are important to consider. The focus of this analysis will be CAPEX. The units for CAPEX are measured as the amount of upfront capital cost necessary to produce one unit of electricity, typically reported in \$/kW or \$/MW. This specific cost unit is especially applicable when considering the capacities of different necessary outside of primary production.

The paper analyzed 45 scenarios for hydrogen adoption, and thus through a scale up to meet worldwide expectations for hydrogen production capacity.

- To determine future scenarios for decreasing the costs of PEM Wright's Law.
- capacity.
- learning rate for specific scenarios.
- The general Wright's Law formula for a decreasing cost reads as: $Y = aX^{-b}$

Where: Y = the cost per unit (\$/kW), X = the cumulative number of units produced (GW), a = cost to produce the first unit (kW), and b = slope of the function^{1,2}.

Estimating PEM Electrolyzer Costs for Hydrogen Production Through 2050

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Table: Summary of various scenarios used in analysis. Includes the number of scenarios used for each type, as specified in the methodologies; the minimum and maximum of the used learning rates for each type; the median learning rate for further clarity on the true middle of the learning rates used; the CAPEX starting points in 2022 for each type of scenario; and the CAPEX ending points in 2050 for each type. Plug Power were the scenarios with the largest possible ending CAPEX but also had the most consistent learning rate, as 25% was used for all years.

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2034	2038	2042	2046	2050
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general learning rate of 25%. Scenario with RD&D 10X starts and ends as the lowest CAPEX scenario as this prediction is the most

MEDIAN	2022 CAPEX	2050 CAPEX
LEARNING	RANGE (\$/KW):	RANGE (\$/KW):
RATE (%):		
29.39	1225-867	100.39-53.35
8.81	1225-867	172.95-71.46
25	1225-867	283.49-132.42
25	1503	339.04-100.3
25	1225-867	500.92-241.65

CAPEX (\$/kW):	1600 1400 1200 800 600 400 200 0	
Fi 205 an	gure: CAP 50. Using a d 2050 are disp	EX bo dis olay
In fr •	order f uition th Contin produc Along essenti The bi further More o electro allow f	for he iue tic wi al ase di ase oly: for
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2022 2030 2050

distributions across all scenarios in analysis for the years 2022, 2030, and ox and whisker plot to display inter-quartile ranges for the years 2022, 2030, splayed. Outside of two outliers in 2030, no other year has outliers present, ying a relatively continual decrease in CAPEX between the years.

Discussion

the most aggressive learning rates to come to following are recommendations: ed support for policies like the IRA, allowing the on of new technologies at a much lower cost. ith research and development, deployment is to the process of furthering the learning curve. es associated with CAPEX as a metric should be liscussed to help with research accuracy. en and transparent access to data regarding zers would increase the accuracy of analysis and ^r more efficient learning processes.

References

, Azevedo, I. M. L., Jaramillo, P., & Yeh, S. (2015). A review of learning ctricity supply technologies. *Energy Policy*, 86, 198–218. https://doi.org/10.1016/j.enpol.2015.06.011

2. S. Ziegler, M., & E. Trancik, J. (2021). Re-examining rates of lithium-ion battery technology improvement and cost decline. Energy & Environmental Science, 14(4), 1635–1651. https://doi.org/10.1039/D0EE02681F

Lazard. (2021). Lazard's Levelized Cost of Hydrogen Analysis—Version 2.0. Lazard. https://www.lazard.com/media/451922/lazards-levelized-cost-of-hydrogen-analysis-

4. Schmidt, O., Gambhir, A., Staffell, I., Hawkes, A., Nelson, J., & Few, S. (2017). Future cost and performance of water electrolysis: An expert elicitation study. International Journal of Hydrogen Energy, 42(52), 30470–30492. https://doi.org/10.1016/j.ijhydene.2017.10.045