# Learning Learning Curves 

Ron Rivest<br>MIT<br>rivest@mit.edu

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## Happy Birthday, Rob!

## Overview

(1) Goal: to give you a nice "open problem"
(2) Learning Curves (aka "experience curves")
(3) Learning one learning curve
(4) Learning multiple learning curves (multi-armed bandit formulation)
(5) Open problem

## Learning Curves

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- Let $X=$ number of units produced so far. Let $C(X)=$ cost of producing $X$-th unit
- $C(2 X)=C(X) \cdot(1-\lambda)$ where $\lambda=$ learning rate (e.g. $\lambda=0.20$ )


## PV Solar Learning Curve



Data: Lafond et al. (2017) and IRENA Database; the reported learning rate is an average over several studies reported by de La Tour et al (2013) in Energy. The rate has remained very similar since then. OurWorldinData.org - Research and data to make progress against the world's largest problems.

## Energy Learning Curves

Electricity from renewables became cheaper as we increased Our World capacity - electricity from nuclear and coal did not
Price per megawatt hour of electricity
This is the global weighted-average of the
levelized costs of energy (LCOEI, without subsidies
logmithmic avis and adiusted for inflation
5300

Source IRENA 2020 for al dati on renewatie wources Lazed for the price of electricity from nuclear and cod - LAEA for nuclear capacity and Cabal Energy Martar for caal
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- This is standard simple linear regression problem. (Use least-squares; details omitted.)


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- For example: $p(t)=1 / \sqrt{t}$.


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- At each step, choose technology $k$ (for some $k$ ):
- increase $x_{k}$ by $\delta_{k}$.

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- Obtain actual cost $c_{k}$.
- Infer new parameters $\alpha_{k}, \beta_{k}$, and $\sigma_{k}^{2}$ using least squares on sample of size $n_{k}$, where $n_{k}$ is number of times technology $k$ has been chosen.


## Choosing best technology $k$

- How to choose best technology $k$ to use at time $t$ ?
- Estimate $\hat{c}_{k}(T)=$ estimated log cost of energy at time $T$ using only technology $k$ from now $(t)$ on:

$$
\hat{c}_{k}(T)=\alpha_{k}+\beta_{k}\left(x_{k}+(T-t) \delta_{k}\right)
$$

- Do this whenever technology $k$ is used.


## Learning the best technology

$\operatorname{Learn}(T, K)$ :
for $k=1,2, \ldots, K$ : use technology $k$ twice.
for $t=2 K+1$ to $T$ :
with probability $p(t)$ : \# Explore
Use technology $k$, where $k=$ a least-used technology.
else: \# Exploit
for $k=1,2, \ldots, K$, estimate $\hat{c}_{k}(T)$ using what's been learned so far, using least-squares to get $\hat{\alpha}_{k}, \hat{\beta}_{k}$ :

$$
\hat{c}_{k}(T)=\hat{\alpha}_{k}+\hat{\beta}_{k}\left(x_{k}+(T-t) \delta_{k}\right)
$$

Use technology $k$, where $k$ minimizes $\hat{c}_{k}(T)$
return $\min _{k} \hat{c}_{k}(T)$

## Conjecture (Open Problem)

For all sets of $K$ learning curves and all $T$, LEARN returns a result $\hat{c}_{k}(T)$ such that with high probability

$$
\hat{c}_{k}(T) \text { is "not much more than" } c_{k_{*}}^{\prime}(T)
$$

where $k_{*}$ is the value of $k$ with minimum expected value $c_{k_{*}^{\prime}}^{\prime}(T)$ (that is, where $k_{*}$ is always used).

## Thanks!

## Happy Birthday, Rob!

## References

(1) Ghemawat P., Building Strategy on the Experience Curve, HBR, 1986.
(2) Harvey H. and Gillis, J. The Big Fix. (Simon \& Schuster; 2022).
(3) Roser, M., Why did renewables become so cheap so fast?, 2020. https://ourworldindata.org/cheap-renewables-growth
(9) Way, R. et al., Empirically grounded technology forecasts and the energy transition, 2021, INET Oxford Working Paper No. 2021-01.

