

Big data, little data, and virtual twins:

*Accelerating process development for
semiconductor device fabrication*

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Why can't we design a process like we design a chip?

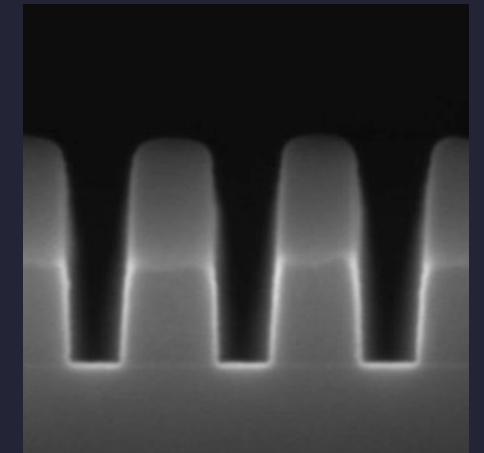
SPEC



RECIPE



RESULT





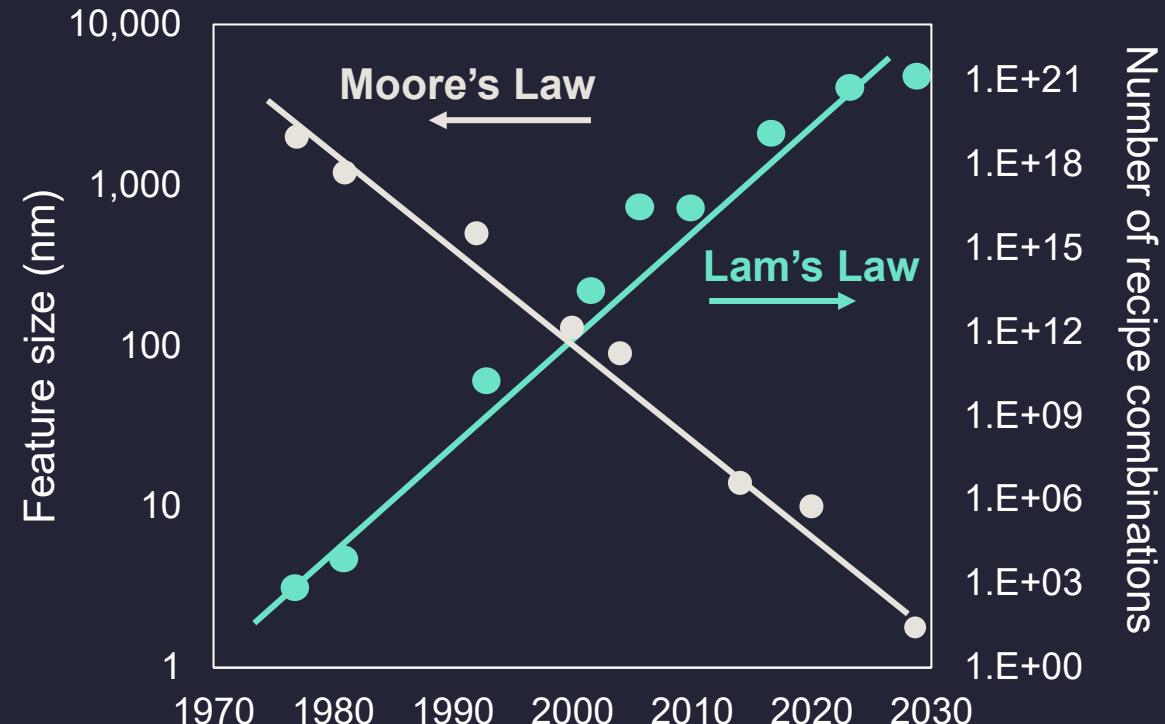
Why not just use a
big data approach?

Simply put,
it costs too
much and
takes too long



Little data
world but *big*
dimensional
space

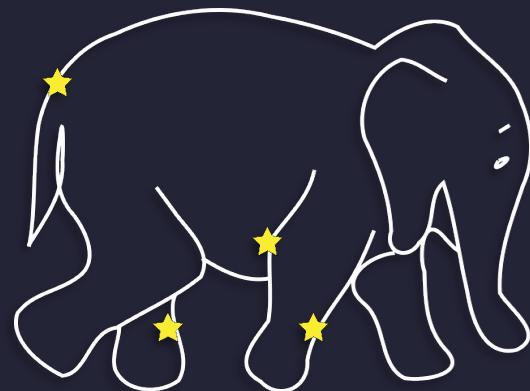
~ AVOGADRO'S NUMBER OF RECIPES



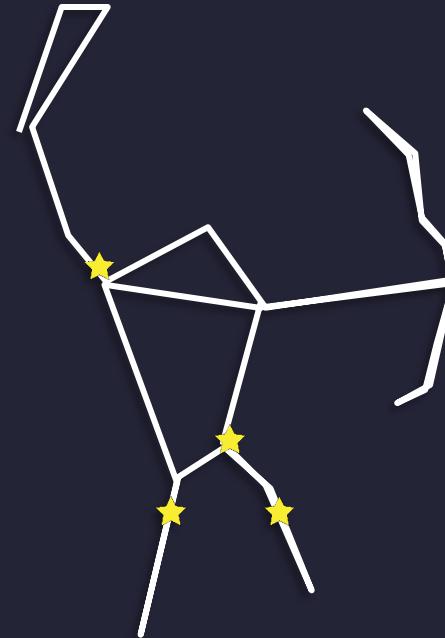
What about physics?...

Exploit little data with right (physics-based) model

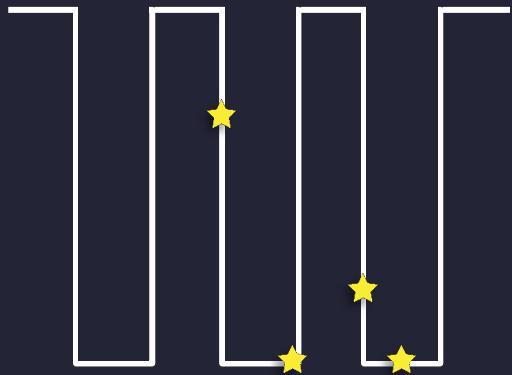
MODEL 1



MODEL 2



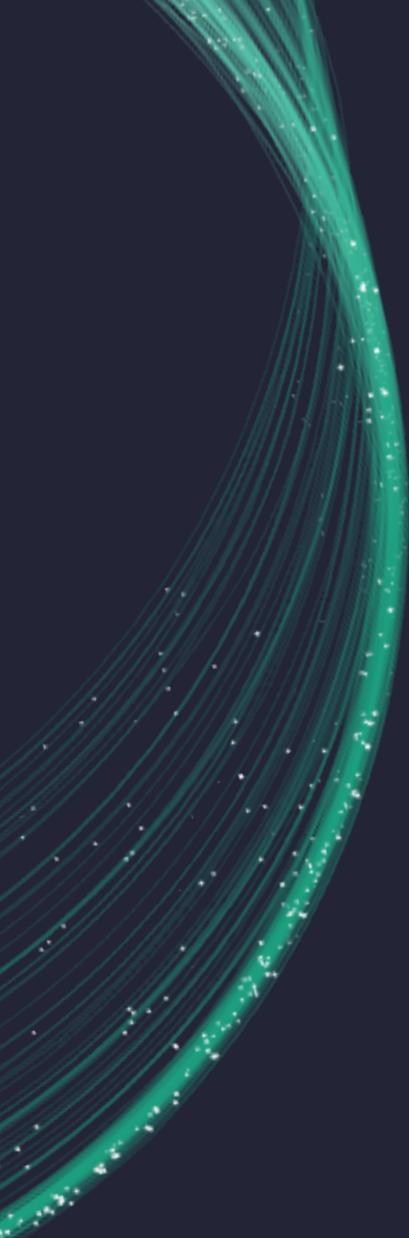
MODEL 3



"With four parameters I can fit an elephant, and with five I can make him wiggle his trunk."

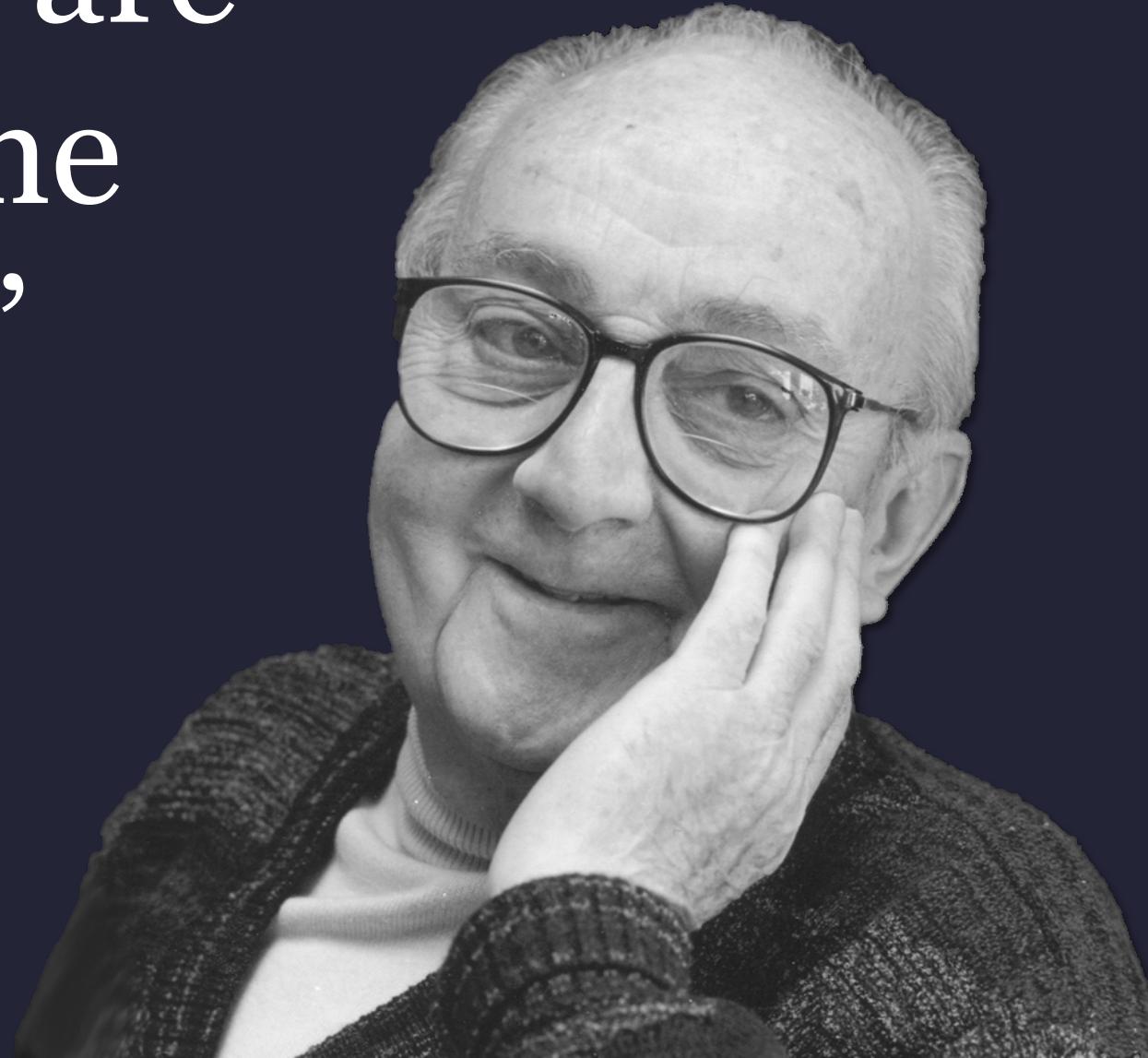
*John von Neumann, as related by Freeman Dyson (2004)
"A meeting with Enrico Fermi," Nature 427 (6972)*



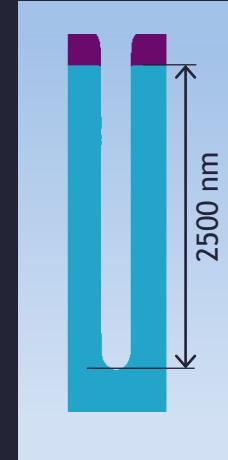


“All models are
wrong, some
are useful.”

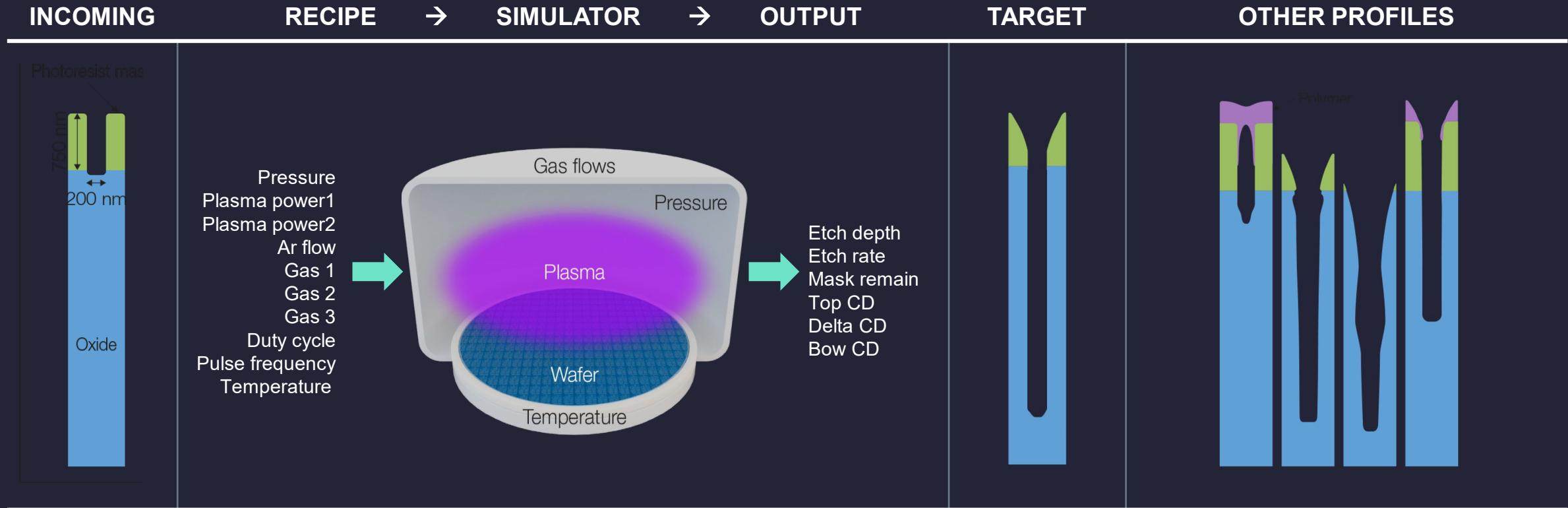
George Box, 1976



Let's play a “game” to find the useful models

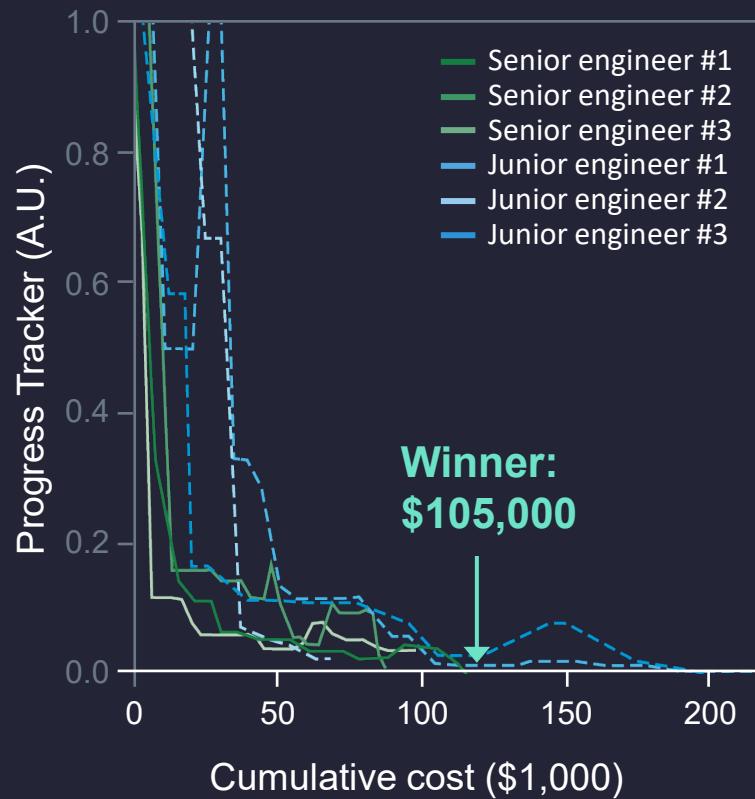


A virtual plasma etch process “cousin”

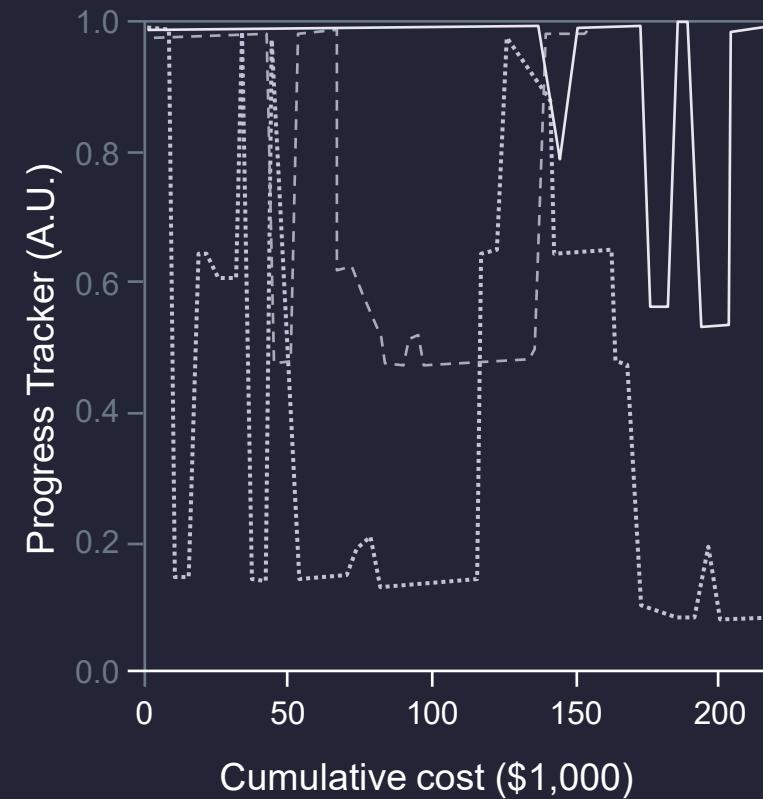


Machine **alone** was no match for expert engineer

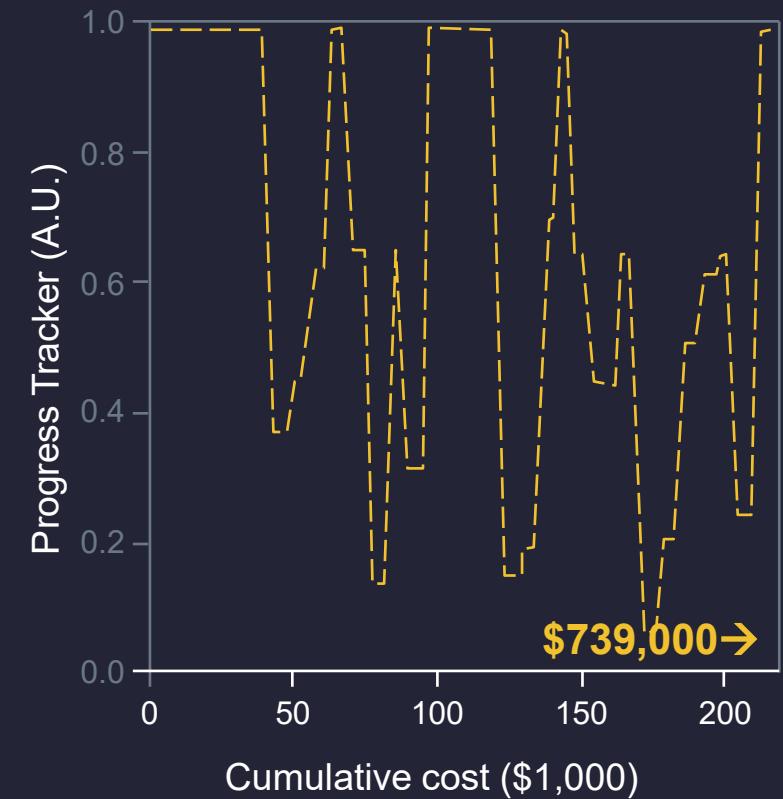
Process engineers



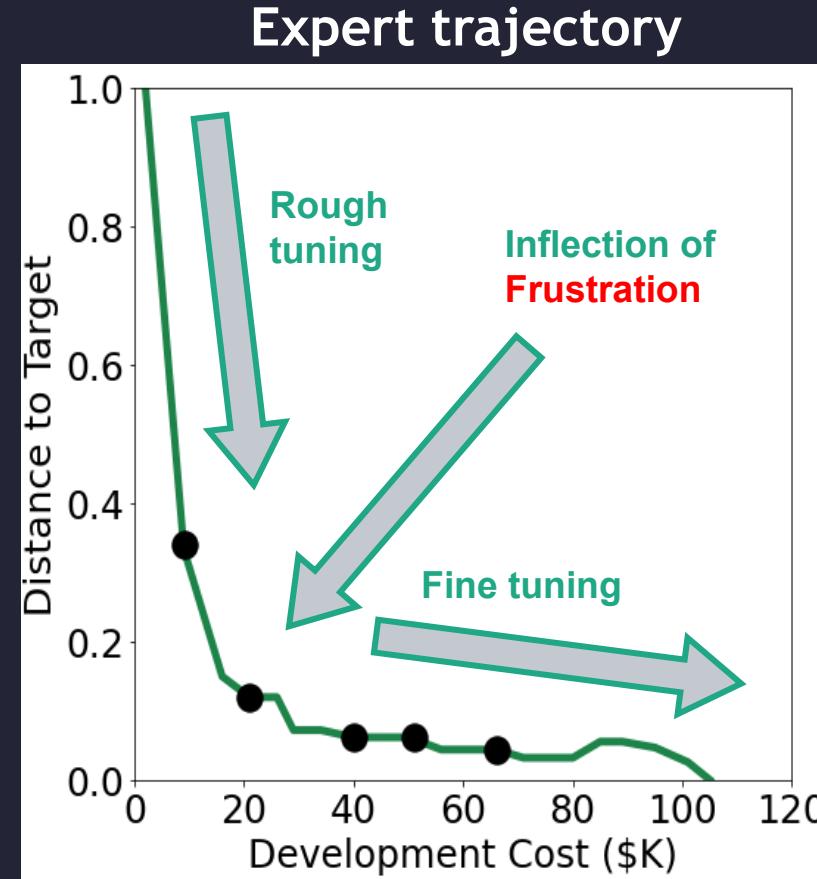
Inexperienced humans



Computer algorithm

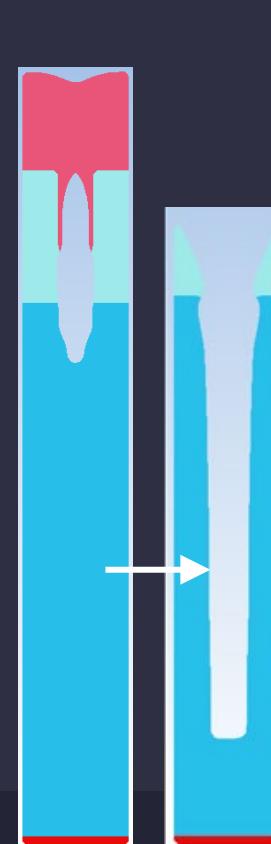


Human learning curve consists of rough and fine tuning



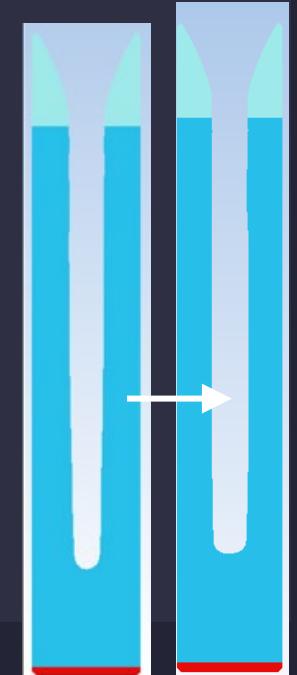
Rough-tuning stage

- Baseline from experience
- Domain knowledge and physical intuition are valuable
- Fulfilling, rapid progress *toward* solution

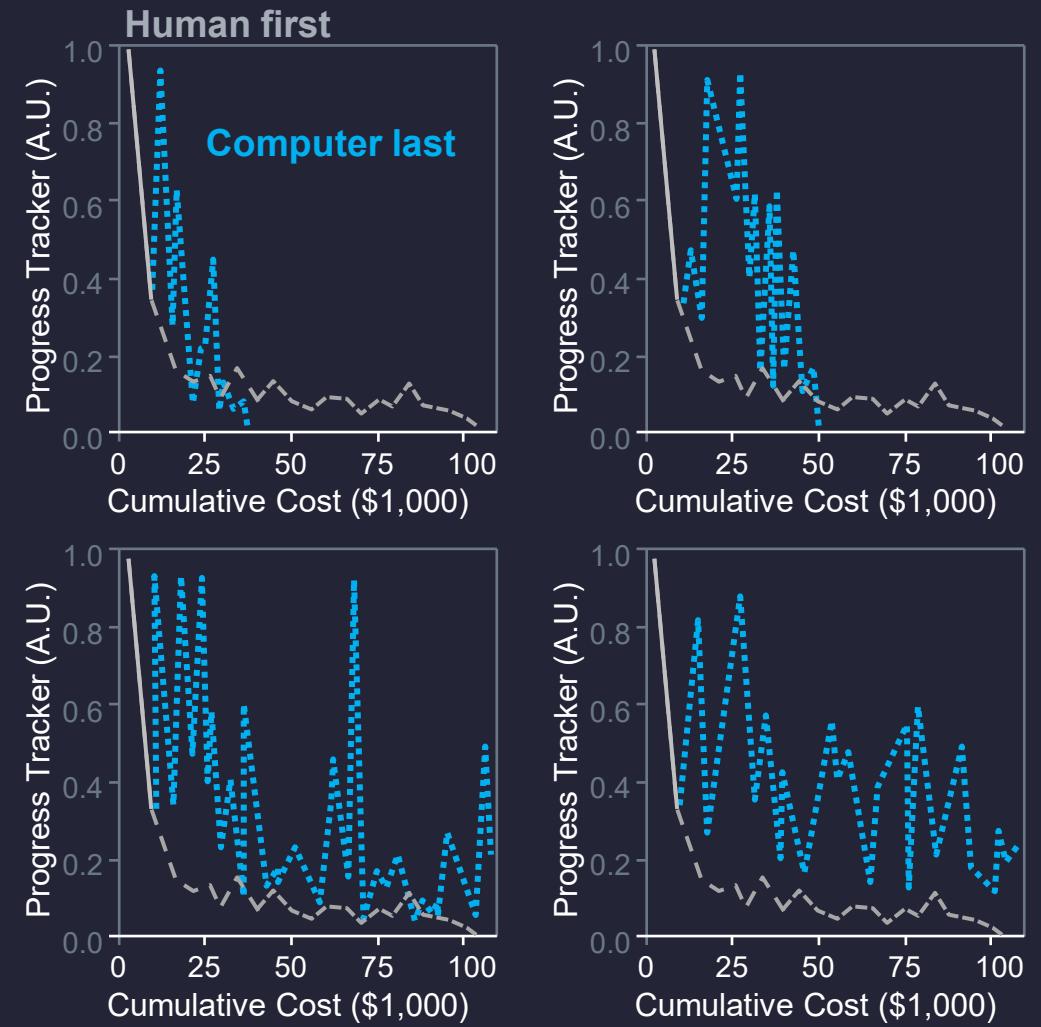


Fine-tuning stage

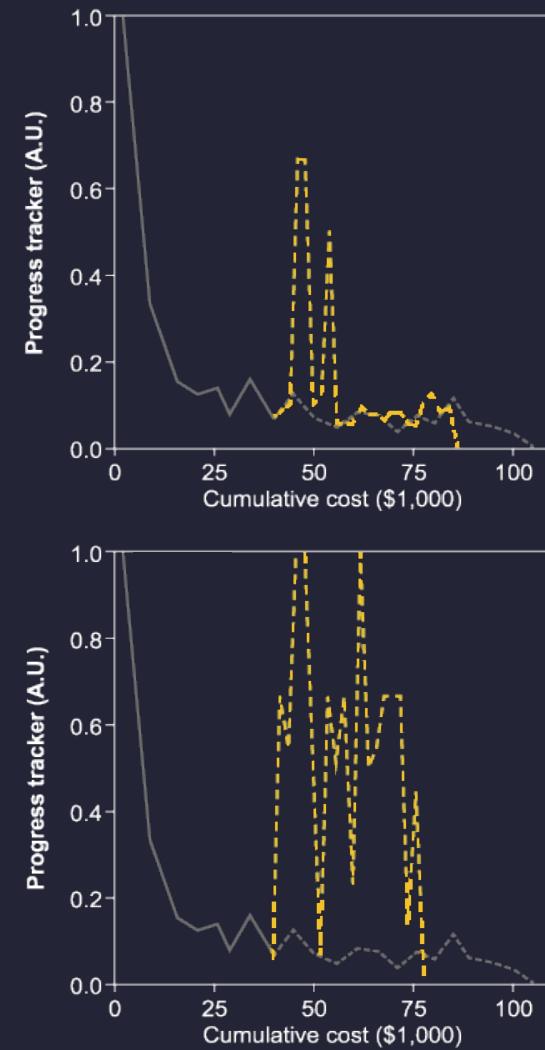
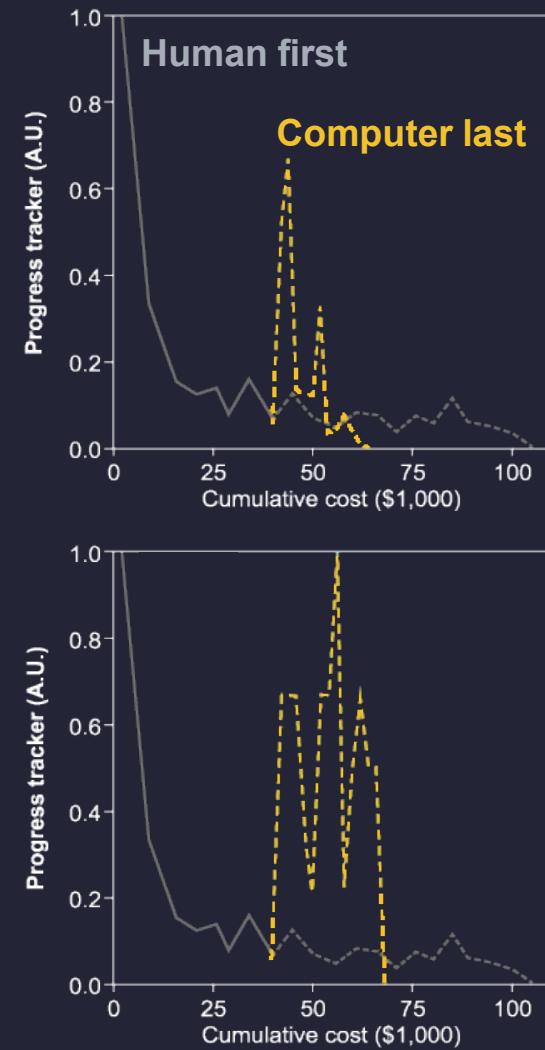
- Close to spec
- Physical intuition and domain knowledge less useful
- Frustrating, low-productivity path to solution



Human-Machine Collaboration: Transfer point “A”

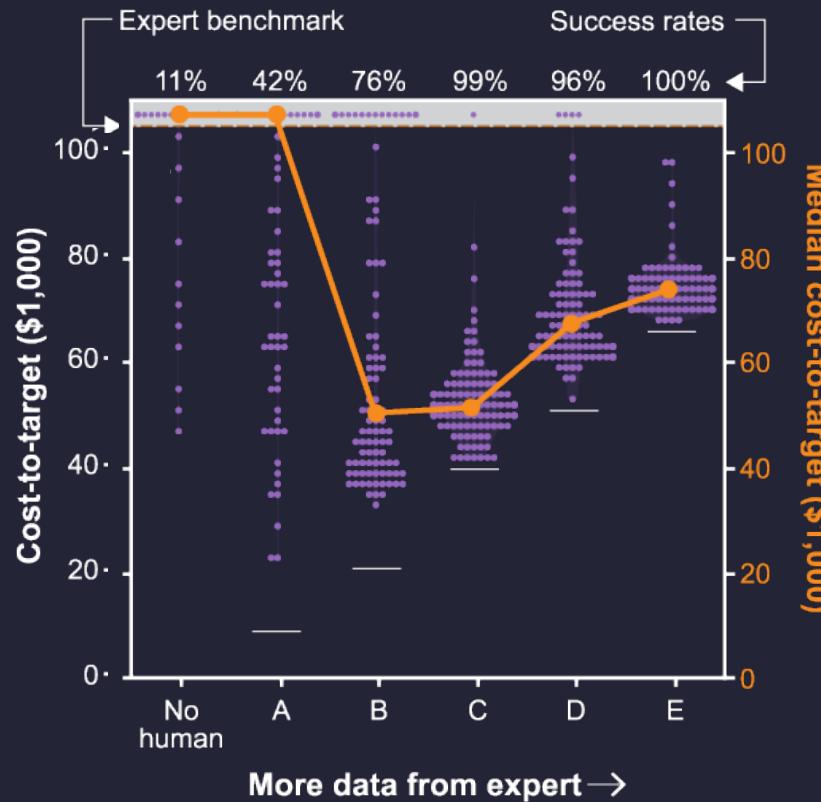


Human-machine collaboration yields cost and time savings

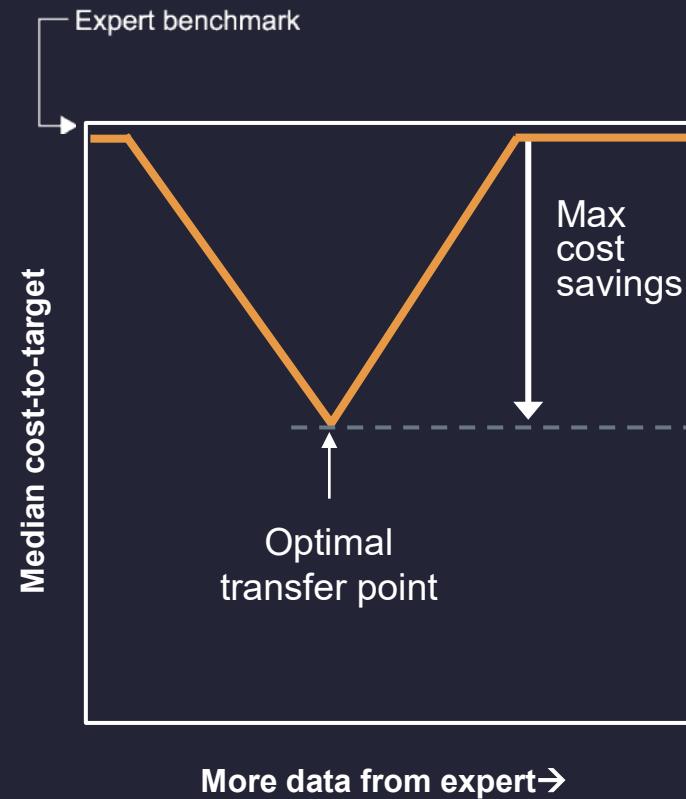


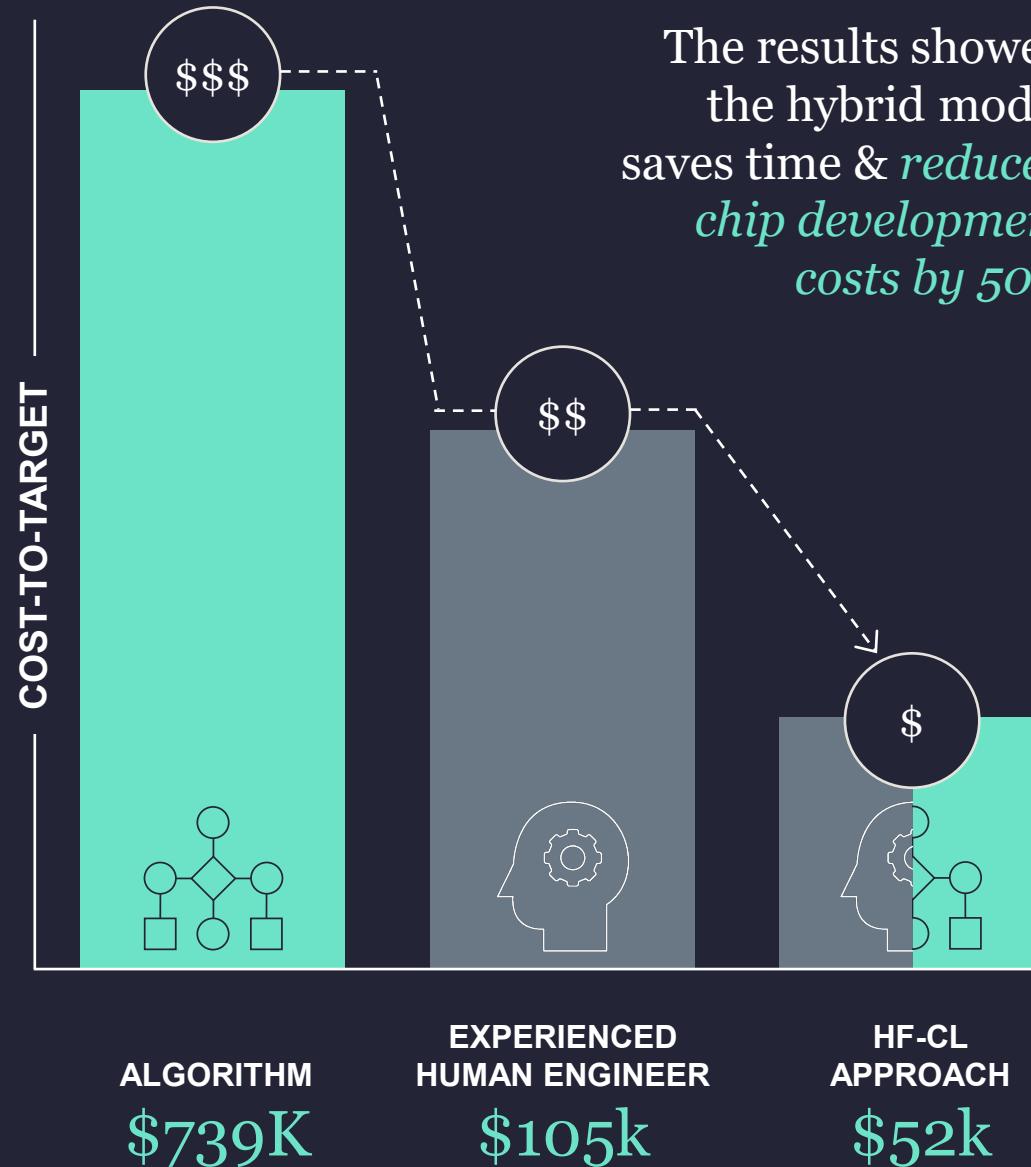
Optimal transfer leverages human investment

Experimental V-curve



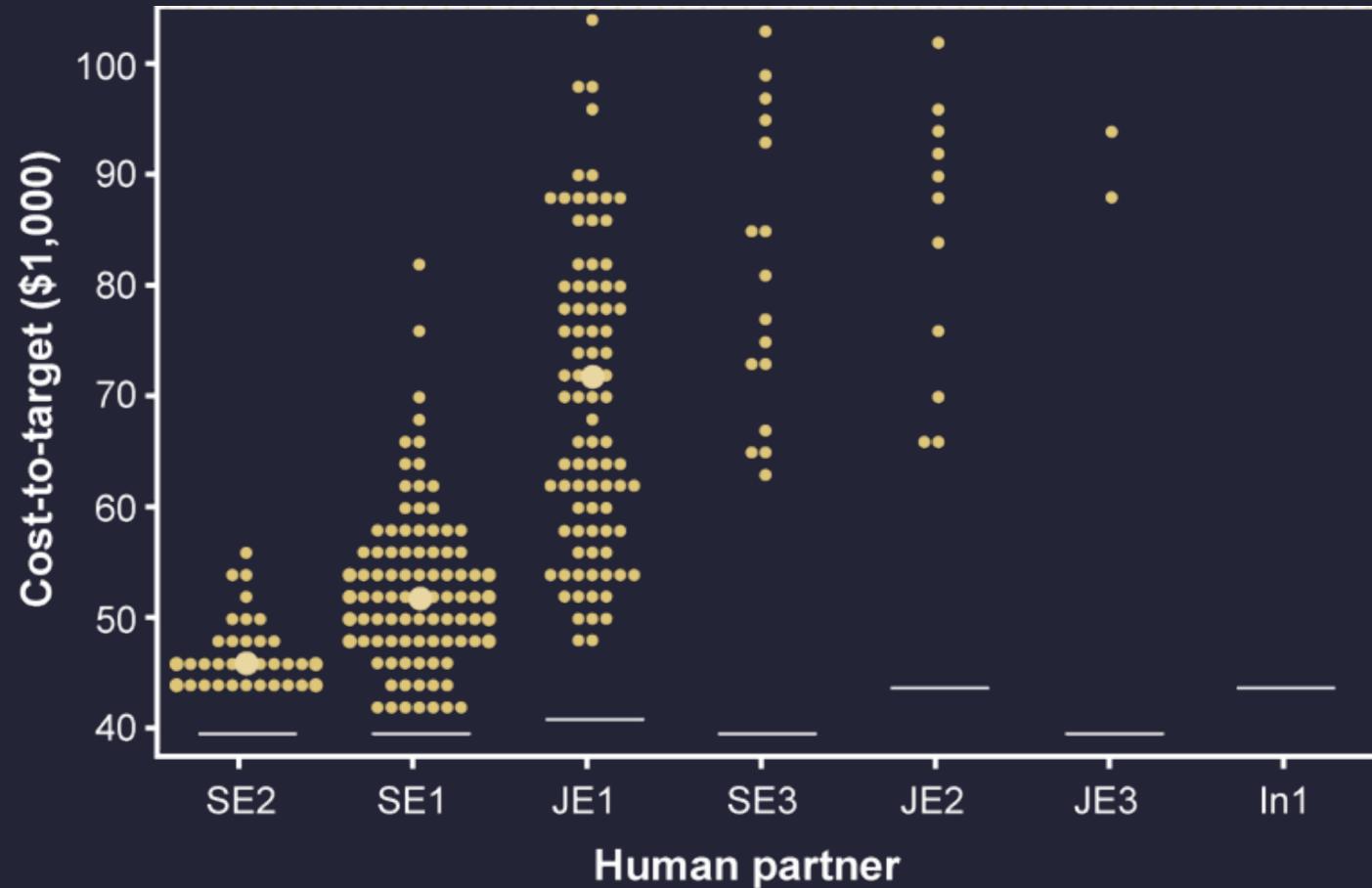
Schematic



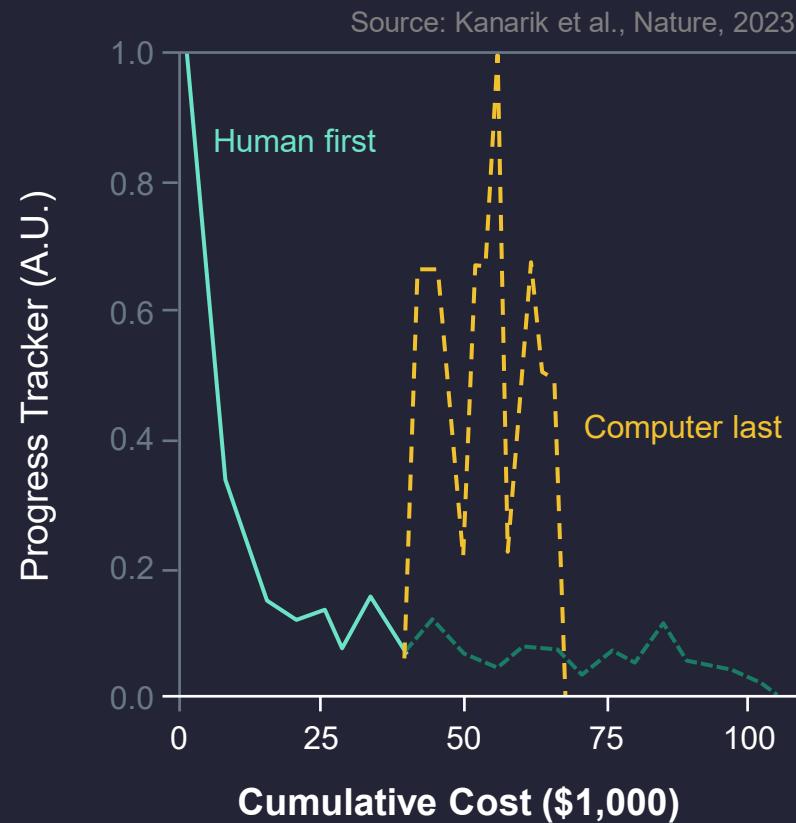


Hybrid approach wins
**Human-first,
machine-last
saves countless
hours and
millions of dollars**

Computer should partner with an **experienced** engineer



Algorithm behaves differently than process engineer



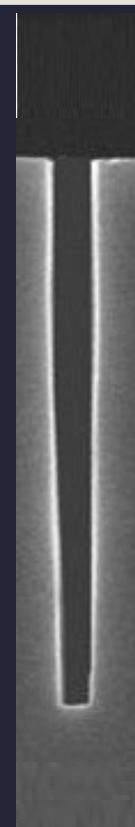
Parameter 1	Parameter 2	Parameter 3	Parameter 4	Parameter 5	Parameter 6	Parameter 7	Parameter 8
1148.6	68.5	4026	90.7	33.9	20.9	220.0	50.9
1165.2	66.5	3594	198.7	33.3	22.3	231.0	58.4
1166.8	67.1	3480	167.6	32.6	21.3	226.1	58.2
1149.3	68.3	3842	109.2	30.7	17.9	252.7	58.3
1160.1	60.5	3110	181.0	27.2	17.8	204.5	58.1
1158.0	60.0	3103	156.8	27.0	17.8	202.9	58.0
1143.9	68.6	3550	90.1	33.4	16.1	180.0	59.5
1137.1	67.3	3715	96.7	34.1	17.4	180.6	59.5
1160.5	67.7	3830	169.9	30.2	18.0	199.4	57.0
1170.7	67.0	3728	196.3	29.2	17.5	195.7	56.3
1161.6	67.2	3687	181.9	30.2	17.7	194.5	56.0

There is high value
learning from virtual
worlds that *are not*
precisely predictive

Real vs. virtual processes : Cost and time comparison

Real process in lab

- Build cost: **>\$1,000,000** to purchase reactor for the laboratory



Per recipe:

- Cost per recipe: **\$1000**
- Time per recipe: **half day**

Simulation of process

- Build cost: **\$100,000** person-hours to program HARC application into SEM3D



Per recipe:

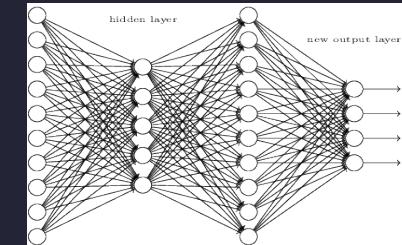
- Cost per recipe*: **\$0.11**
- Time per recipe: **8 min**

Emulator (model of simulator)

- Build cost: Used **\$30,000** of data (240,533 simulations) to train the neutral network

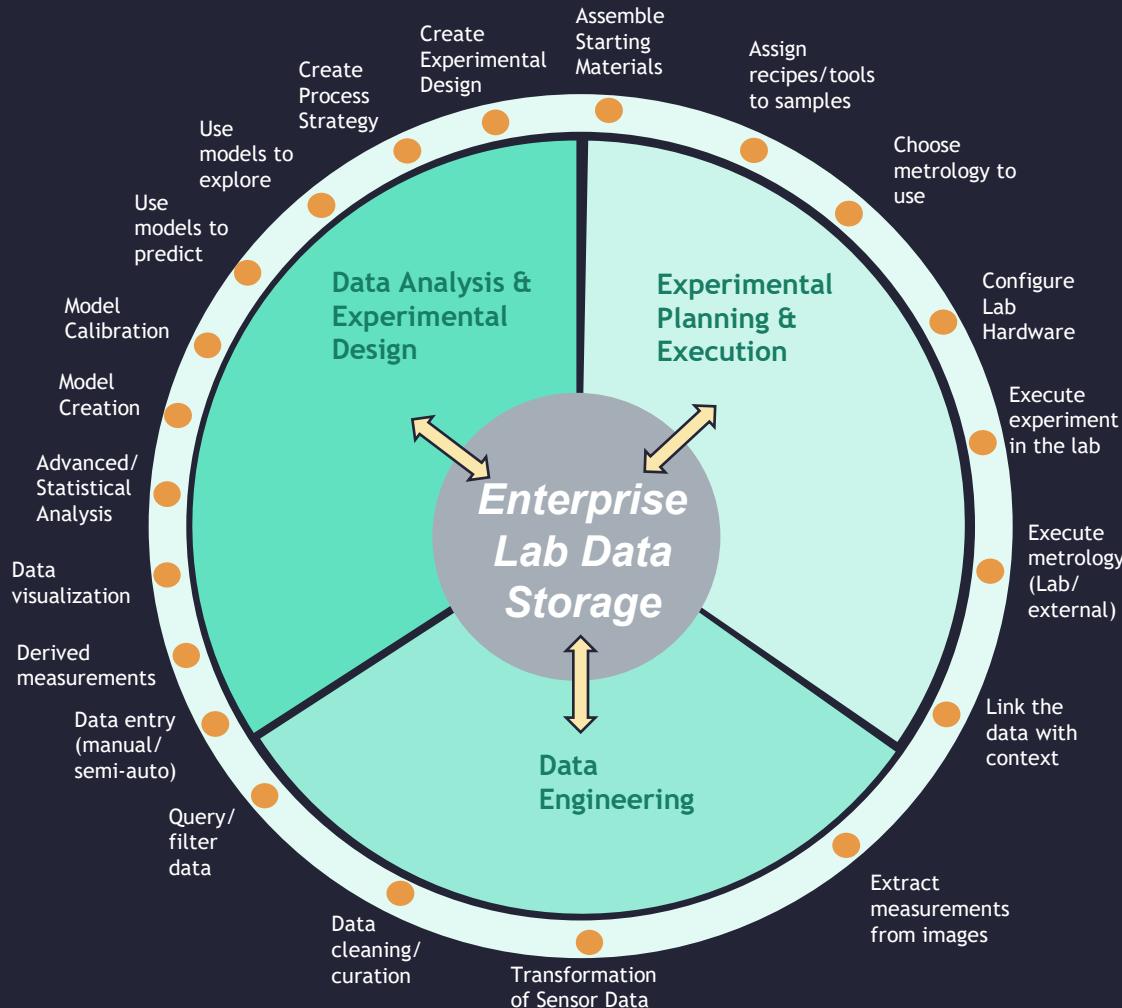
Per recipe:

- Cost*: **\$3e-07**
- Time: **0.0013 s**



Virtual Process Development

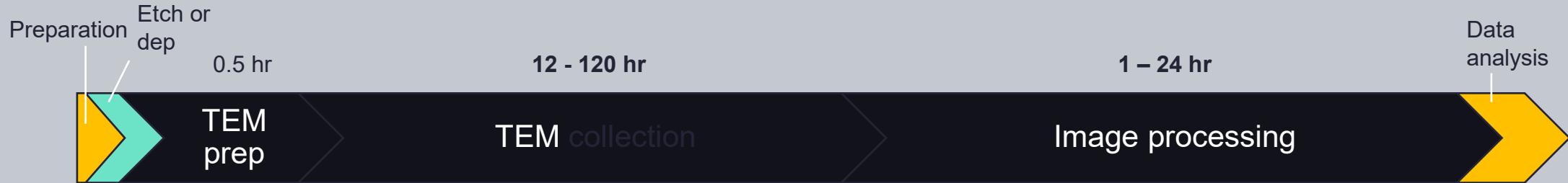
Transform process development through digitalization, automation, simulation & data analysis



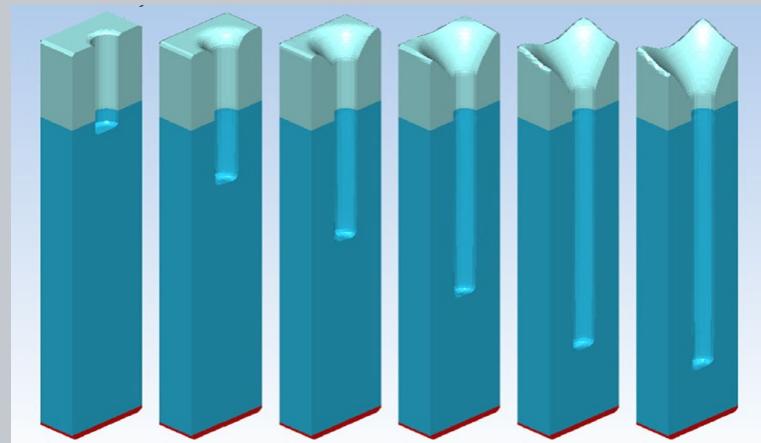
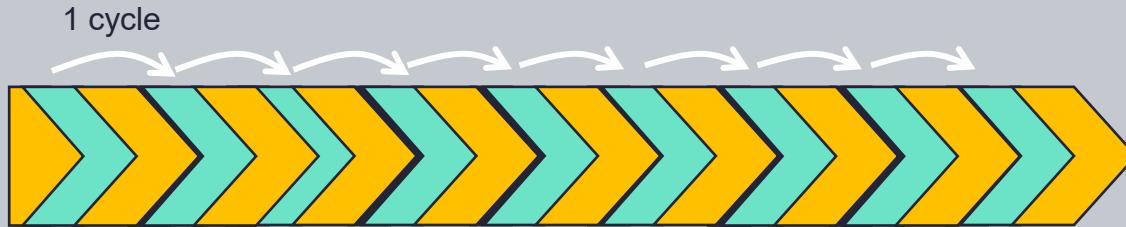
- Process Development is not one monolithic workflow. It is many different paths through a variety of different activities. Catering to these varied workflows requires a **holistic strategy**.
- The activities largely reside in three disciplines, with specific requirements, and must be **connected through enterprise-scale storage of experimental process data**.
- Modernizing and **automating** physical experimental activities in the lab is key to delivering the contextual data to the data store
- **Image analysis** and **flexible platforms for data science**, machine learning and advanced analytics are critical for data engineering.
- Connecting platforms and systems to create efficient, friction-free workflows = **Virtual Process Development**

Real-time profile metrology for *100x* cycle time reduction

Metrology for high aspect ratio solution development **costly, time-consuming, and destructive**



Real-time metrology offers **100x cycle time reduction**



Meet the authors

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