

Inverse Design: Why Aren't We There Yet?

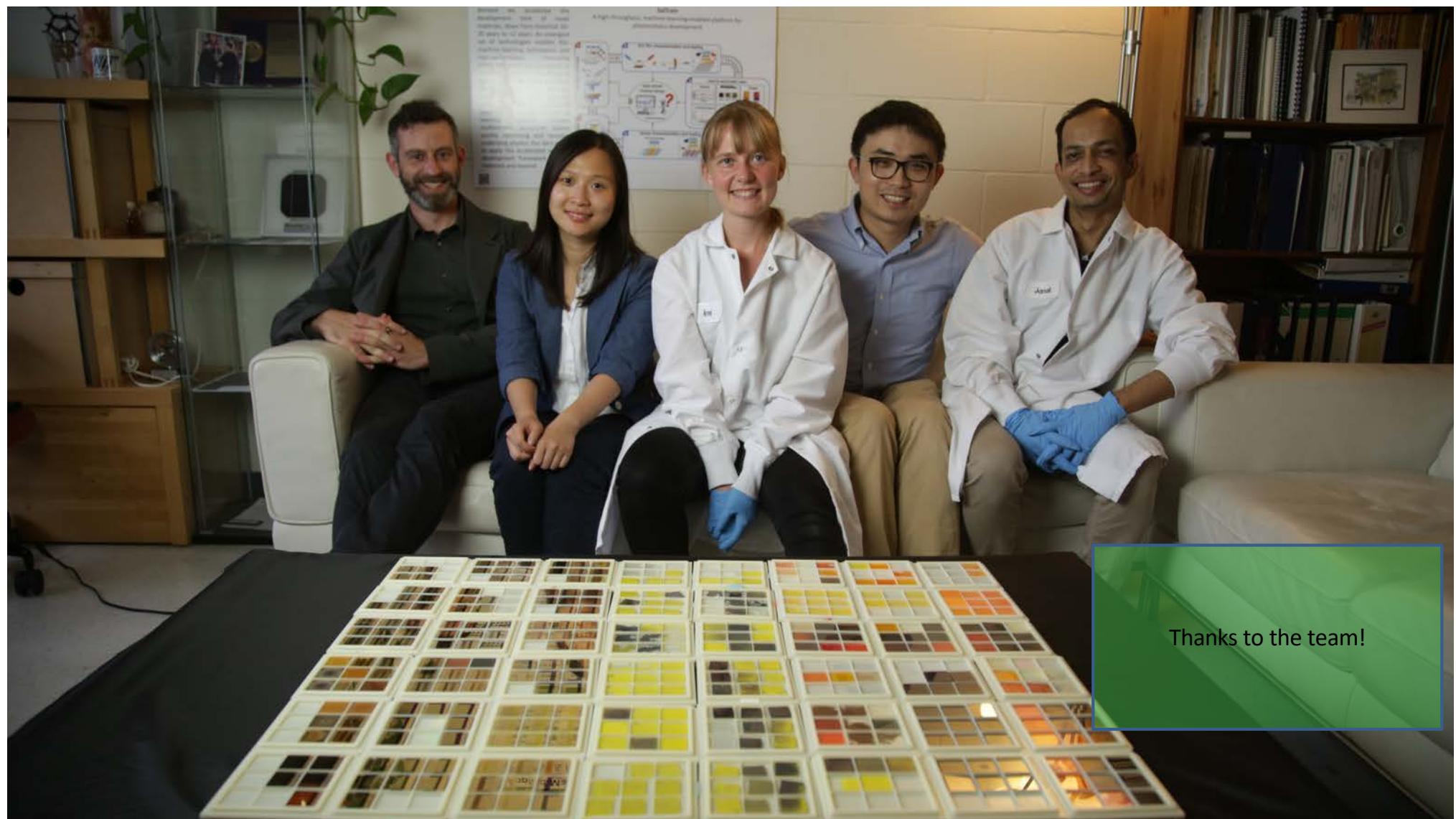
Tonio Buonassisi (MechE, MIT)



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General Inverse Design of Crystalline Materials: Why Aren't We There Yet?

A refinement of the broader title.
This presentation focuses on
crystalline inorganic materials.



Thanks to the team!



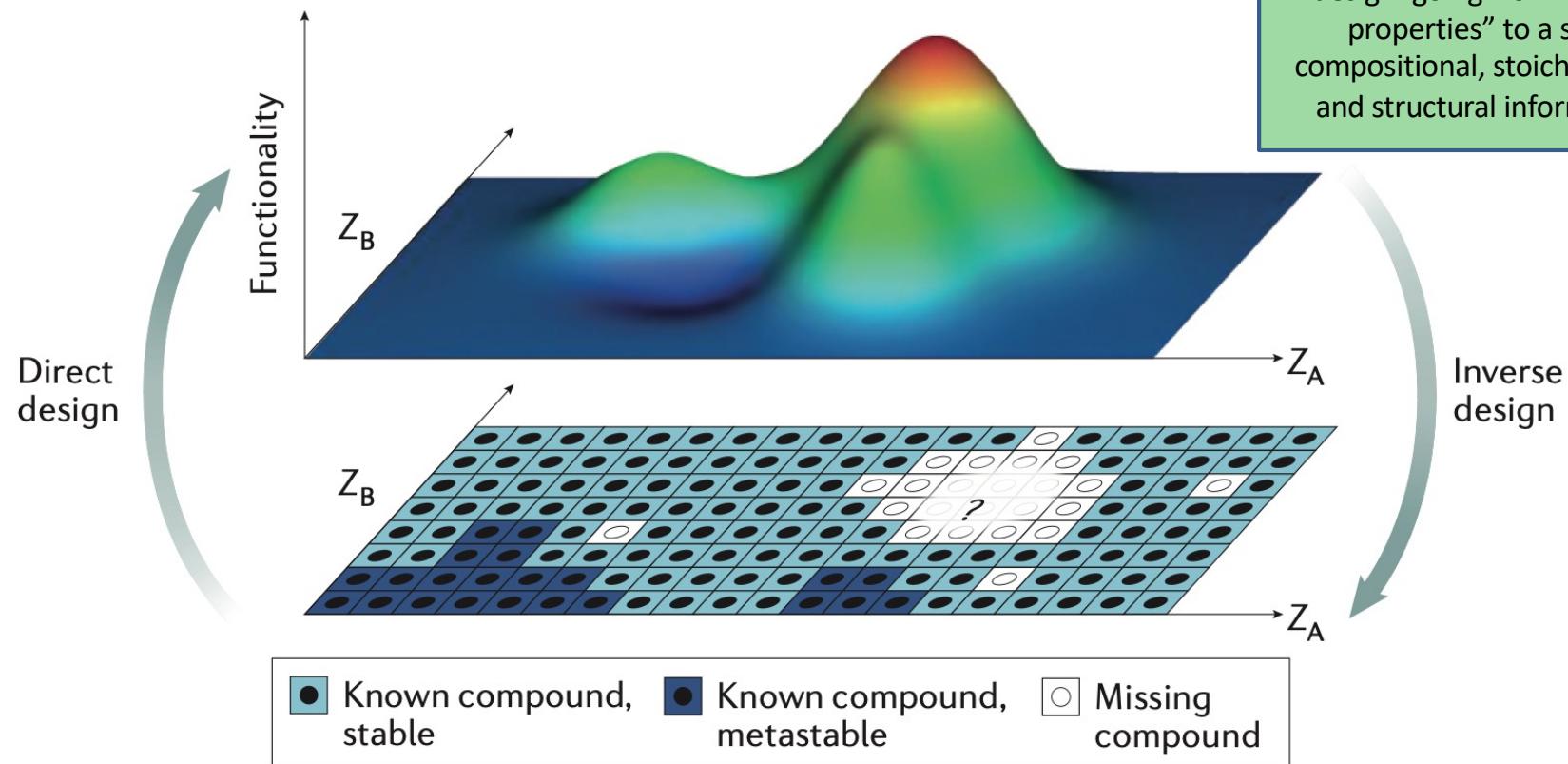
Goal:

**You give me a set of properties,
I give you a material.**

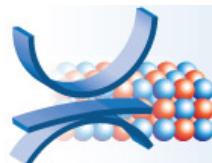
This is the goal of inverse design

Materials Prediction from Property Sets

Alex Zunger's framing of inverse design: going from "desired properties" to a set of compositional, stoichiometric, and structural information.



History: US-Based Research Centers



Center for Inverse Design

Achieving the grand challenge of materials and nanostructures by design



These EFRCs (Energy Frontier Research Centers), led out of Colorado, focused heavily on inverse design. The latter, included the Materials Project team and the presenter.

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PERSPECTIVES

Inverse design in search of materials with target functionalities

Alex Zunger

Abstract | Solid-state chemists have been consistently successful in envisioning and making new compounds, often enlisting the tools of theoretical solid-state physics to explain some of the observed properties of the new materials. Here, a new style of collaboration between theory and experiment is discussed, whereby the desired functionality of the new material is declared first and theoretical

Because properties live in certain ACSs and not in others, the fundamental question in solid-state chemistry, condensed-matter physics, metallurgy and organic matter is: can we uncover the ‘genetic code’ of structure–property (or body–soul) relations?

The properties required to realize a particular device are often known, but the specific materials that harbour such properties are generally unknown and are difficult to identify. The conditions needed for water splitting, the defining qualities of coexisting transparency and conductivity in transparent conductors, and the reason

A. Zunger, *Nat. Rev. Chem.* **2**, 121 (2018)
<https://doi.org/10.1038/s41570-018-0121>

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PERSPECTIVES

Invers with t

Challenge #1: “Technological applications often demand compounds with what would appear to be a counterintuitive combination of two or more properties. Such compounds harbouring what appears as contradicting attributes are rather difficult to identify and discover.”

Alex Zunger

Abstract | Solid-state chemists have been consistently successful in envisioning and making new compounds, often enlisting the tools of theoretical solid-state physics to explain some of the observed properties of the new materials. Here, a new style of collaboration between theory and experiment is discussed, whereby the desired functionality of the new material is declared first and theoretical

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Alex Zunger

Challenge #2: “Because the effective interactions used as input are generally not expressible (or mappable) in terms of the periodic table, there is generally no recognizable rational path between model Hamiltonian predictions and the realization of actual materials. In other words, the theory is not invertible.”

Abstract | Solid-state chemists have been consistently successful in envisioning and making new compounds, often enlisting the tools of theoretical solid-state physics to explain some of the observed properties of the new materials. Here, a new style of collaboration between theory and experiment is discussed, whereby the desired functionality of the new material is declared first and theoretical

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Alex Zunger

Challenge #3: “because the total electron and ion energy of the system is generally not computed or optimized with respect to the structural degrees of freedom, it is difficult to assess whether the predicted properties (such as exotic superconductivity and quantum spin liquids) will be found in stable (or nearly stable) recognizable compounds”

Abstract | Solid-state chemists have been consistently successful in envisioning and making new compounds, often enlisting the tools of theoretical solid-state physics to explain some of the observed properties of the new materials. Here, a new style of collaboration between theory and experiment is discussed, whereby the desired functionality of the new material is declared first and theoretical

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A. Zunger, *Nat. Rev. Chem.* **2**, 121 (2018)
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Generative Design & Inverse Design

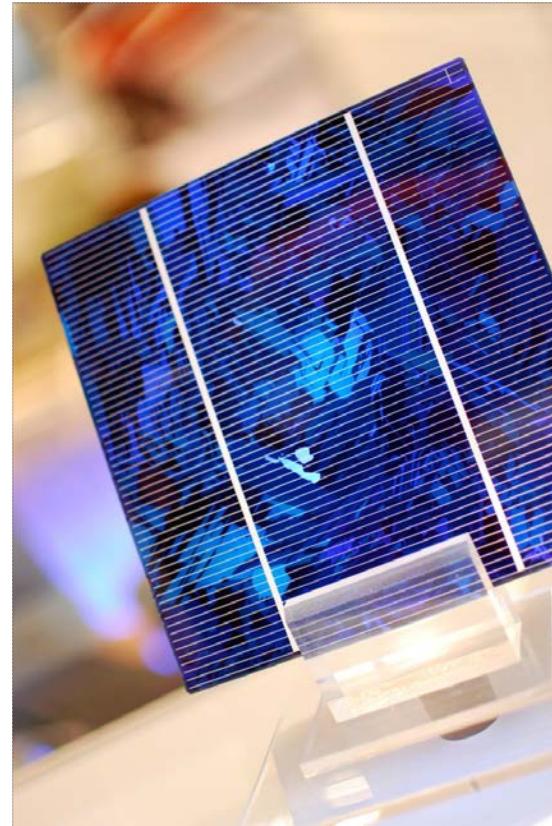
As our community struggles for common definitions, here are my definitions of these two terms with examples.

Generative Design

“Generative design is an iterative design process that involves a program that will generate a certain number of outputs that meet certain constraints, and a designer that will fine tune the feasible region by selecting specific output or changing input values, ranges and distribution.”

https://en.wikipedia.org/wiki/Generative_design

2014: Generative Design in Silicon Solar Cell Manufacturing

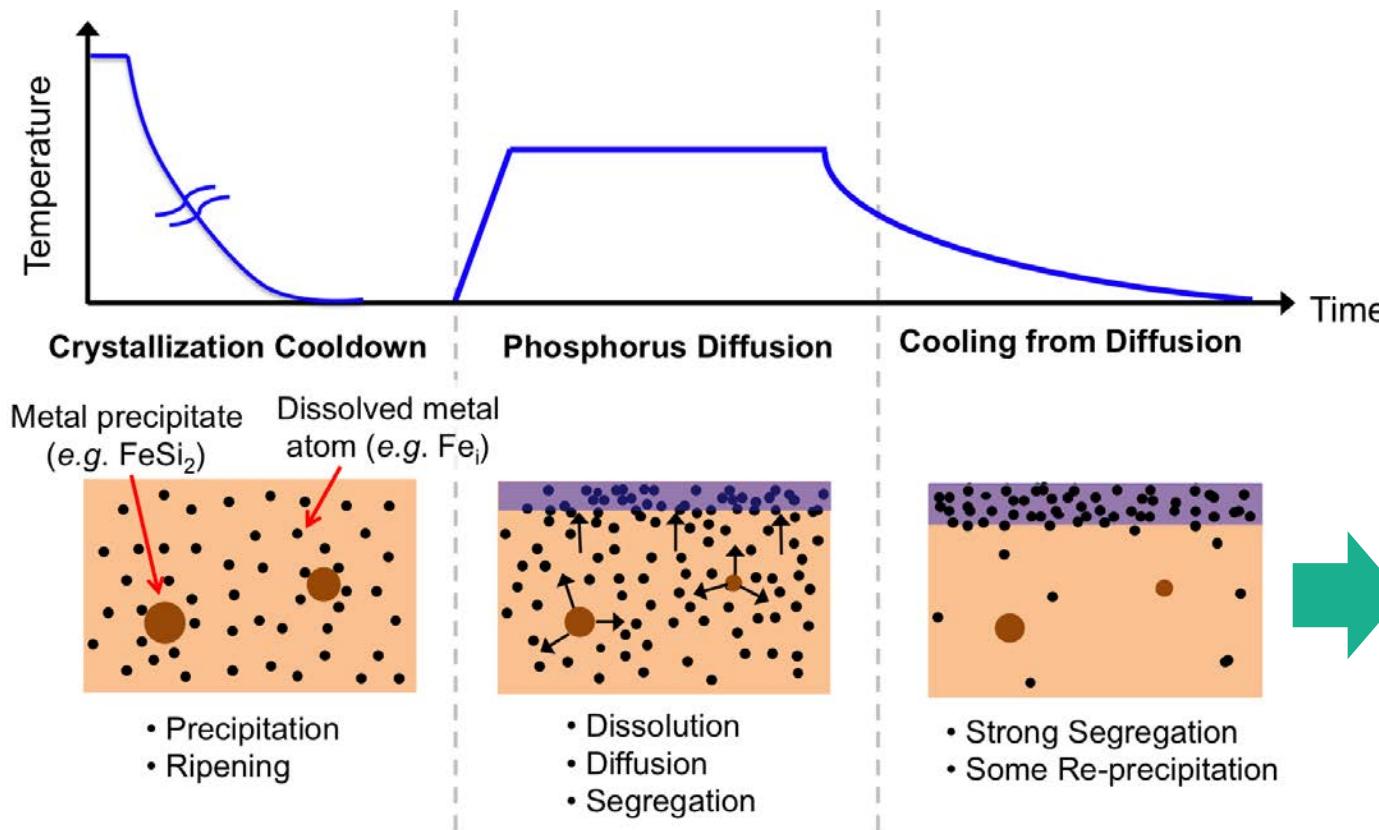


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2014: Generative Design in Silicon Solar Cell Manufacturing



David Fenning
UCSD
[@FRG_UCSD](https://twitter.com/@FRG_UCSD)



carrier lifetime → PC1D → solar cell efficiency

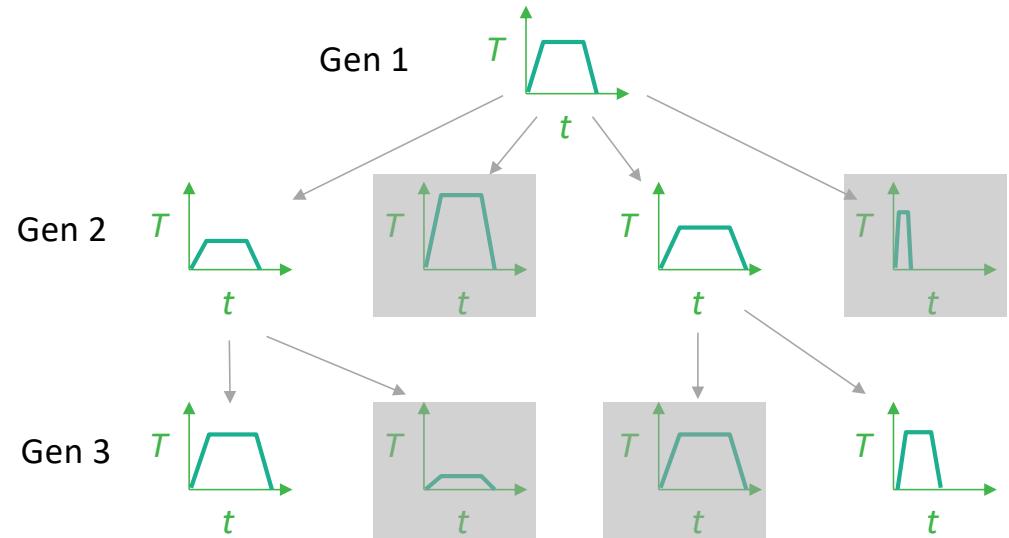
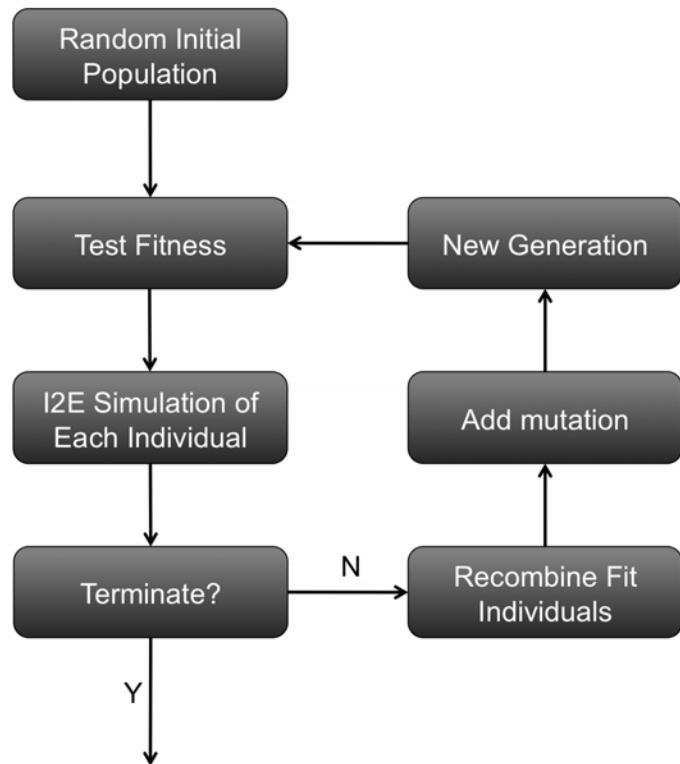
The I2E, or “Impurities to Efficiency” simulator, predicts Si solar cell efficiency from impurities in the initial material + process conditions

D.P. Fenning et al., *Adv. Energy Mater.* 1400459 (2014)

<https://doi.org/10.1002/aenm.201400459>

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2014: Generative Design in Silicon Solar Cell Manufacturing



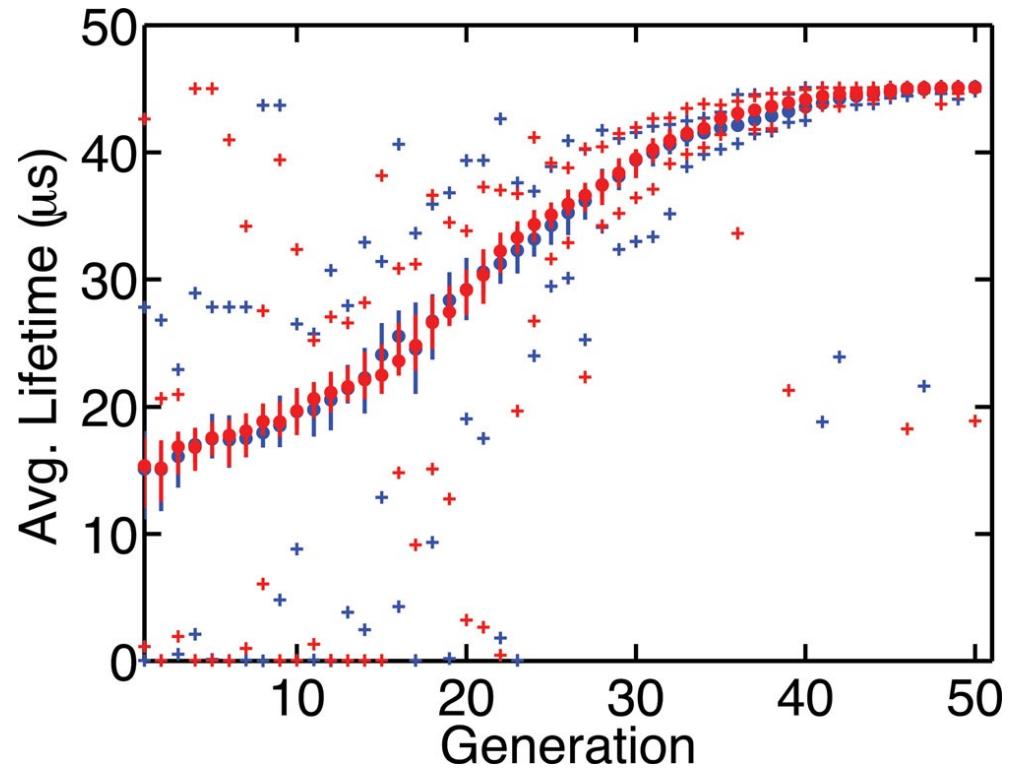
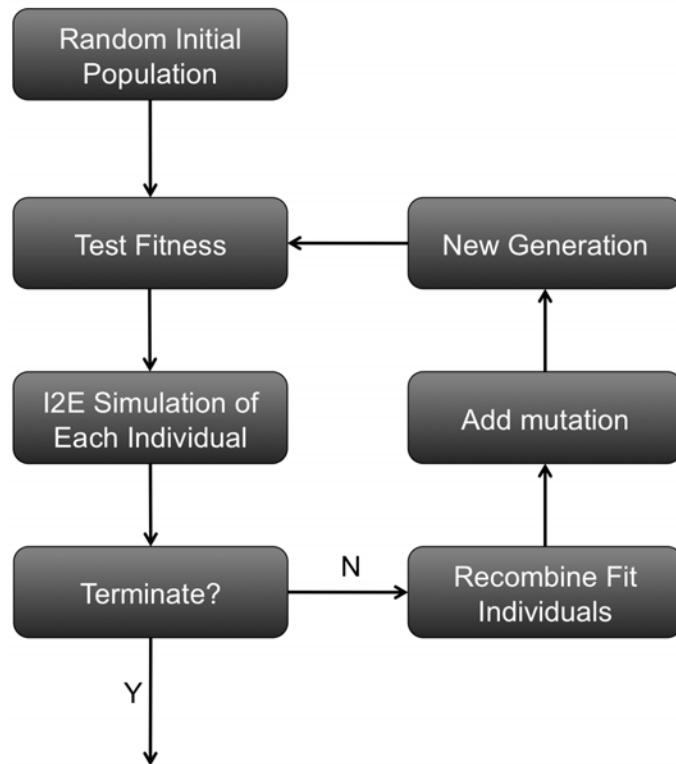
The I2E tool can be wrapped in a Genetic Algorithm, creating new time-temperature profiles from an original time-temperature profile.

D.P. Fenning et al., *Adv. Energy Mater.* 1400459 (2014)

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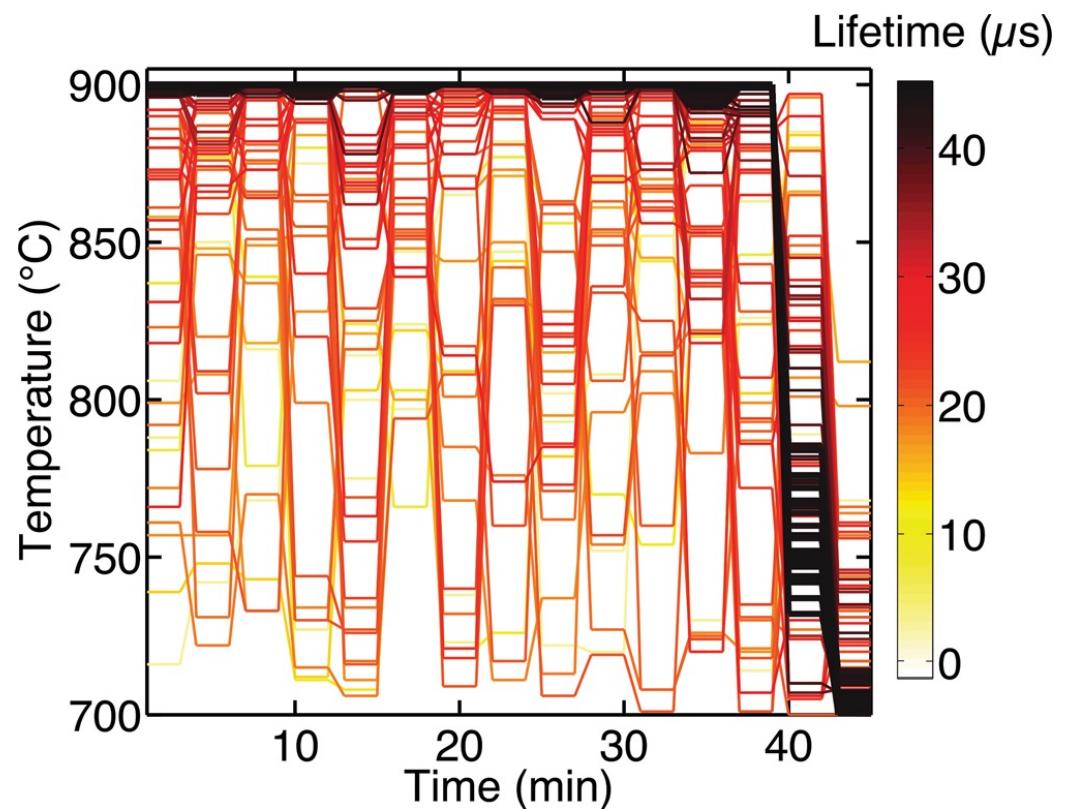
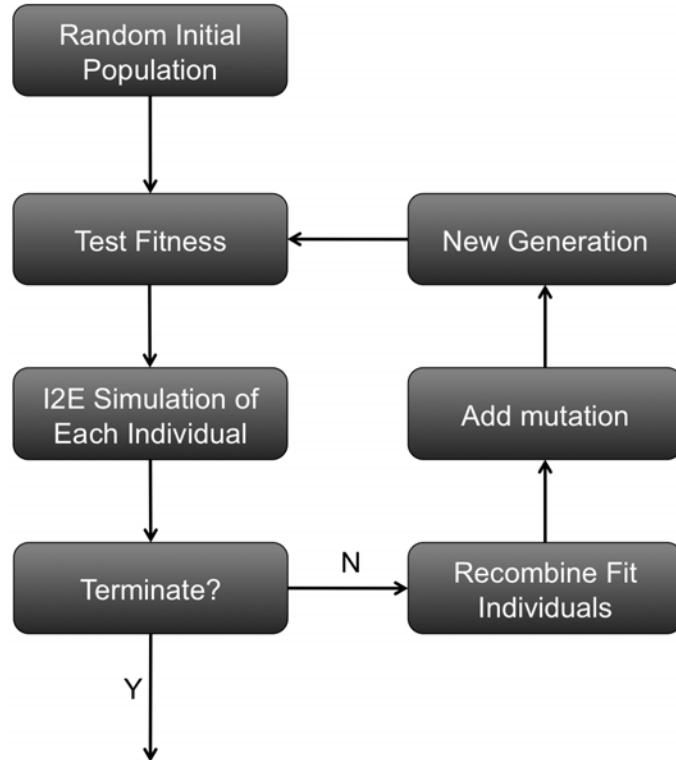


D.P. Fenning et al., *Adv. Energy Mater.* 1400459 (2014)

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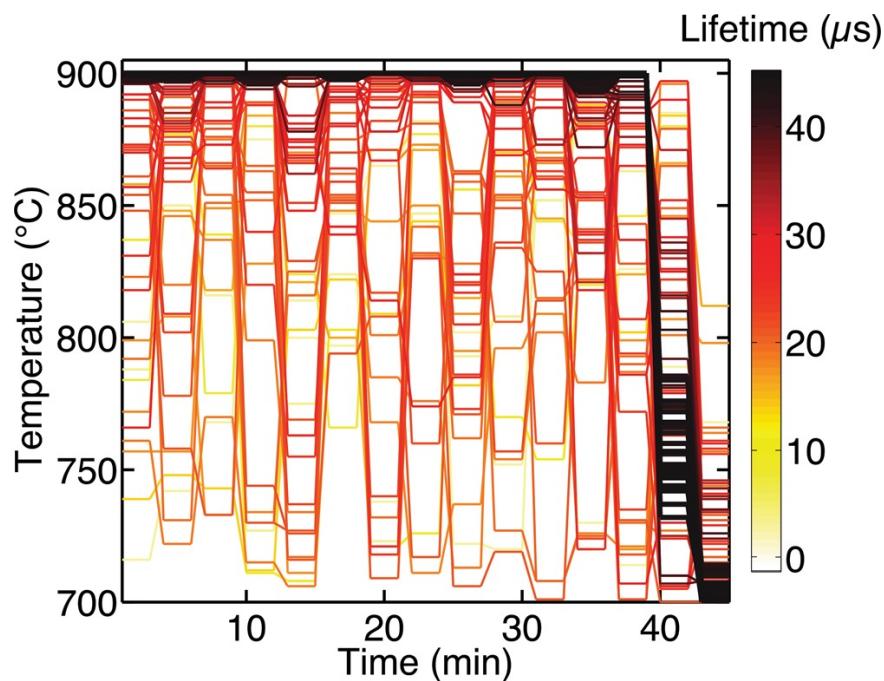
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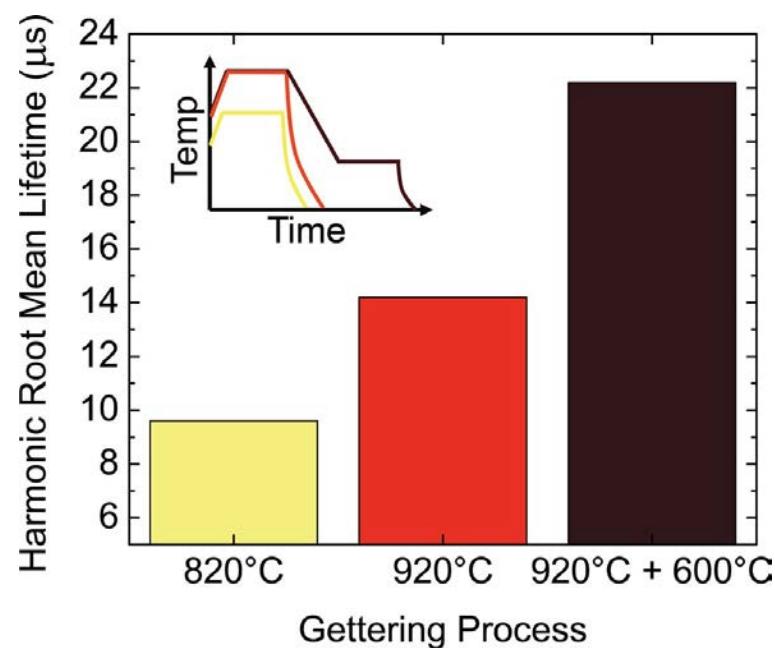
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2014: Generative Design in Silicon Solar Cell Manufacturing

Prediction



Experiment

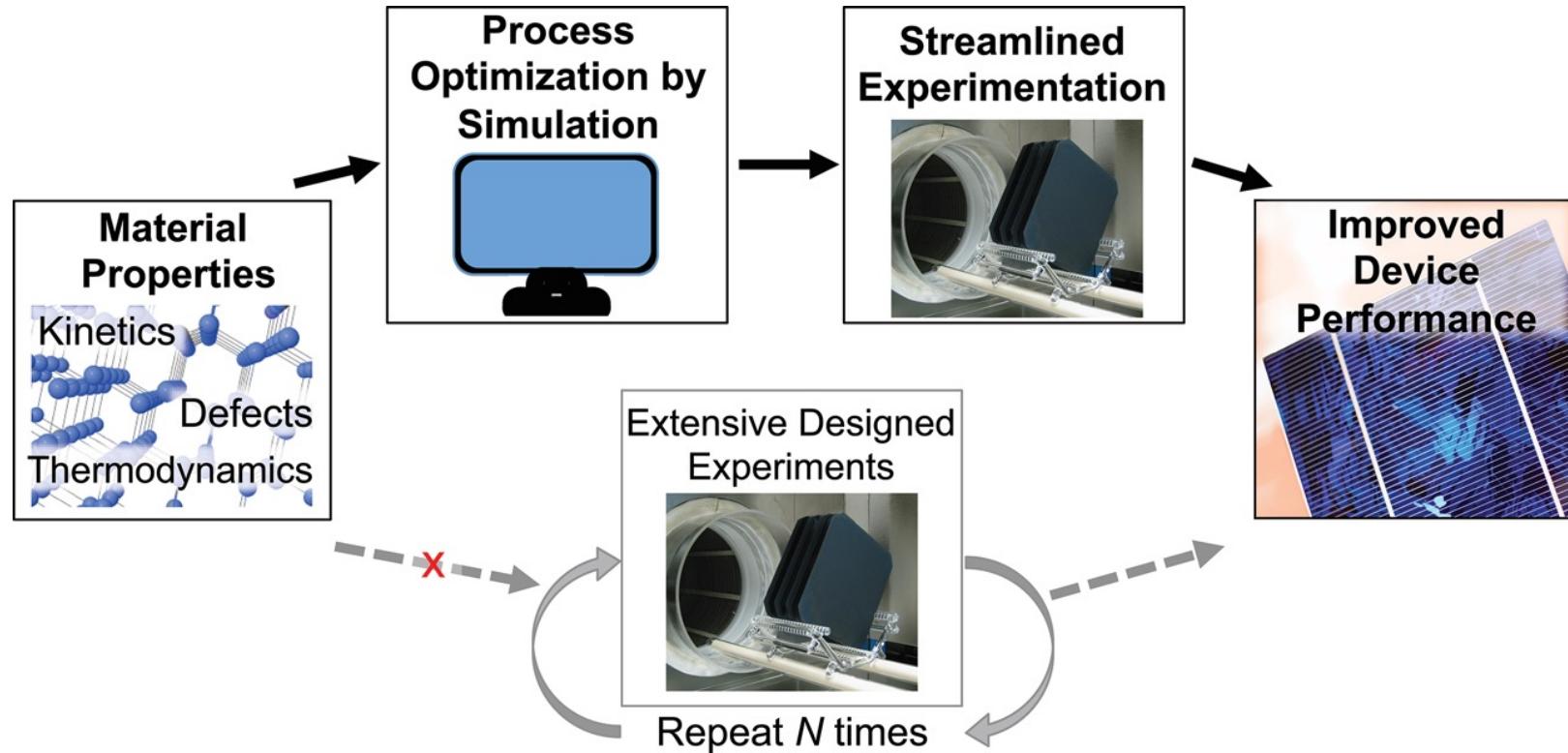


D.P. Fenning *et al.*, *Adv. Energy Mater.* 1400459 (2014)

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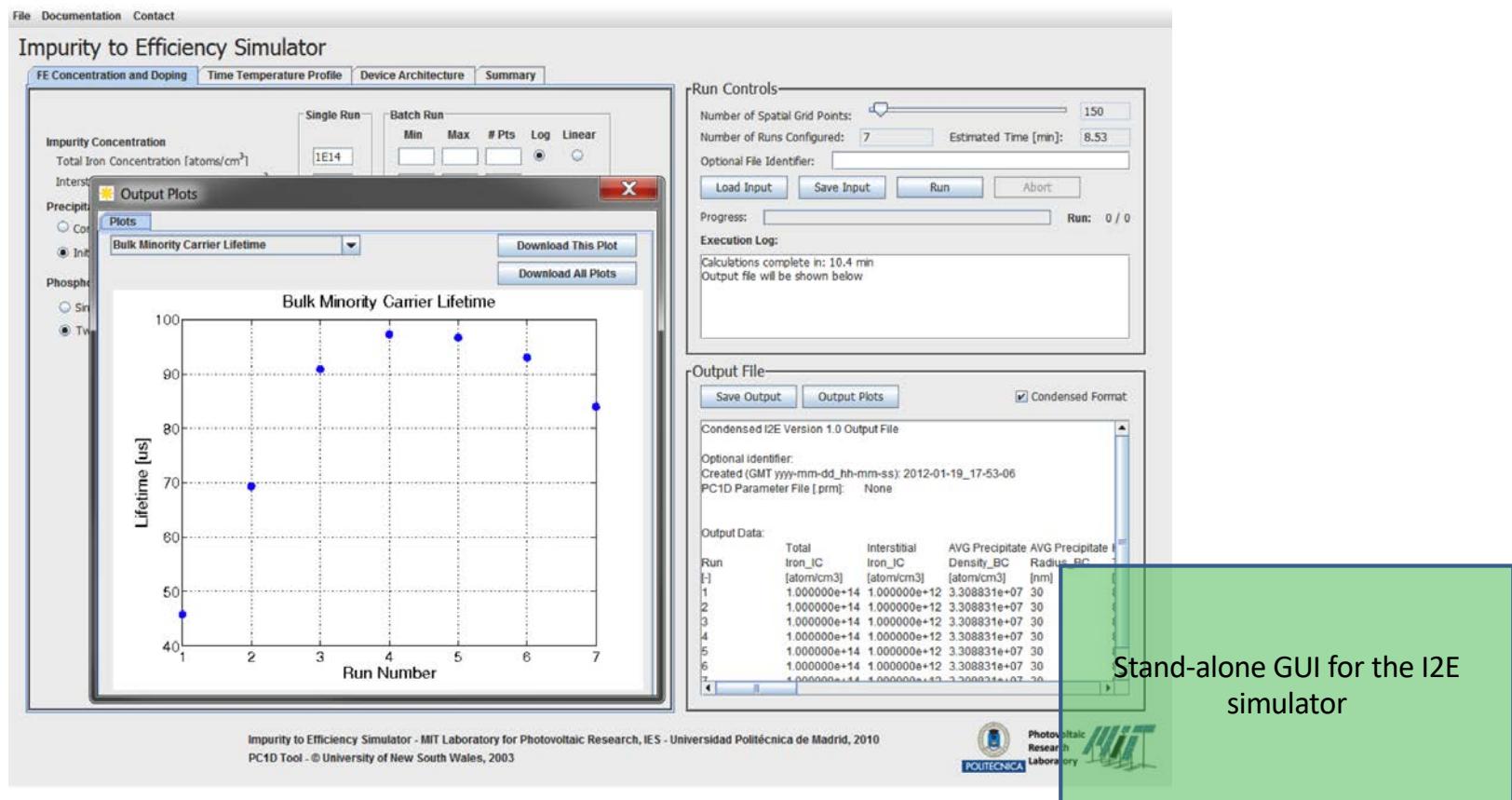


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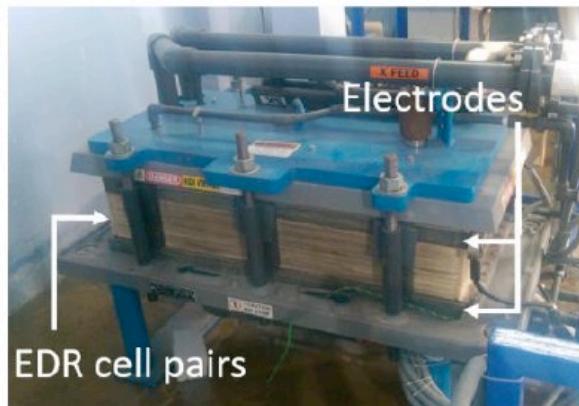
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2017: Solar-Powered Desalination System Optimization



Sterling Watson
MIT; Natel Energy



Another example of generative design in sustainability, a collaboration with the GEAR Lab at MIT (Amos Winter)

S. Watson *et al.*, *Solar Energy* **162**, 132–139 (2018)

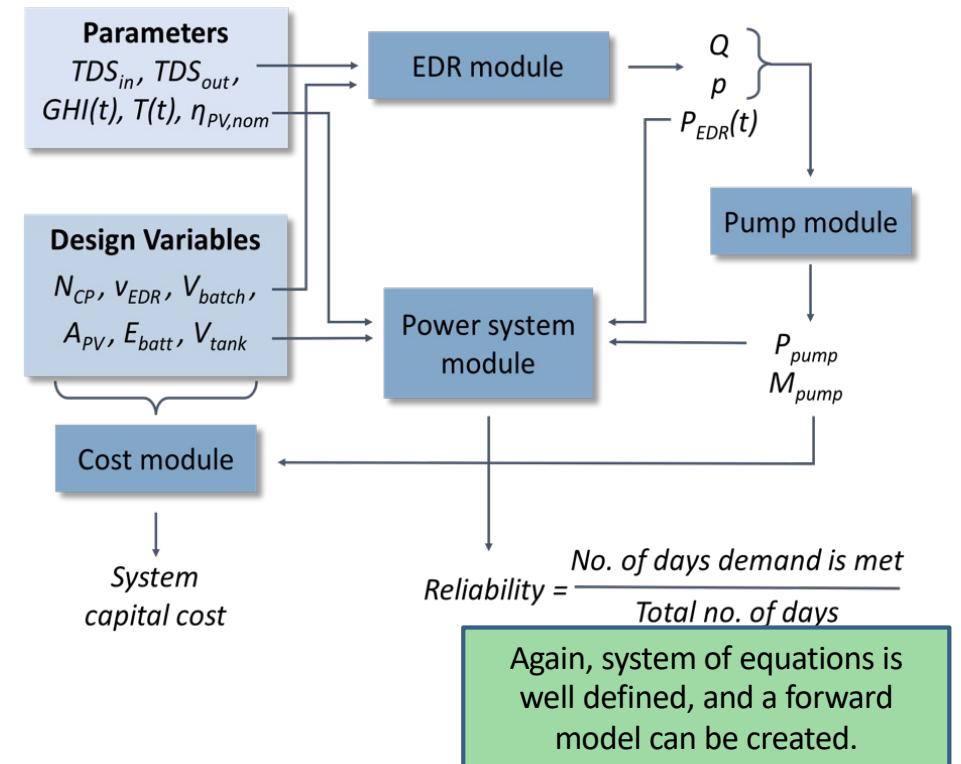
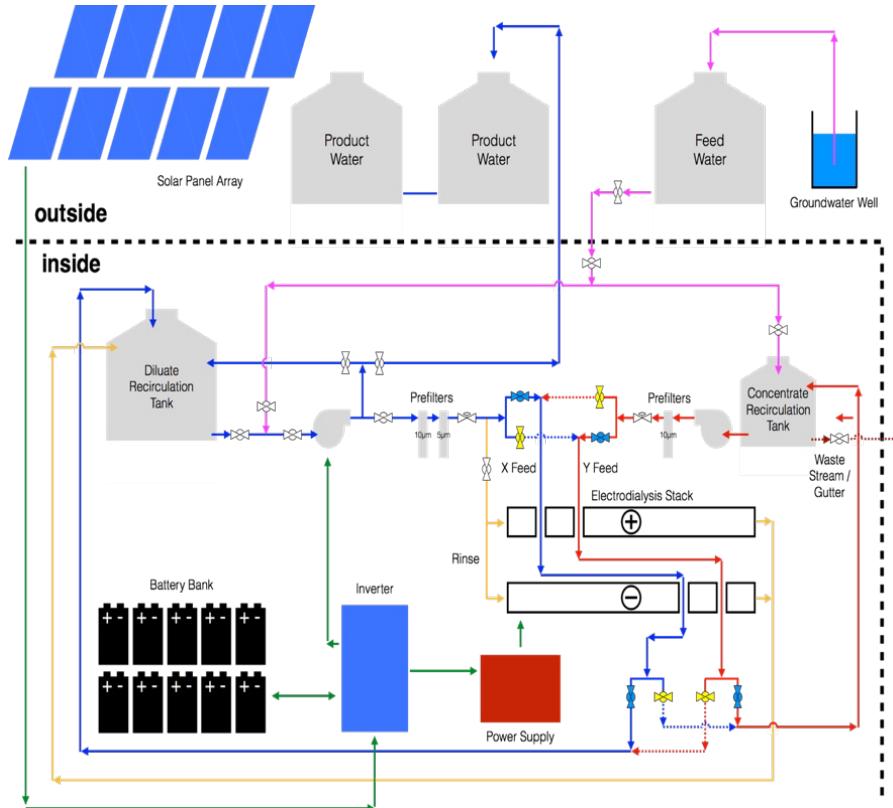
<https://doi.org/10.1016/j.desal.2019.114217>

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2017: Solar-Powered Desalination System Optimization



S. Watson *et al.*, *Solar Energy* **162**, 132–139 (2018)

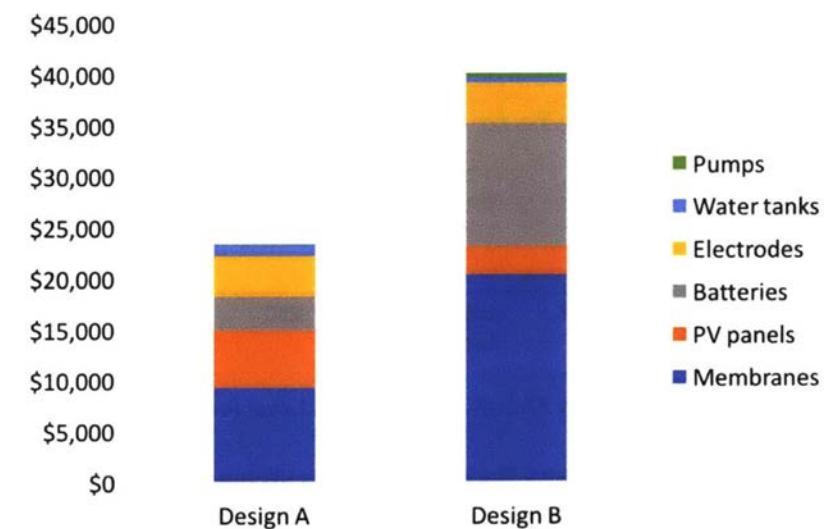
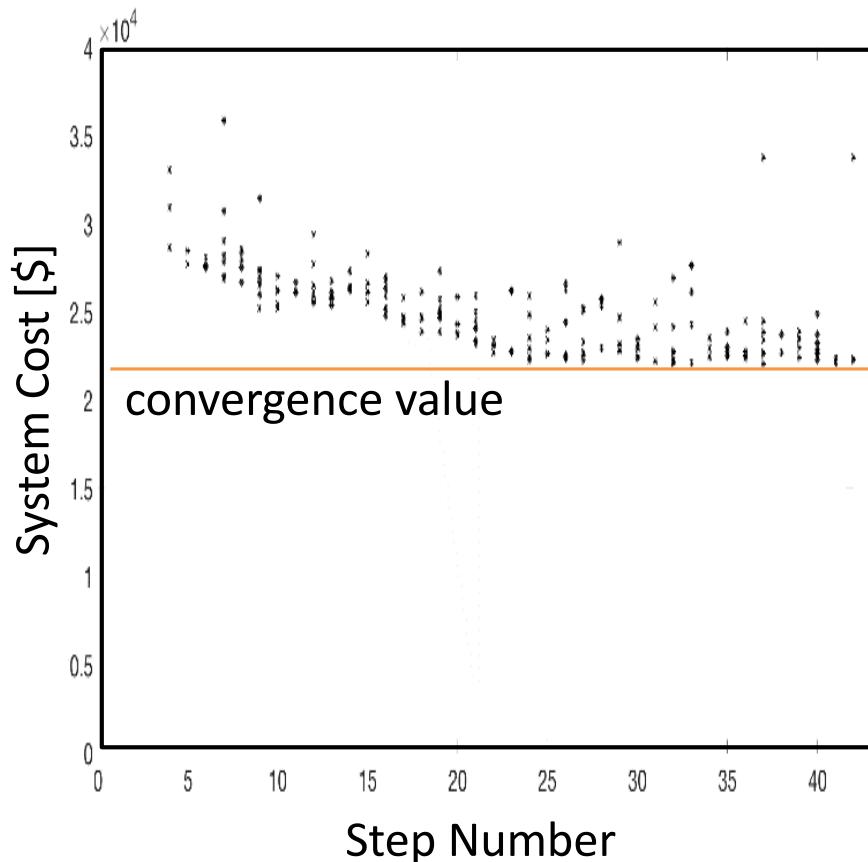
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2017: Solar-Powered Desalination System Optimization



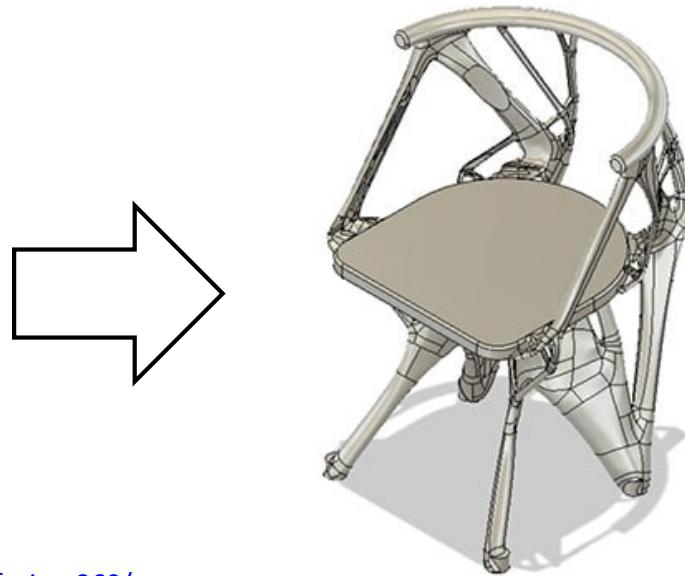
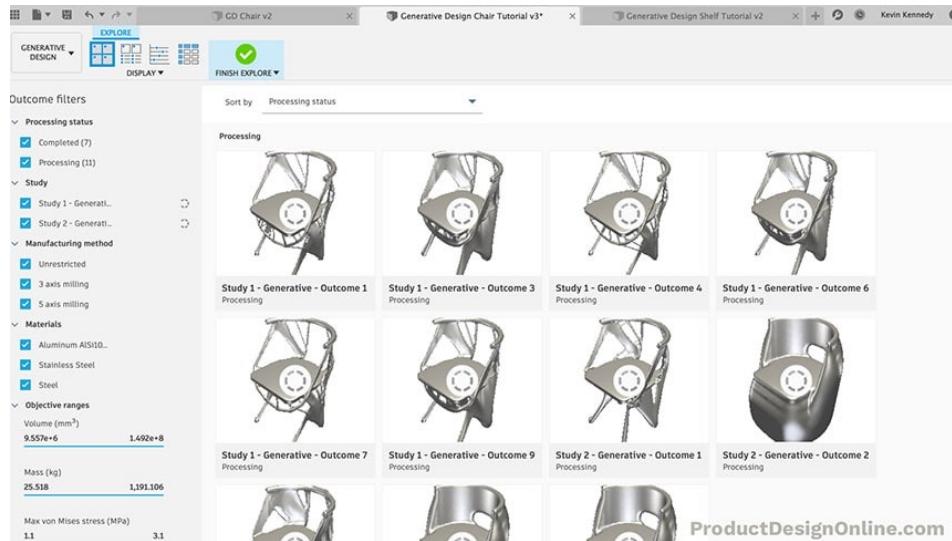
A nearly 40% drop in system cost can be achieved through particle swarm optimization, with this forward model.

2017: Solar-Powered Desalination System Optimization



Designed and built a cost-optimized PV-EDR system in Chellur, India

AutoCAD - Fusion 360: Constrained Generative Design

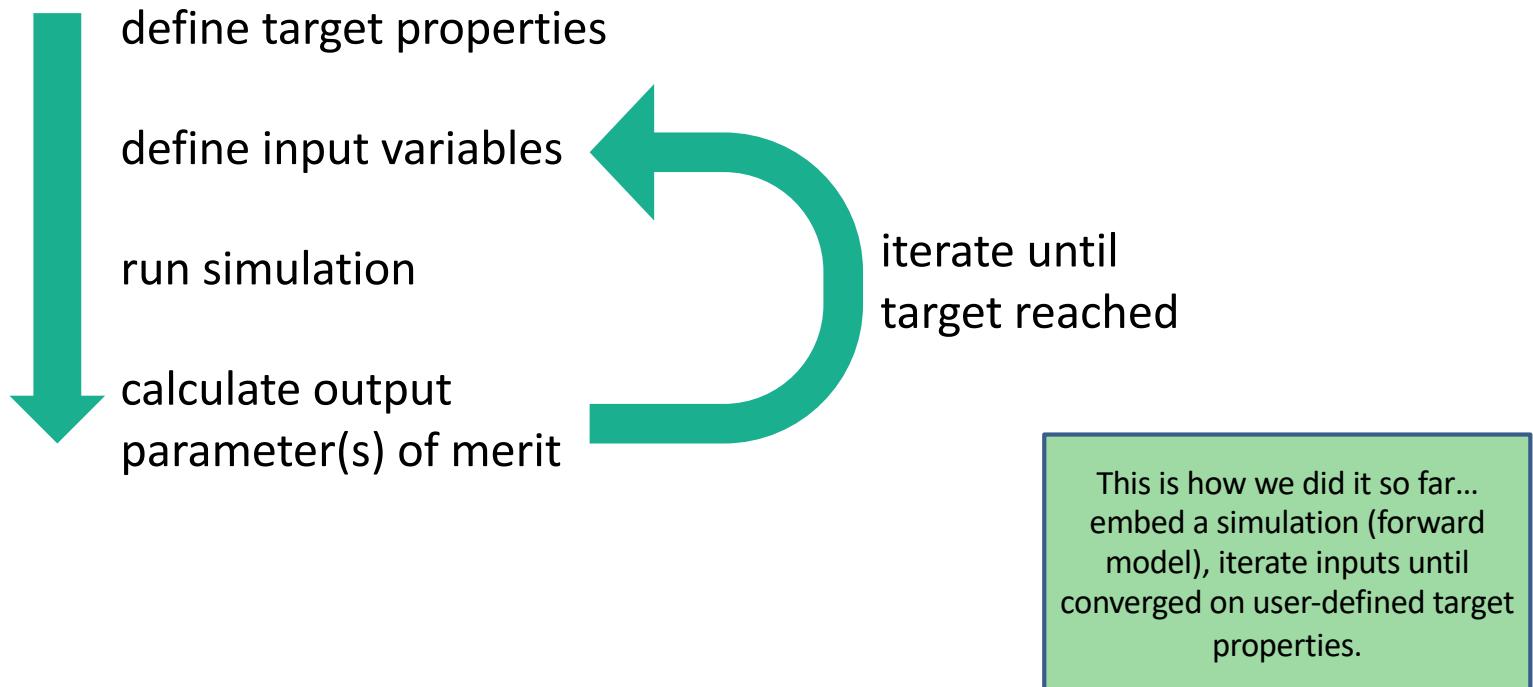


<https://productdesignonline.com/fusion-360-tutorials/generative-design-a-stronger-chair-in-fusion-360/>

A commercial example

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Generative Design Approach

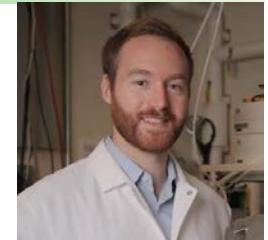


Generative Design Approach

- 
- define target properties
 - define input variables
 - search database
 - identify materials with target properties

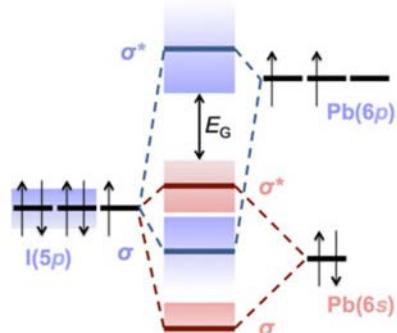
With Materials Project and other open-source databases, we could now search those databases for materials with user-defined properties.

2015: Screening Materials Project with “Transport” Proxies

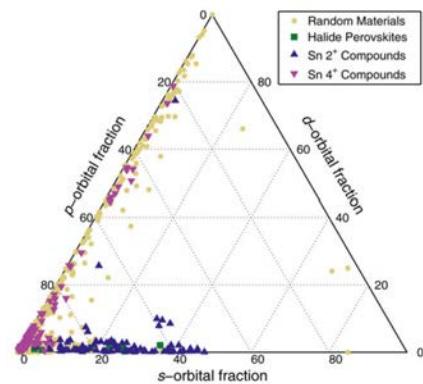


Riley Brandt

Determine Params of Merit



Screen Params of Merit



R.E. Brandt *et al.*, MRS Communications 5, 265 (2015)

<http://dx.doi.org/10.1557/mrc.2015.26>

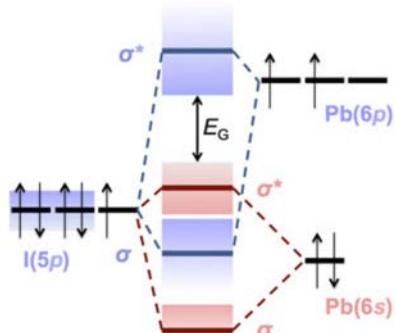
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2015: Screening Materials Project with “Transport” Proxies

Determine Params of Merit



Screen Params of Merit

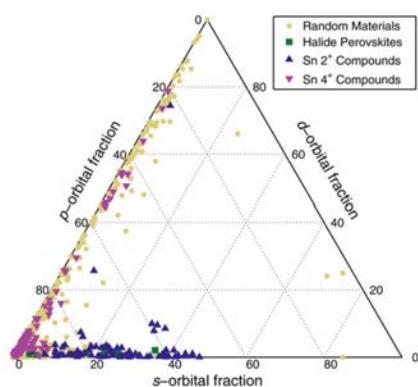


Table I. DFT-computed band structures, band dispersion (with SO coupling), effective masses, and ionic dielectric constants for several compounds of interest, as compared with the (MA)BX₃ family of perovskites.

Formula	Space Group (International S. G. number)	E_G (eV) (DFT+SO)	E_G direct (eV) (DFT + SO)	E_G (DFT + SO) — E_G (DFT)	ϵ_V (DFT + SO) — ϵ_V (DFT)	E_C (DFT + SO) — E_C (DFT)	m_e^* (DOS)	m_h^* (DOS)	VBM band degeneracy	CBM band degeneracy	Ionic dielectric constant
InI	Cmcm (63)	1.29	1.29	-0.09	0.08	-0.01	0.25	0.15	2	2	33.51
SnI ₂	C12/m1 (12)	1.58	1.74	-0.18	0.10	-0.08	1.49	0.35	4	2	28.84
SbI ₂	R3H (148)	1.88	1.96	-0.26	0.18	-0.08	9.57	2.12	1	1	8.32
SbI ₃	P12/c1 (14)	1.86	1.93	-0.25	0.18	-0.07	8.88	1.72	2	1	8.43
BiI ₃	R3H (148)	1.73	1.82	-0.78	0.23	-0.55	10.39	1.85	1	1	5.70
BiI ₃	P3 1m (162)	1.66	1.72	-0.81	0.23	-0.57	9.34	0.79	1	2	5.78
Bi ₂ S ₃	Pnma (62)	1.28	1.30	-0.02	-0.01	-0.03	3.89	0.88	2	3	44.94
Bi ₂ S ₃	Pnma (62)	1.14	1.14	-0.24	-0.17	-0.41	2.86	0.49	2	1	400.33
BiOI	P4/nmmS (129)	1.38	1.49	-0.13	0.00	-0.14	3.75	0.37	9	1	46.32
BiSI	Pnam (62)	1.18	1.32	-0.69	0.05	-0.64	4.79	0.53	2	1	29.59
BiSel	Pnma (62)	0.91	1.03	-0.68	0.04	-0.65	5.89	0.25	2	1	26.83
BiSBr	Pnma (62)	1.32	1.35	-0.53	-0.11	-0.64	6.21	0.24	2	1	30.10
St ₂ Si	P2 ₁ 2 ₁ c (19)	1.28	1.46	-0.23	0.11	-0.11	2.84	0.91	2	2	31.59
St ₂ Si	Pnma ₂ (33)	1.45	1.60	-0.17	0.11	-0.06	2.06	1.31	2	2	69.72
St ₂ Si	Pnma (62)	1.45	1.60	-0.17	0.11	-0.06	2.06	1.25	4	2	69.38
St ₂ Si	Pnma (62)	1.16	1.29	-0.20	0.10	-0.10	4.37	0.59	2	2	43.94
RbBiS ₂	R3mH (166)	1.12	1.47	-0.22	-0.04	-0.27	10.96	0.20	1	2	37.94
InAlI ₄	P12 ₁ /m1 (11)	2.87	2.87	-0.15	0.12	-0.03	1.49	3.90	2	2	19.59
CaInBr ₃	Cmcm (63)	3.12	3.28	-0.04	0.02	-0.02	1.43	0.27	1	2	69.26
KSnS ₂	P6 ₃ mc (166)	0.18	0.38	-0.08	0.12	0.04	0.25	0.04	0.5	4	2.92
Cs ₂ Sn ₂ I ₉	P3m1 (164)	1.40	1.41	-0.14	0.20	0.05	2.19	0.25	1	7	13.06
Cs ₂ Sn ₂ I ₉	P6 ₃ /mmc (194)	1.79	1.91	-0.10	0.16	0.06	3.73	0.50	2	2	9.68
Cs ₂ Bi ₂ I ₉	P6 ₃ /mmc (194)	1.90	2.04	-0.46	0.24	-0.22	4.63	1.79	2	3	9.63
(MA)GeI ₃	Pm3m (221)	0.82	0.82	-0.18	0.11	-0.06	0.43	0.51	2	2	6.89
(MA)SnI ₃	Pm3m (221)	0.12	0.12	-0.29	0.05	-0.24	0.12	0.65	1	1	25.57
(MA)PbI ₃	Pm3m (221)	0.31	0.31	-1.12	0.13	-1.00	0.10	0.16	1	1	20.07

R.E. Brandt *et al.*, MRS Communications 5, 265 (2015)

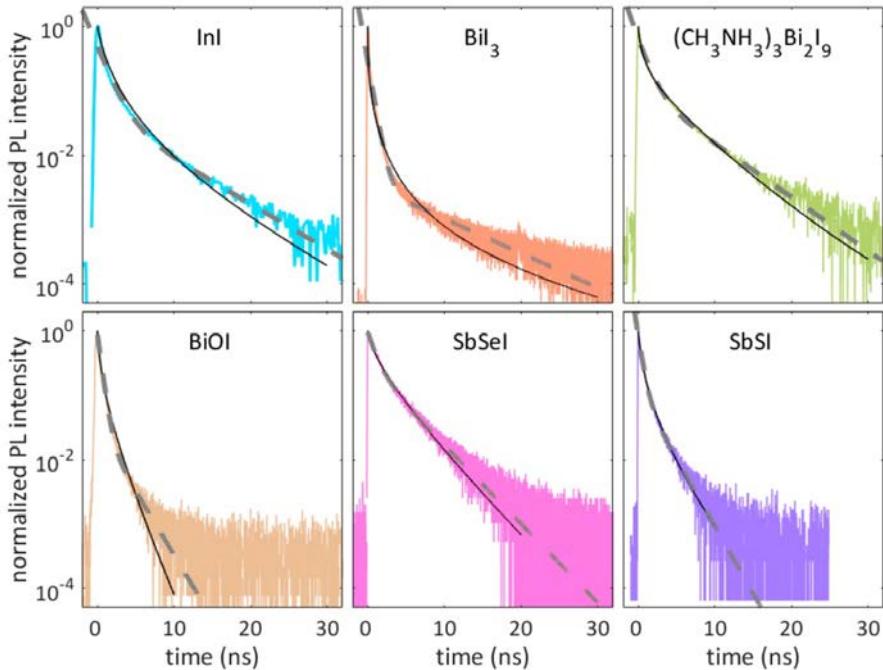
<http://dx.doi.org/10.1557/mrc.2015.26>

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2016: “High-Throughput Experiment”: 1 Material per Month



Article
pubs.acs.org/cm

Searching for “Defect-Tolerant” Photovoltaic Materials: Combined Theoretical and Experimental Screening

Riley E. Brandt,^{*†‡} Jeremy R. Poindexter,[†] Prashun Gorai,^{‡§} Rachel C. Kurchin,[†] Robert L. Z. Hove,^{†,¶} Lea Nienhaus,[‡] Mark W. B. Wilson,^{†,#} J. Alexander Polizzotti,[†] Raimundas Sereika,^{||} Raimundas Žaltauskas,^{||} Lana C. Lee,[†] Judith L. MacManus-Driscoll,[†] Moungi Bawendi,[†] Vladan Stevanović,^{‡§} and Tonio Buonassisi[†]

[†]Massachusetts Institute of Technology, Cambridge, Massachusetts 02139, United States

[‡]Colorado School of Mines, Golden, Colorado 80401, United States

[§]National Renewable Energy Laboratory, Golden, Colorado 80401, United States

^{||}Faculty of Science and Technology, Lithuanian University of Educational Sciences, Vilnius 08106, Lithuania

[#]Department of Materials Science and Metallurgy, University of Cambridge, Cambridge CB3 0FF, United Kingdom

Our “high-throughput” search in
2016–2017... 1 material per
month, validating predictions for
defect-tolerant compounds with
large charge-carrier lifetimes

R.E. Brandt *et al.*, *Chemistry of Materials* **29**, 4667 (2017)

<https://doi.org/10.1021/acs.chemmater.6b05496>

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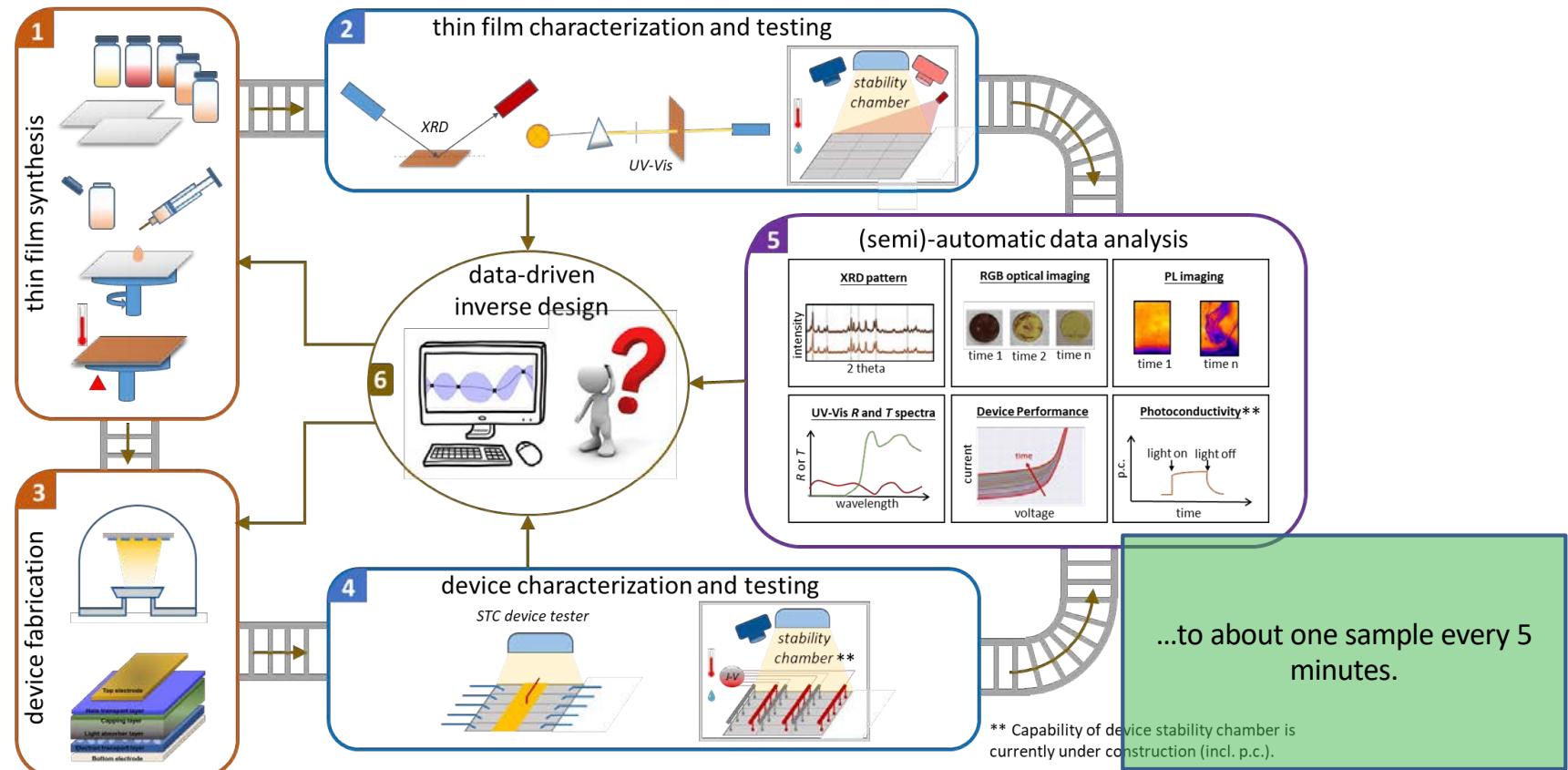
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Generative Design Approach

- 
- define target properties
 - define input variables
 - search chemical space
 - identify materials with target properties

We can increase throughput...

2019: SolTrain: MIT's PV R&D platform



S. Sun et al., Joule 3, 1437 (2019)

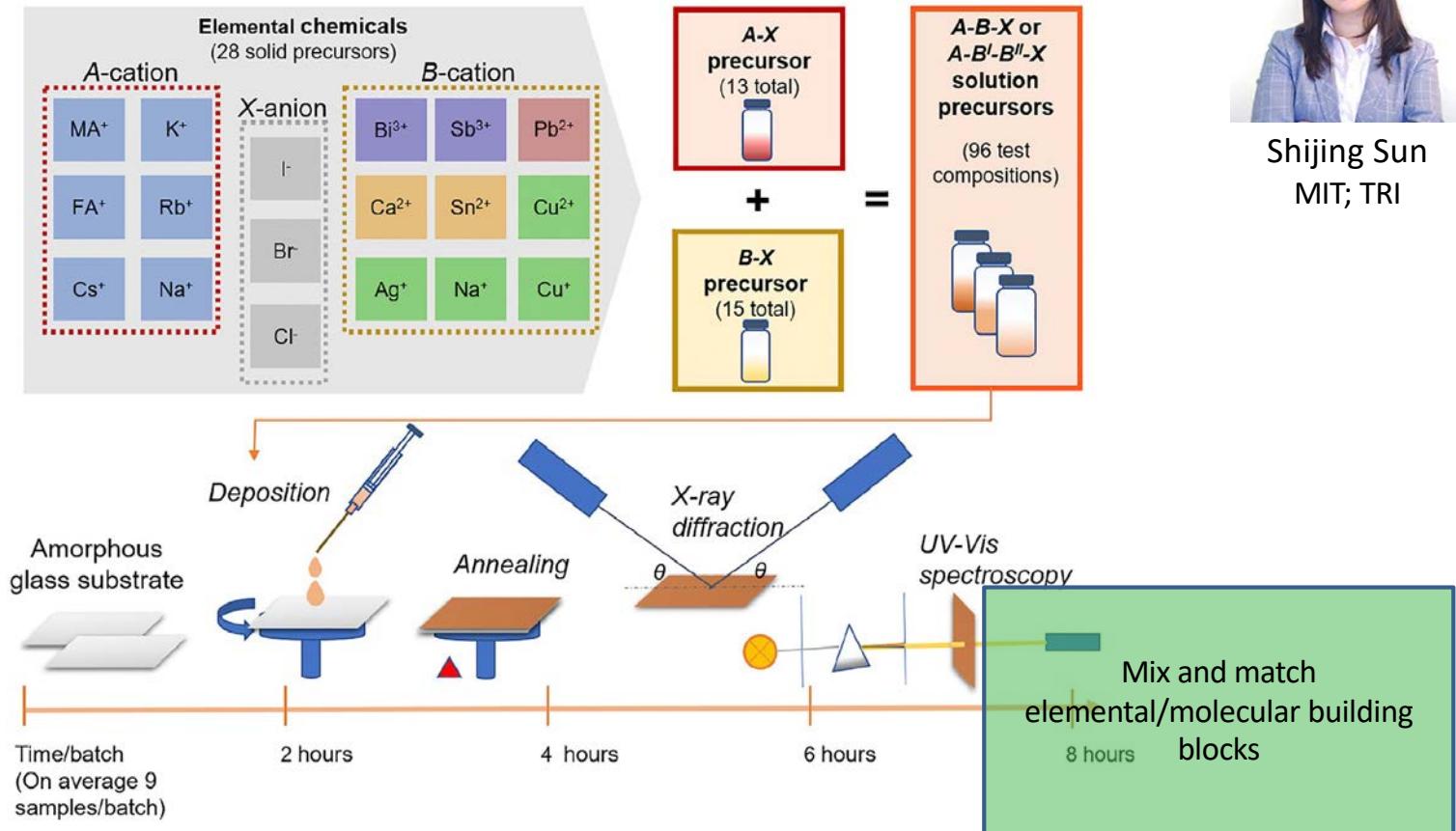
<https://doi.org/10.1016/j.joule.2019.05.014>

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2019: Wide elemental range, single platform

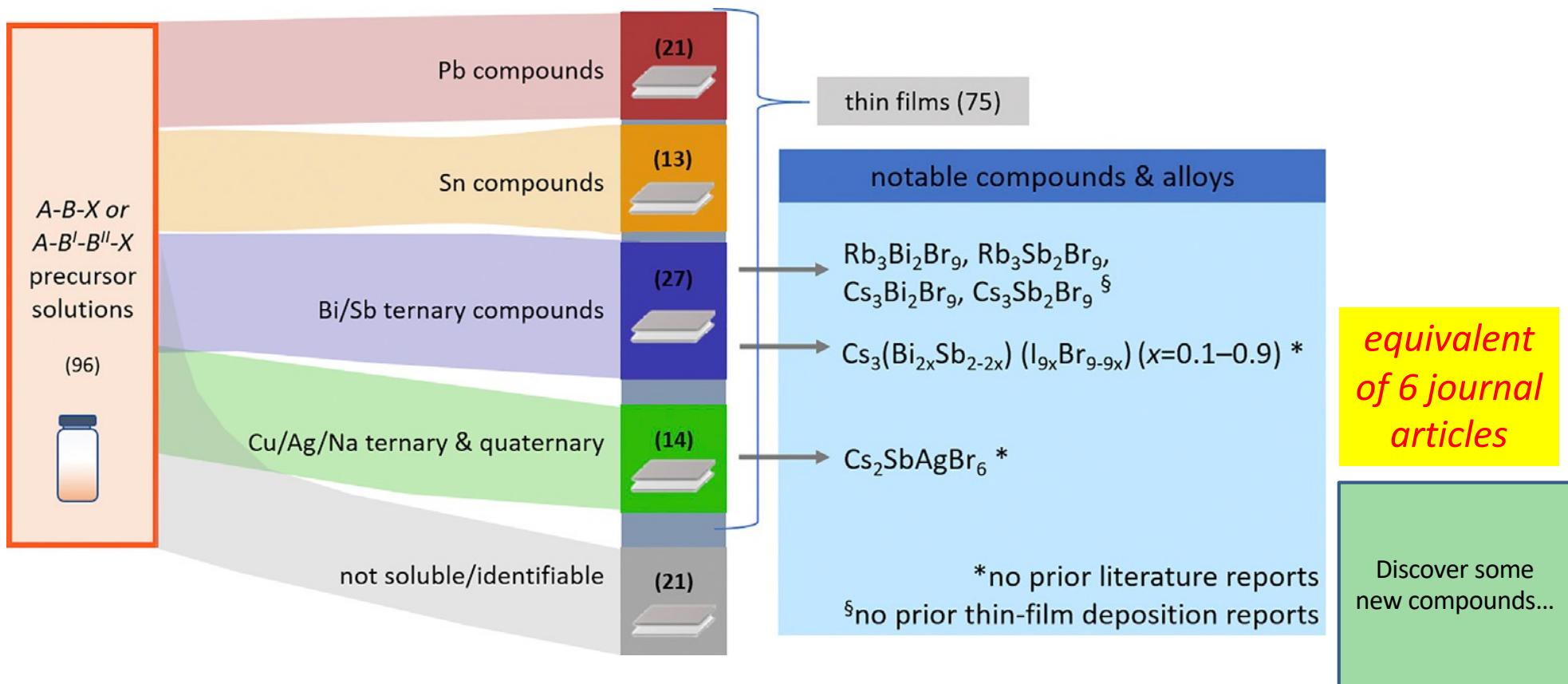


S. Sun et al., Joule 3, 1437 (2019)

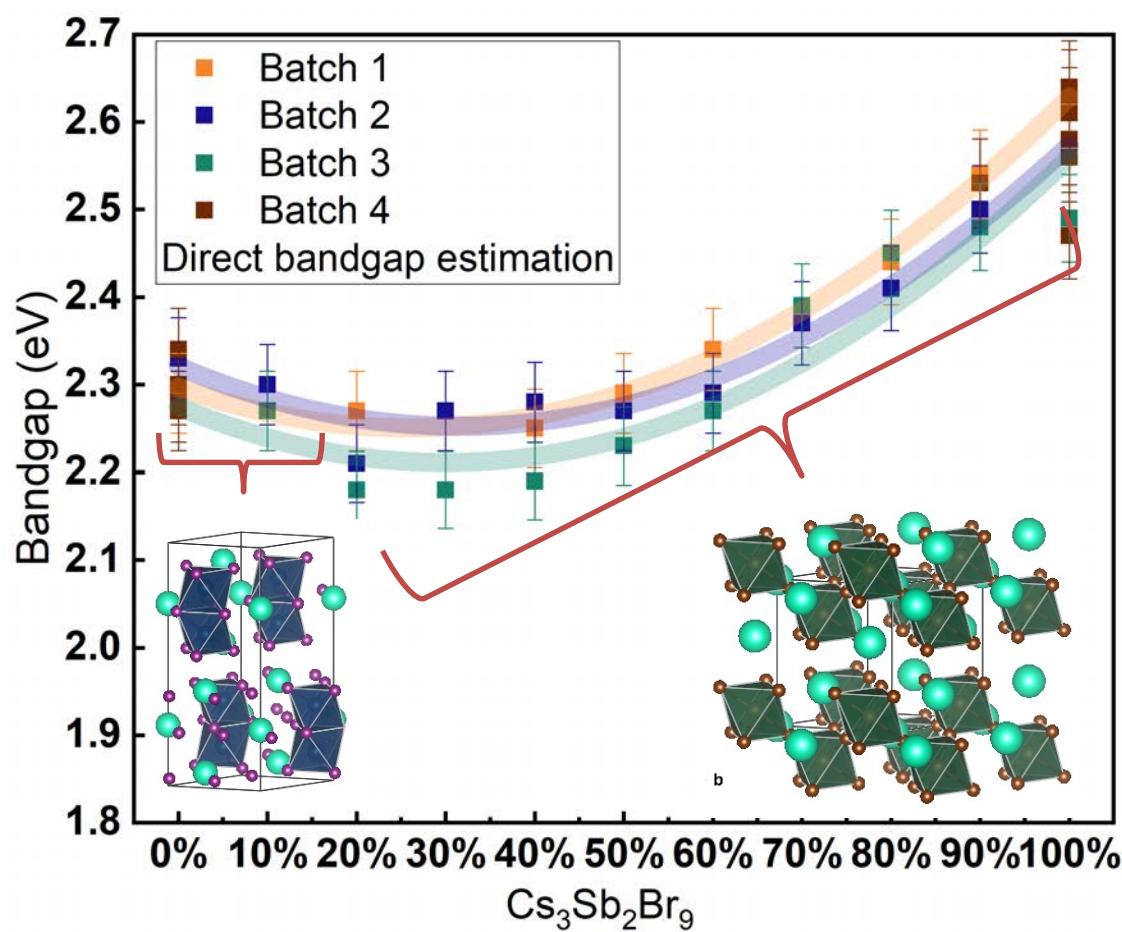
<https://doi.org/10.1016/j.joule.2019.05.014>

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2019: Results from 2-mo Campaign



Bandgap Bowing in $\text{Cs}_3(\text{Bi}_{1-x}\text{Sb}_x)_2(\text{I}_{1-x}\text{Br}_x)_9$



...and some new science

PERSPECTIVES

But databases and brute-force approaches fail, when you have many (contradicting!) materials properties to simultaneously optimize. New, faster approaches are needed...

Invers with t

Alex Zunger

Abstract | Solid-state chemists have been consistently successful in envisioning and making new compounds, often enlisting the tools of theoretical solid-state physics to explain some of the observed properties of the new materials. Here, a new style of collaboration between theory and experiment is discussed, whereby the desired functionality of the new material is declared first and theoretical

Challenge #1: “Technological applications often demand compounds with what would appear to be a counterintuitive combination of two or more properties. Such compounds harbouring what appears as contradicting attributes are rather difficult to identify and discover.”

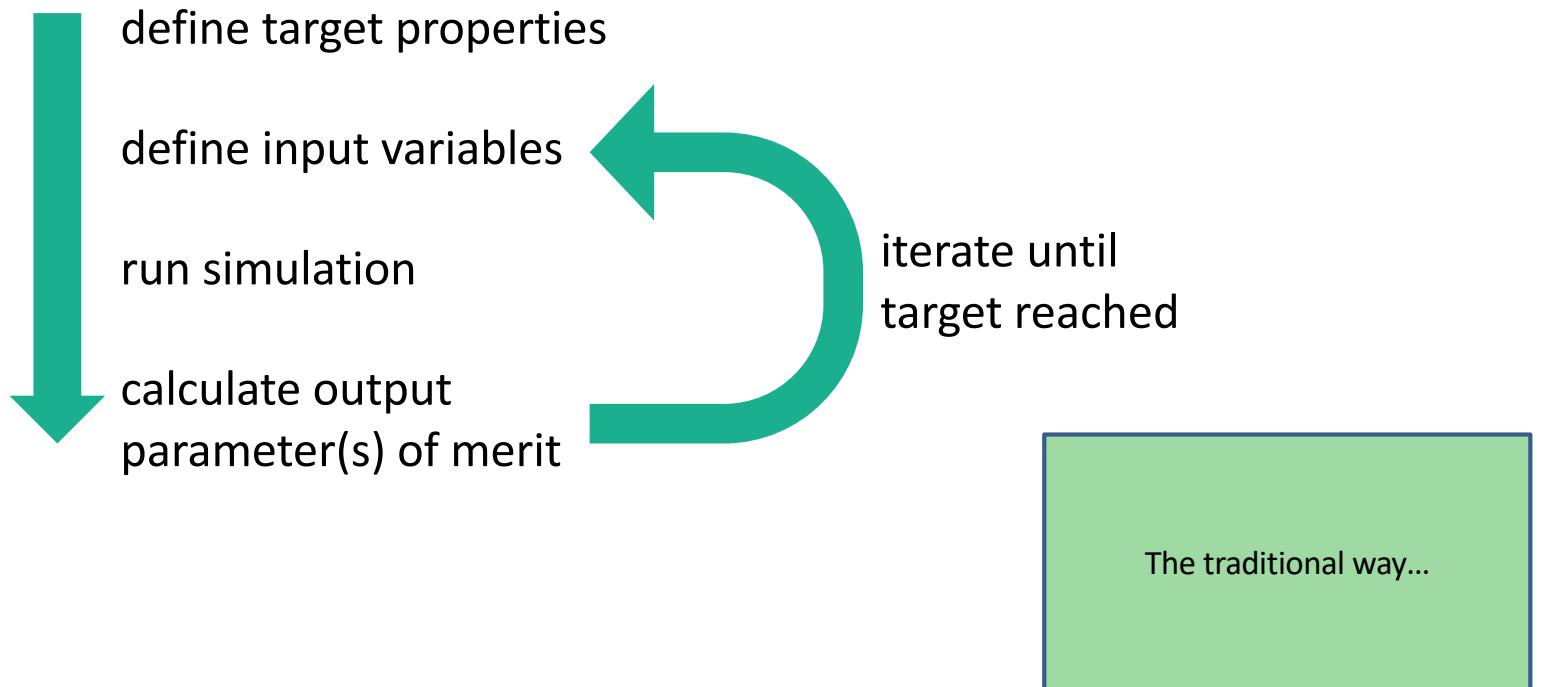
n ACSs and question sed- organic matter ode' of soul) relations?

The properties required to realize a particular device are often known, but the specific materials that harbour such properties are generally unknown and are difficult to identify. The conditions needed for water splitting, the defining qualities of coexisting transparency and conductivity in transparent conductors, and the reason

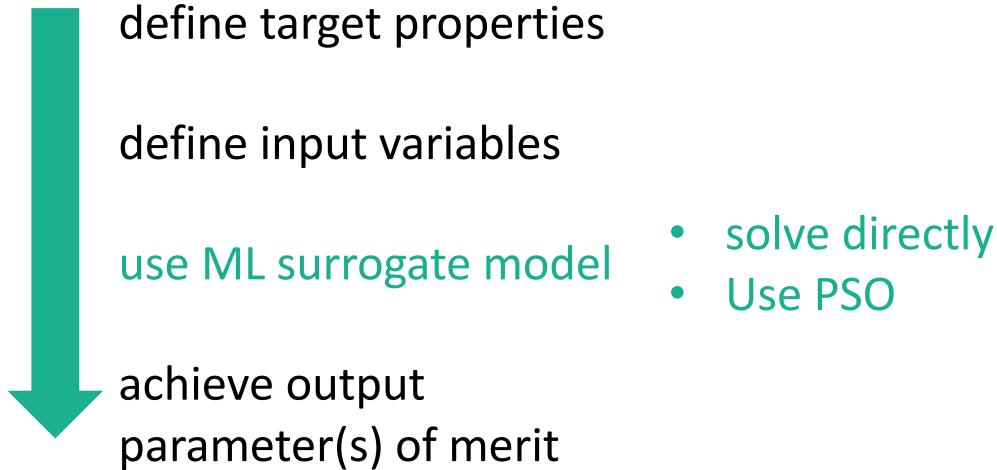
A. Zunger, *Nat. Rev. Chem.* **2**, 121 (2018)
<https://doi.org/10.1038/s41570-018-0121>

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Generative Design Approach



Generative Design Approach



...and an emergent way with the diffusion of machine learning in materials research

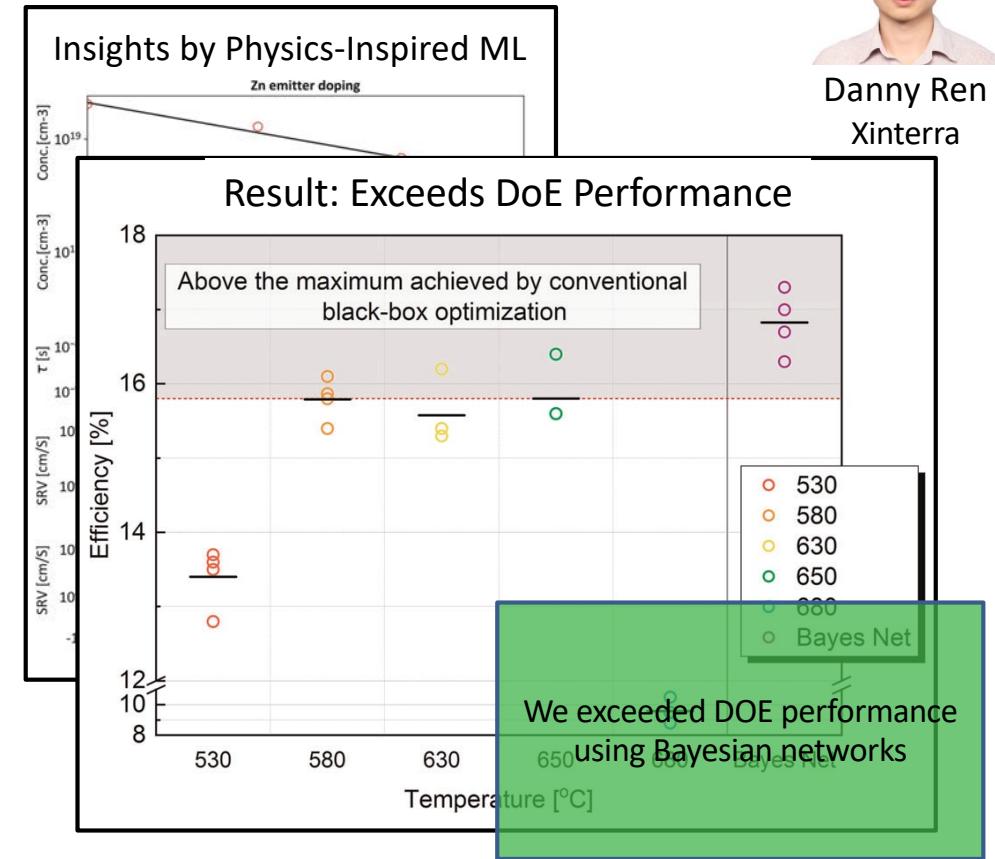
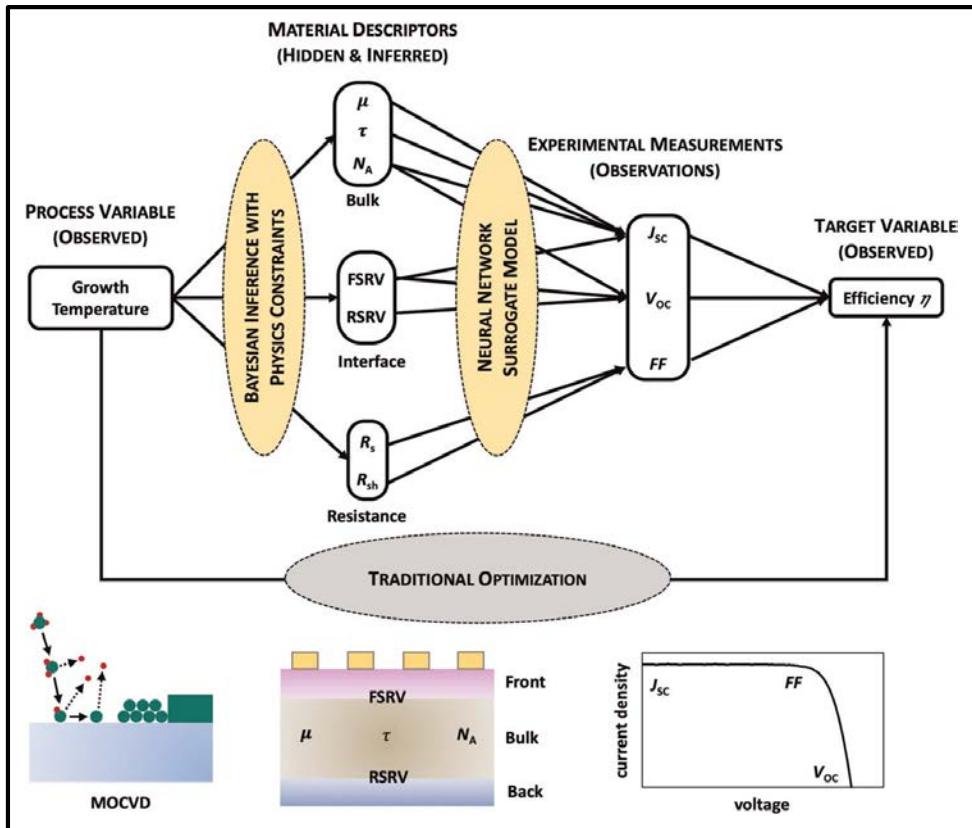
SMART: Singapore-MIT Alliance for Research & Technology



Implementing Heuristics into Bayesian Inference



Danny Ren
Xinterra



Z. Ren, F. Oviedo *et al.*, *npj Computational Materials* **6**, 9 (2020)

<https://doi.org/10.1038/s41524-020-0277-x>

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Accelerated
Materials
Laboratory for
Sustainability

Agency for Science, Technology & Advanced Research (A*STAR)

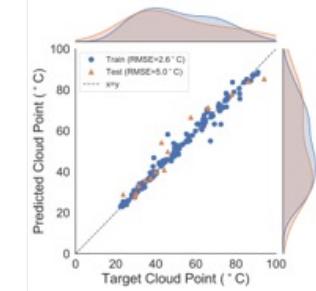
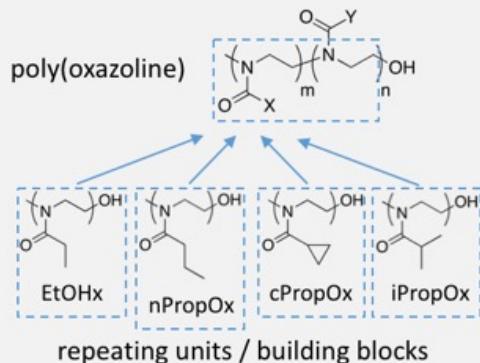


I took a year leave of absence to start up a \$18M, 5-year program at A*STAR, called Accelerated Materials Development for Manufacturing (AMDM).

- Polymers
- Nanoparticles
- Integrated Circuits
- Metals
- Magnetic Materials

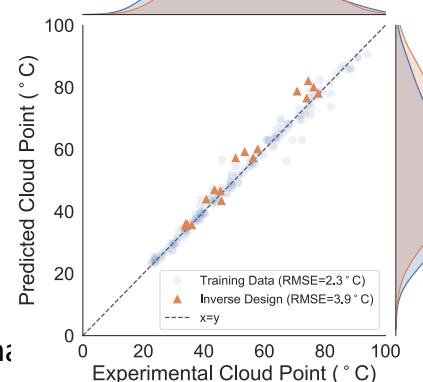
Polymer Inverse Design

Descriptor: Polymer Architecture



Model Training
Gradient Boosting w Decision Trees

Inverse Design
Particle Swarm Optimization

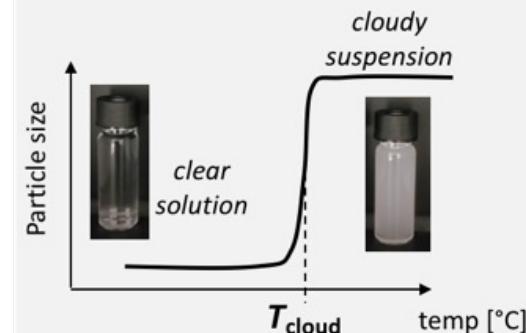


Here, a trained regressor could be explored using particle swarm optimization, to identify combinations of input conditions satisfying a user-defined target



Jatin Kumar
A*STAR; Xinterra

Feature: Cloud Point

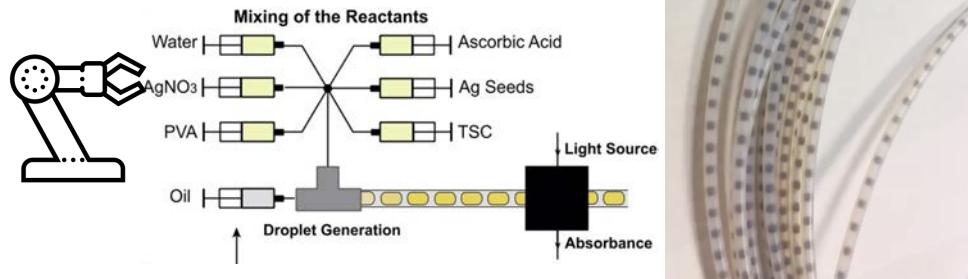


Created 17 new polymers with user-defined properties via inverse design

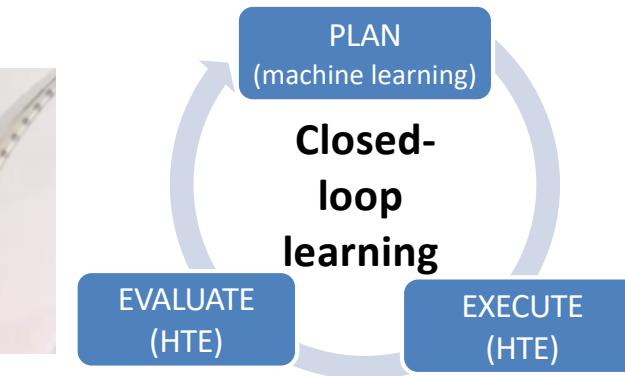
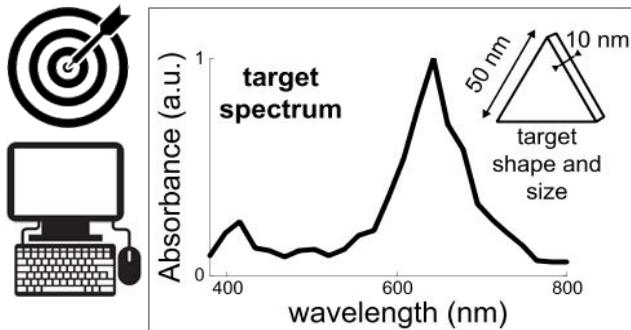


Identify Nanoparticle Process Recipe in 2 weeks (compared to 1 yr)

➤ High-throughput synthesis (microfluidics)



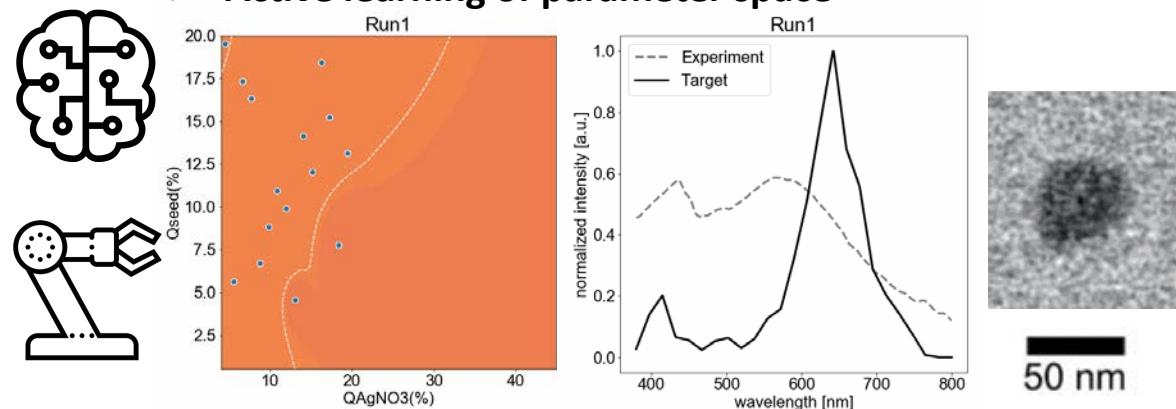
➤ Define Target (simulation)



Flor M.B.
NUS

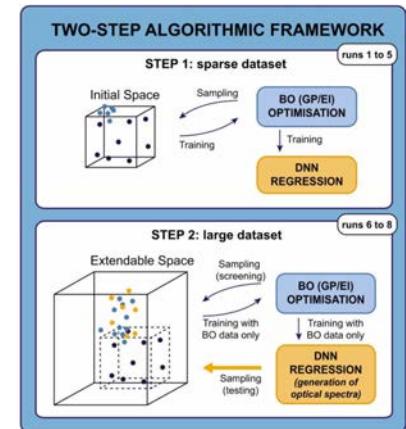
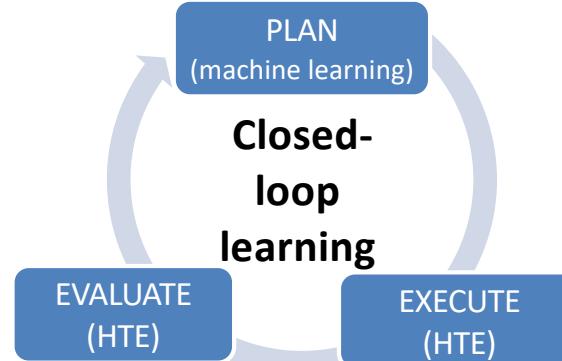
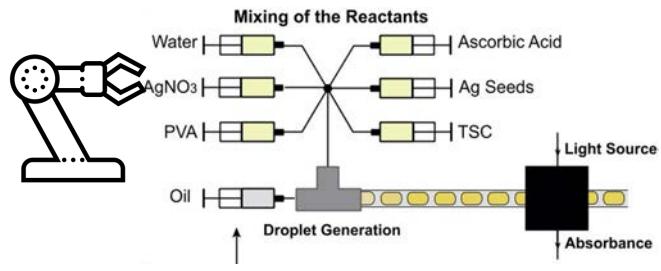
Similar, but for silver nanoparticles. We ran BO and a NN in parallel, to both optimize and regress (toward inverse design).

➤ Active learning of parameter space

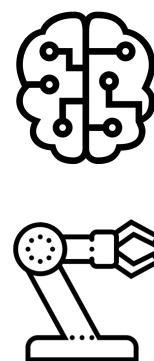
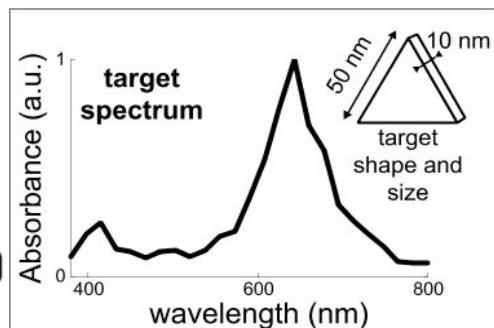
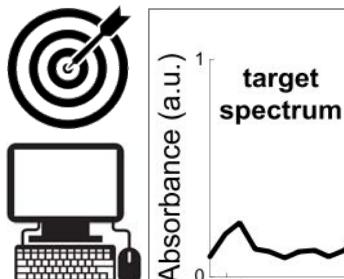


Identify Nanoparticle Process Recipe in 2 weeks (compared to 1 yr)

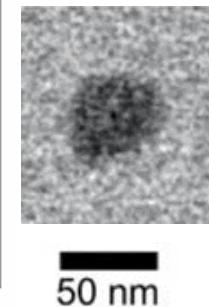
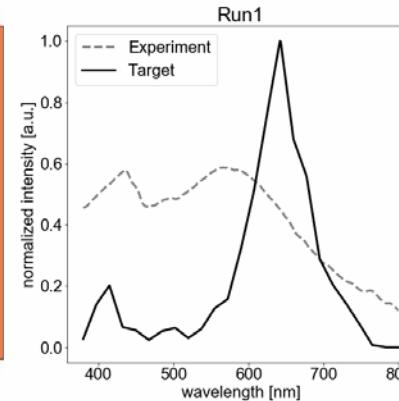
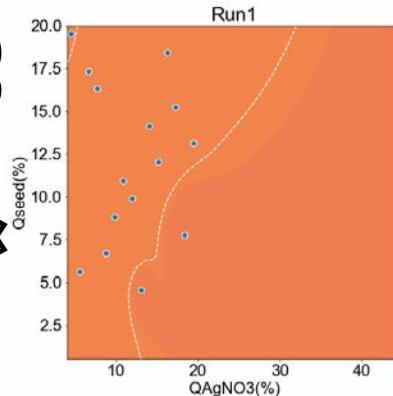
➤ High-throughput synthesis (microfluidics)



➤ Define Target (simulation)

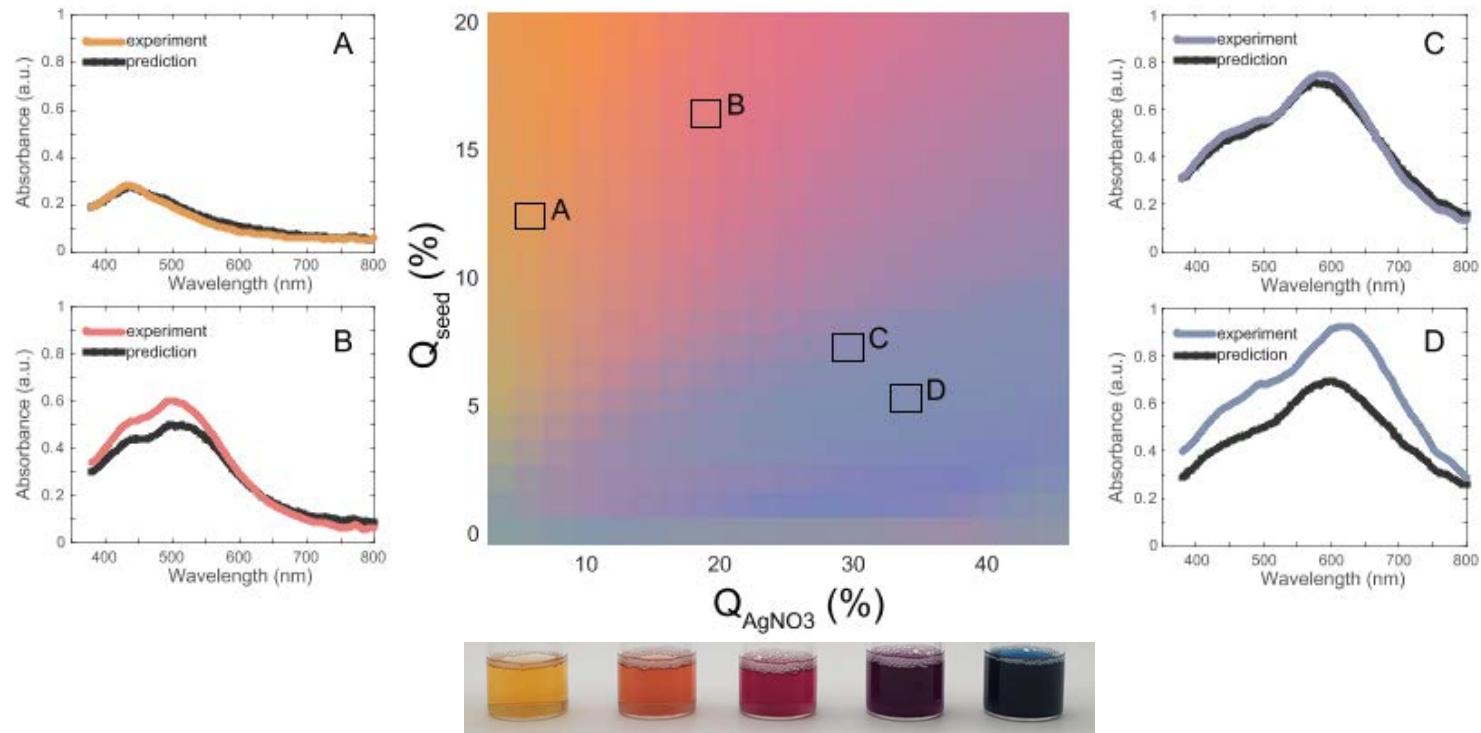


➤ Active learning of parameter space



Identify Nanoparticle Process Recipe in 2 weeks (compared to 1 yr)

Machine-Learnt Silver Nanoparticle Color Palette



Inverse design
with trained
regressors...
almost like a
lookup table.

PERSPECTIVES

Invers With t

Alex Zunger

Challenge #2: “Because the effective interactions used as input are generally not expressible (or mappable) in terms of the periodic table, there is generally no recognizable rational path between model Hamiltonian predictions and the realization of actual materials. In other words, the theory is not invertible.”

Abstract | Solid-state chemists have been consistently successful in envisioning and making new compounds, often enlisting the tools of theoretical solid-state physics to explain some of the observed properties of the new materials. Here, a new style of collaboration between theory and experiment is discussed, whereby the desired functionality of the new material is declared first and theoretical

particular device are often known, but the specific materials that harbour such properties are generally unknown and are difficult to identify. The conditions needed for water splitting, the defining qualities of coexisting transparency and conductivity in transparent conductors, and the reason

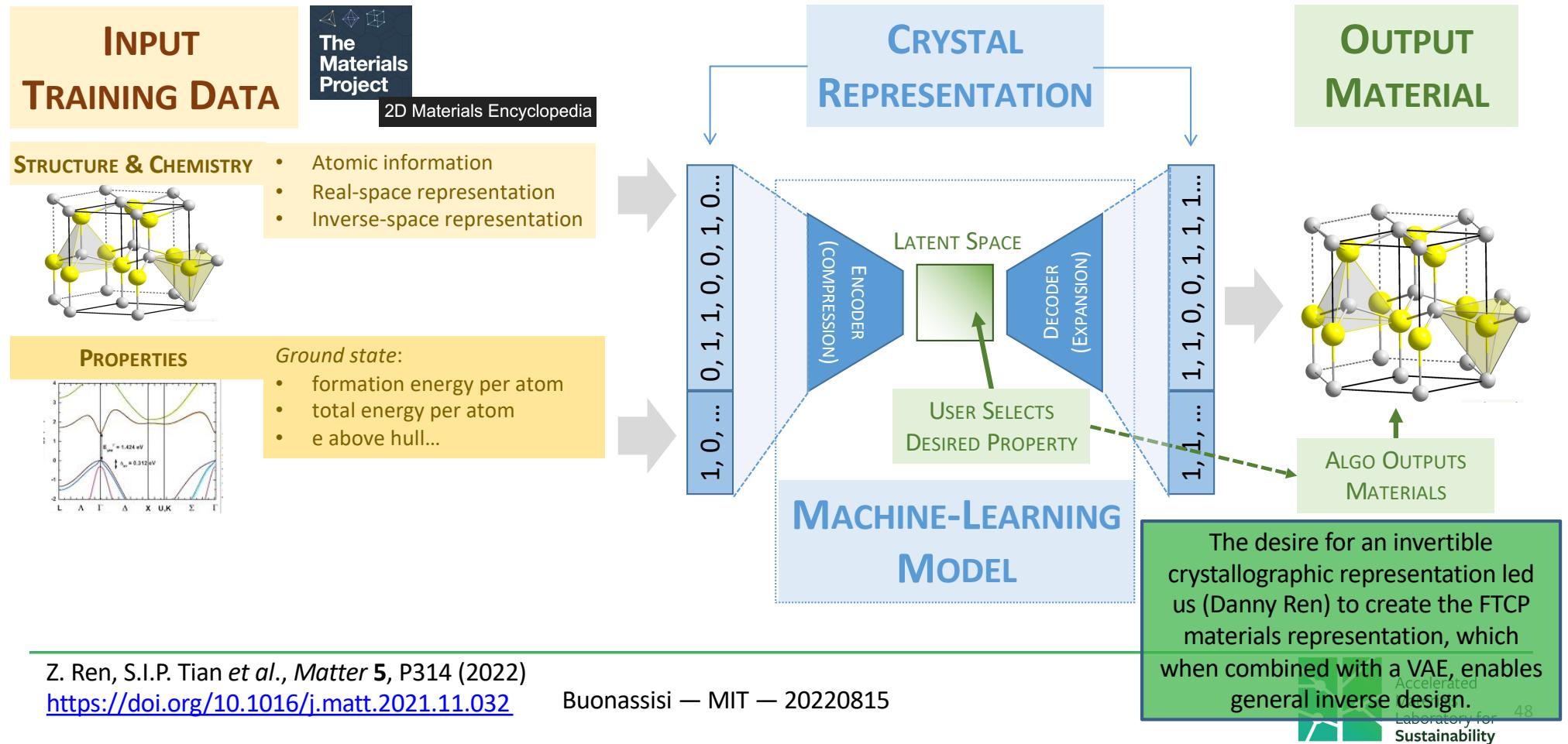
A. Zunger, *Nat. Rev. Chem.* **2**, 121 (2018)
<https://doi.org/10.1038/s41570-018-0121>

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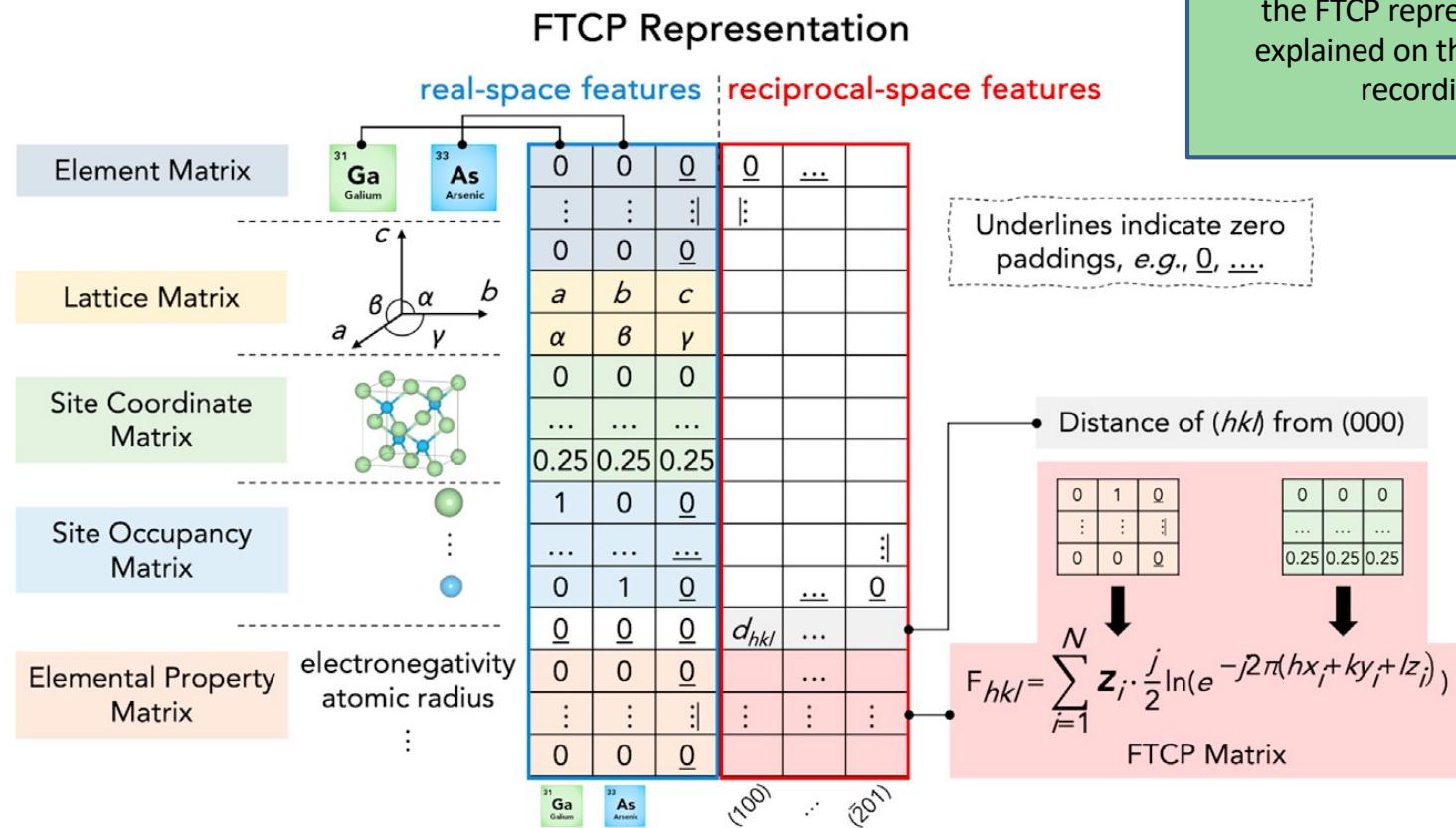


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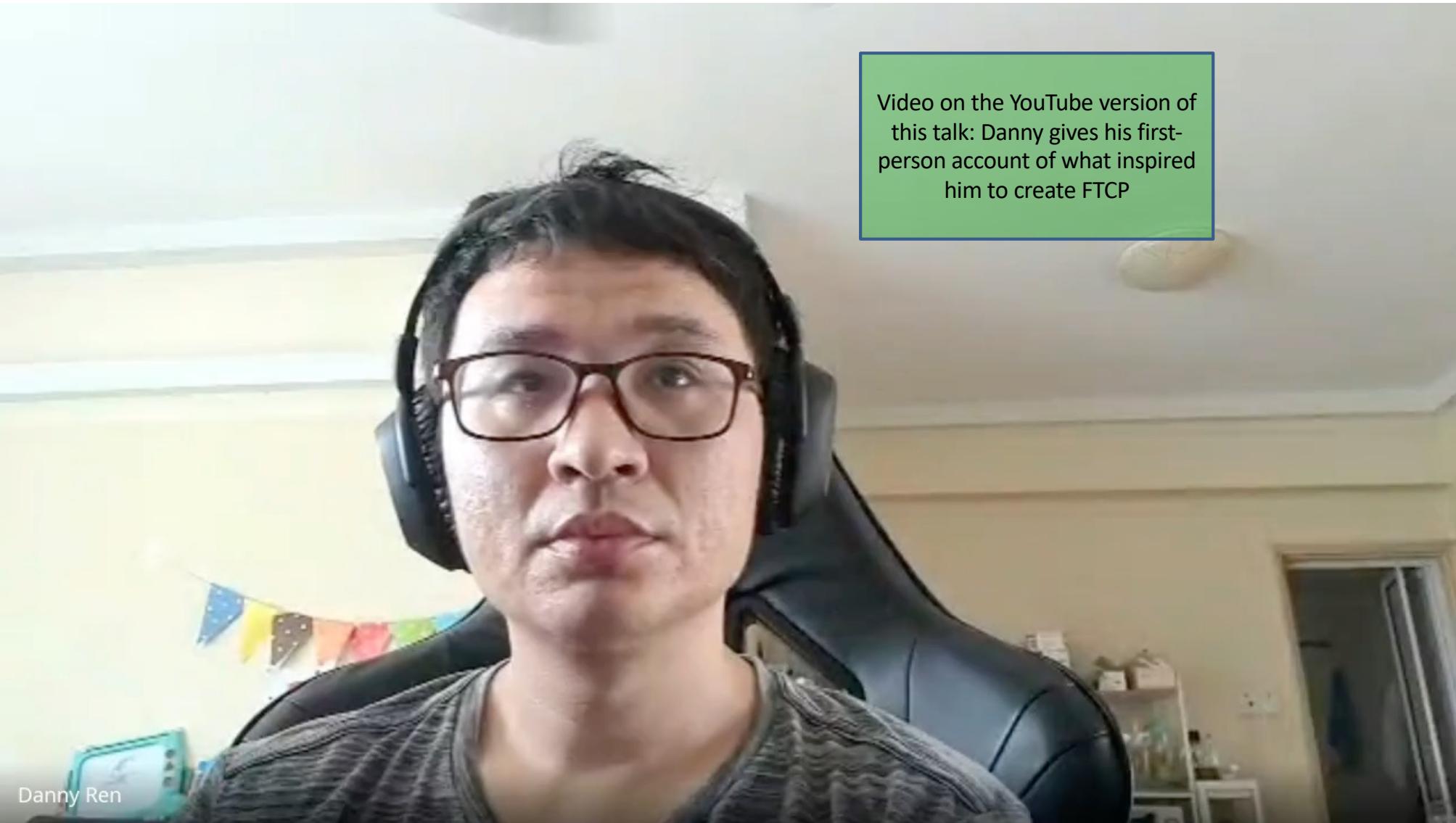
Inverse Design using FTCP



FTCP Representation



Here are the different pieces of the FTCP representation, explained on the YouTube recording.



Video on the YouTube version of
this talk: Danny gives his first-
person account of what inspired
him to create FTCP

Danny Ren

PERSPECTIVES

This is the big one. Many groups out there trying to predict synthesizability, synthesis routes, metastability...

Invers With t

Alex Zunger

Abstract | Solid-state chemists have been consistently successful in envisioning and making new compounds, often enlisting the tools of theoretical solid-state physics to explain some of the observed properties of the new materials. Here, a new style of collaboration between theory and experiment is discussed, whereby the desired functionality of the new material is declared first and theoretical

Challenge #3: “because the total electron and ion energy of the system is generally not computed or optimized with respect to the structural degrees of freedom, it is difficult to assess whether the predicted properties (such as exotic superconductivity and quantum spin liquids) will be found in stable (or nearly stable) recognizable compounds”

in ACSs and question sed- organic matter ode' of soul) relations? realize a

particular device are often known, but the specific materials that harbour such properties are generally unknown and are difficult to identify. The conditions needed for water splitting, the defining qualities of coexisting transparency and conductivity in transparent conductors, and the reason

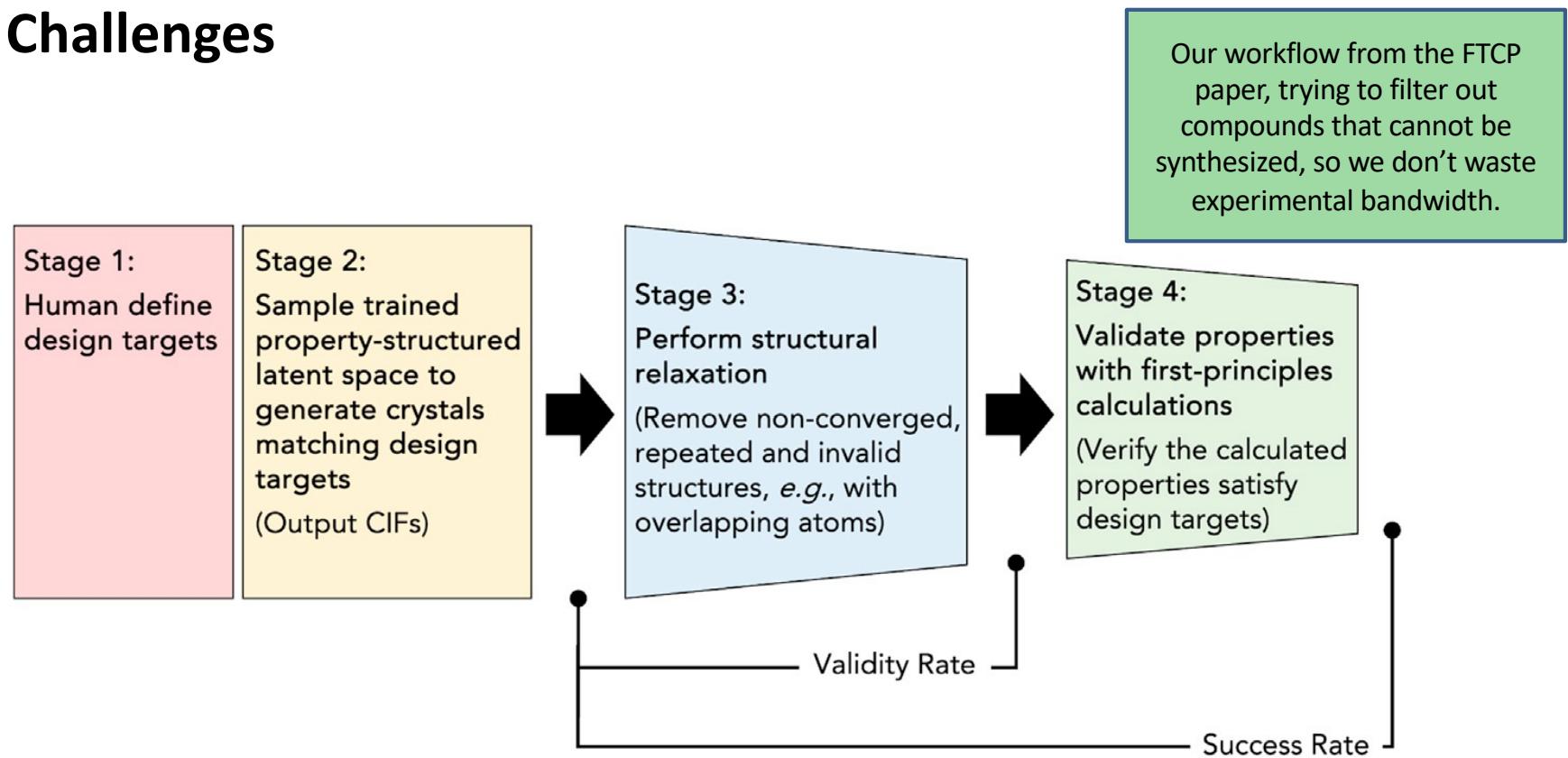
A. Zunger, *Nat. Rev. Chem.* **2**, 121 (2018)
<https://doi.org/10.1038/s41570-018-0121>

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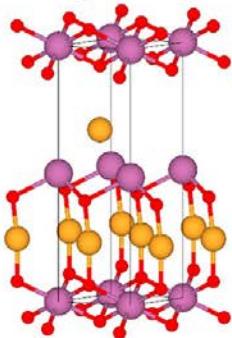
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Open Challenges



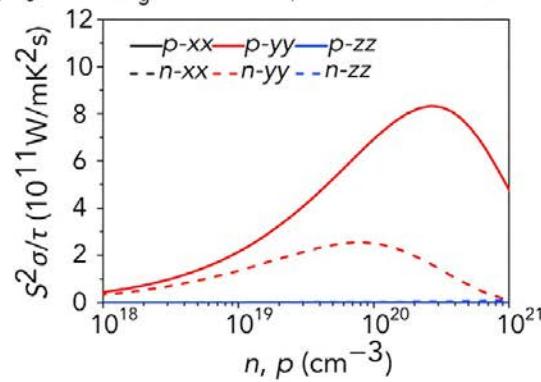
Examples

A Designed Crystal 1: $\text{Au}_2\text{Sc}_2\text{O}_3$

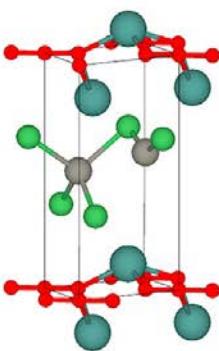


● Sc
● Au
● O

(E_g 0.95 eV, E_f -2.34 eV/atom)

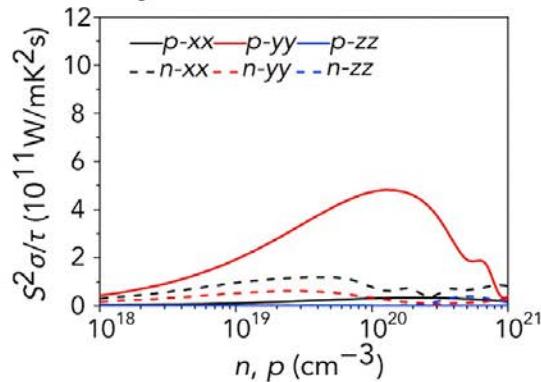


B Designed Crystal 2: $\text{Y}_2\text{Zn}_2\text{As}_2\text{O}_3$

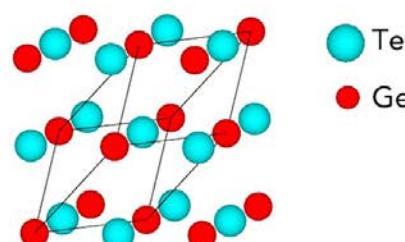


● Y
● Zn
● As
● O

(E_g 0.70 eV, E_f -1.66 eV/atom)

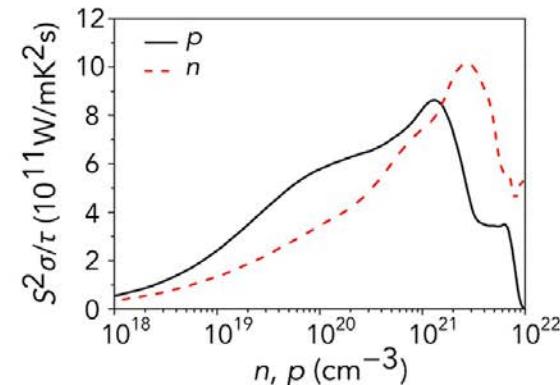


C State-of-the-art



● Te
● Ge

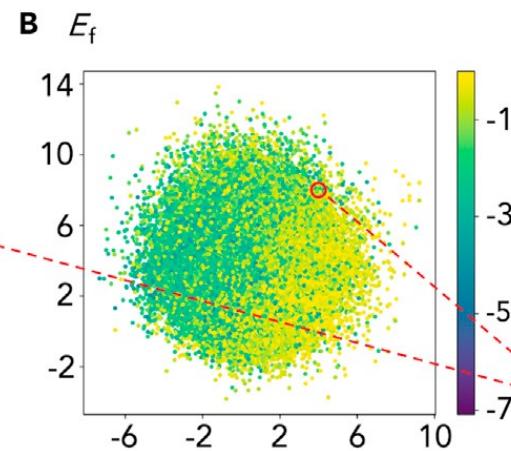
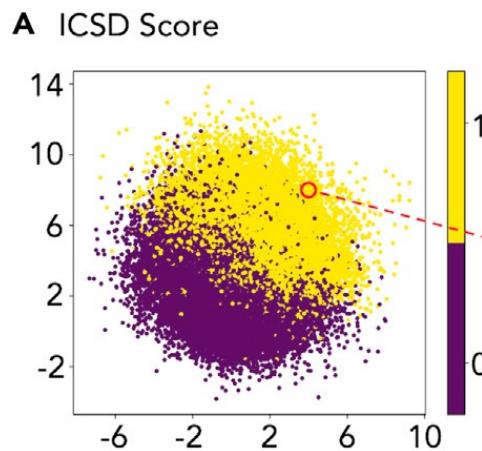
GeTe c-axis ground truth
(state-of-the-art material)



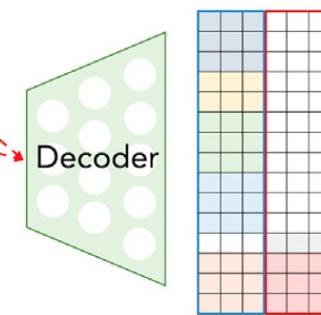
Examples of candidate thermoelectric compounds predicted by FTCP that were validated by BoltzTraP and DFT... but not experiment.



“ICSD Score” as a Synthesizability Metric



One approach to predict synthesizability: How closely does the candidate resemble a compound in the ICSD (*i.e.*, already synthesized)? This can be a target in property-segmented latent space, just like any other material property (formation energy, bandgap...). (*NB*: We are not the first to propose this...)



What Information is Necessary and Sufficient to Predict Materials Properties using Machine Learning?

Siyu Isaac Parker Tian, Aron Walsh, Zekun Ren, Qianxiao Li, Tonio Buonassisi

S.I.P. Tian *et al.*, arXiv (2022)
<https://arxiv.org/abs/2206.04968>

Materials representations:
Video! Isaac presents his new
paper, “What information is
necessary & sufficient to predict
materials properties?”

Here's one of
the earlier
conversations
about that
paper.



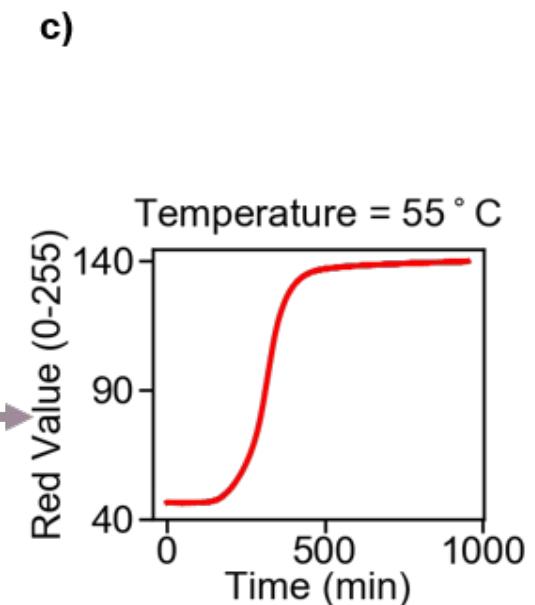
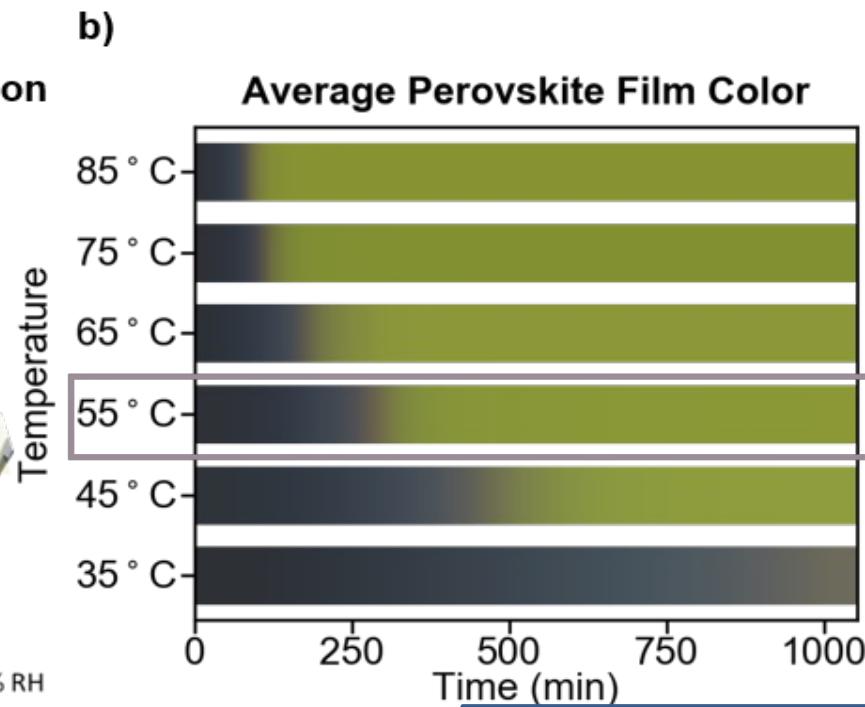
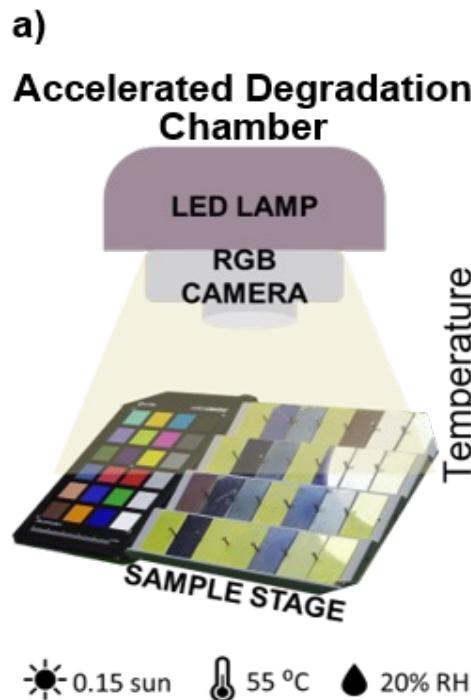
March 4, 2021

I didn't have time to cover this
during the talk, so here are some
bonus slides...

Gaining Human Trust

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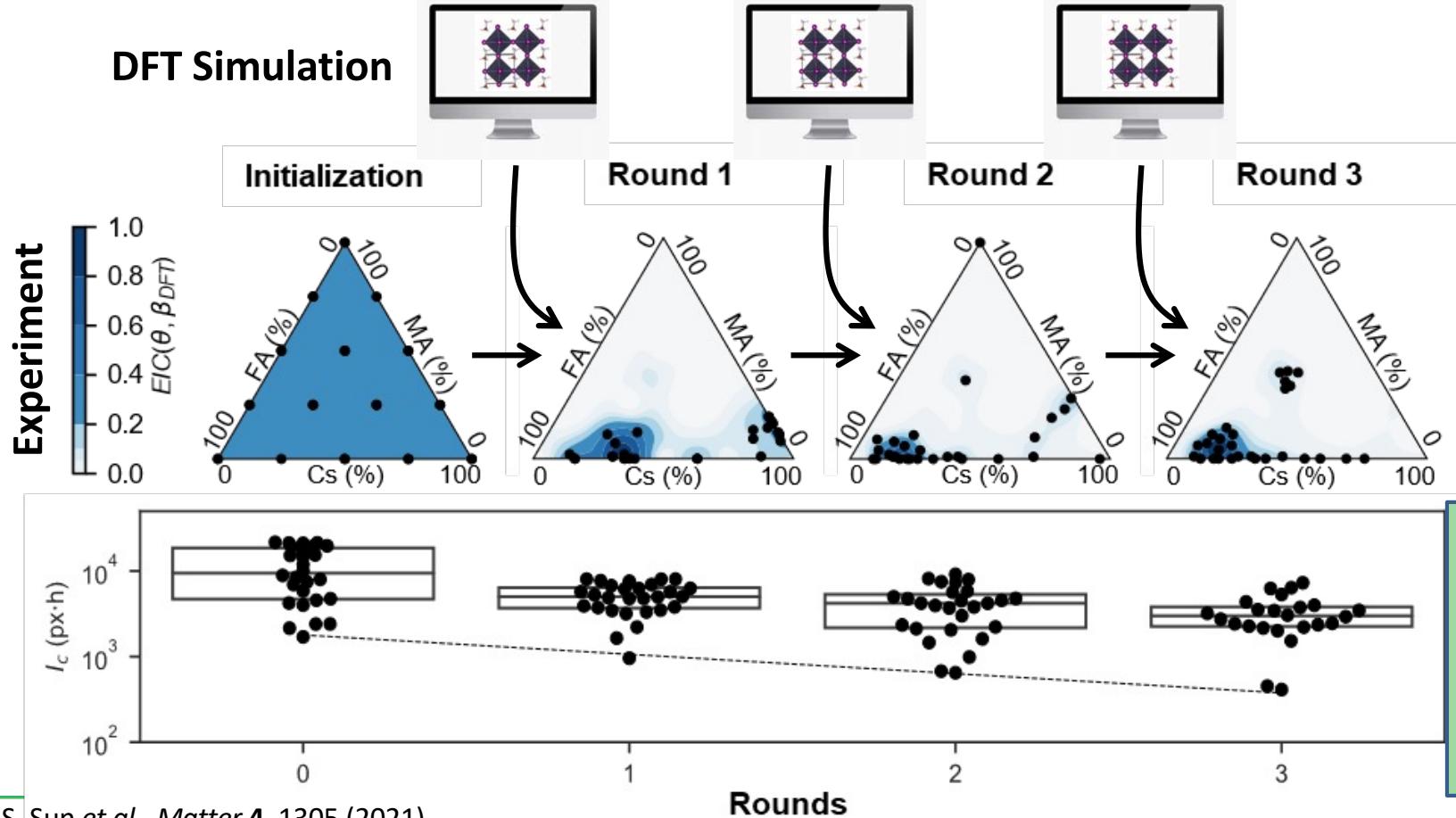
Study of Perovskite Film Degradation as a Function of Temperature



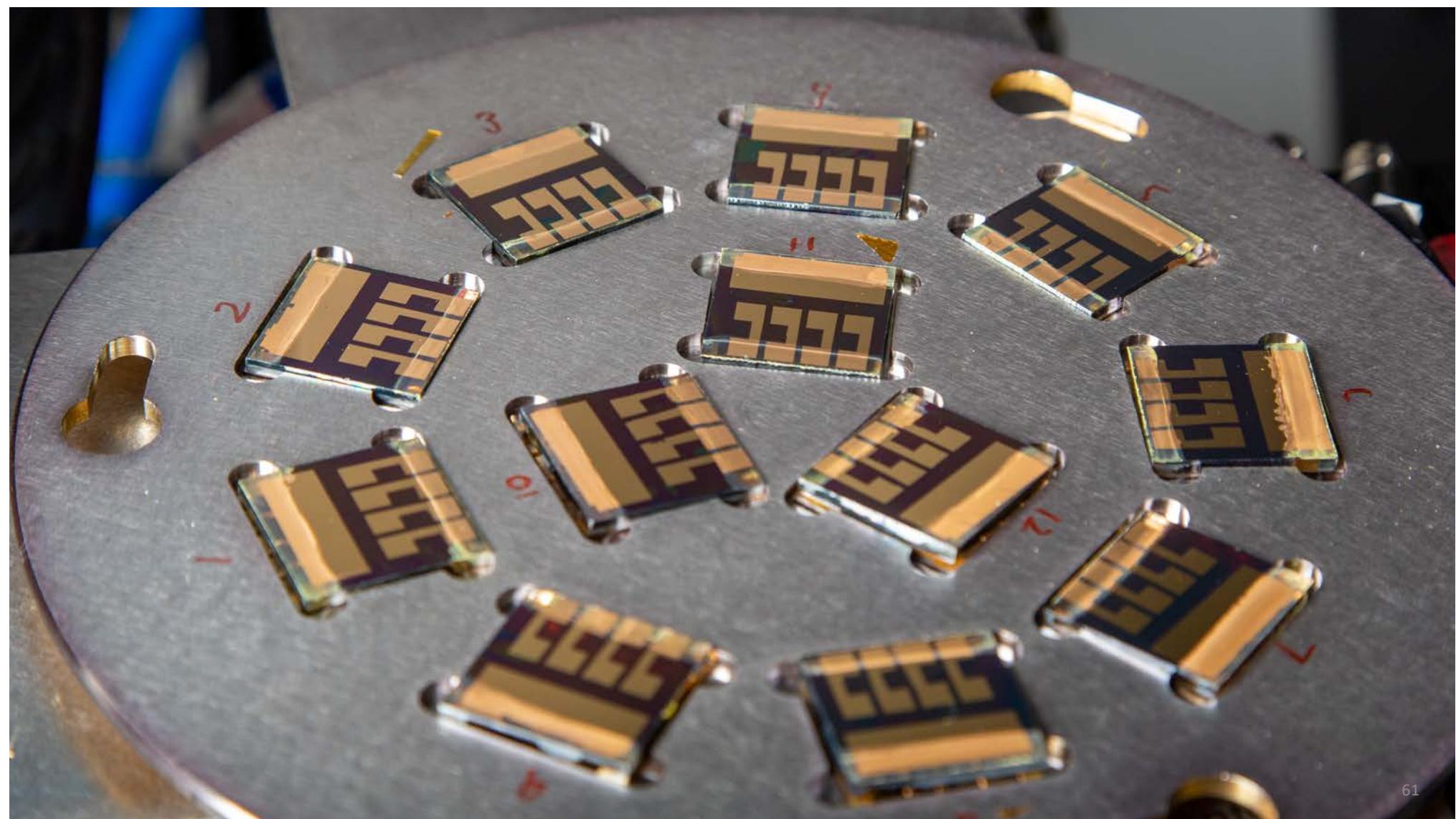
Here's an example set of experiments, whereby scientific knowledge can be input and extracted from ML-guided experiments. Here, we used color-calibrated, camera-based imaging to detect the onset of perovskite (MAPI) film degradation.



Step 2: Optimize Composition Using Hybrid Model

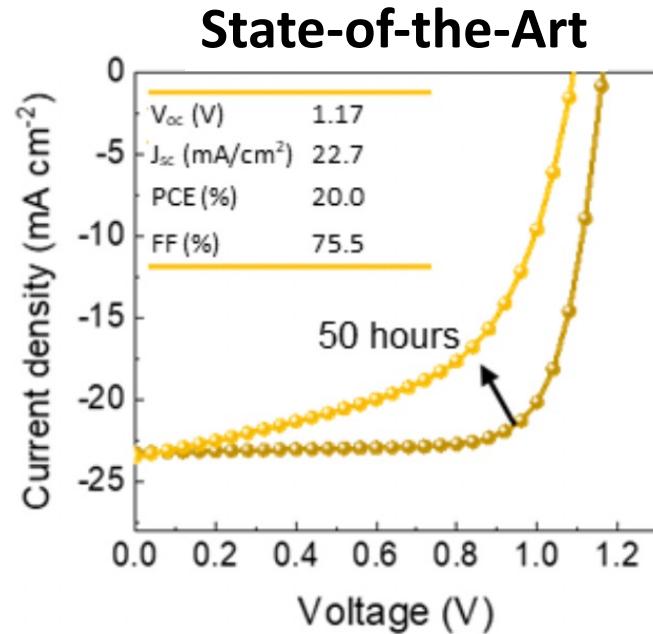


We used interpretable DFT data as a “soft constraint” to nudge the Bayesian optimization cycles, toward identifying more stable perovskite compositions.

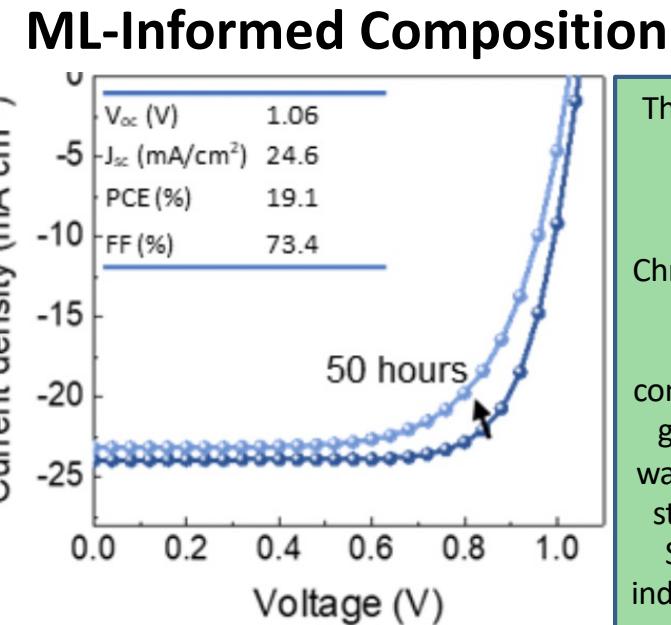


Bonus: Demonstrate Scale-Up with Fewer Attempts

Device fabrication performed by the team of Christoph Brabec



Environmental testing: 50 hrs under 85% RH / 85°C in air

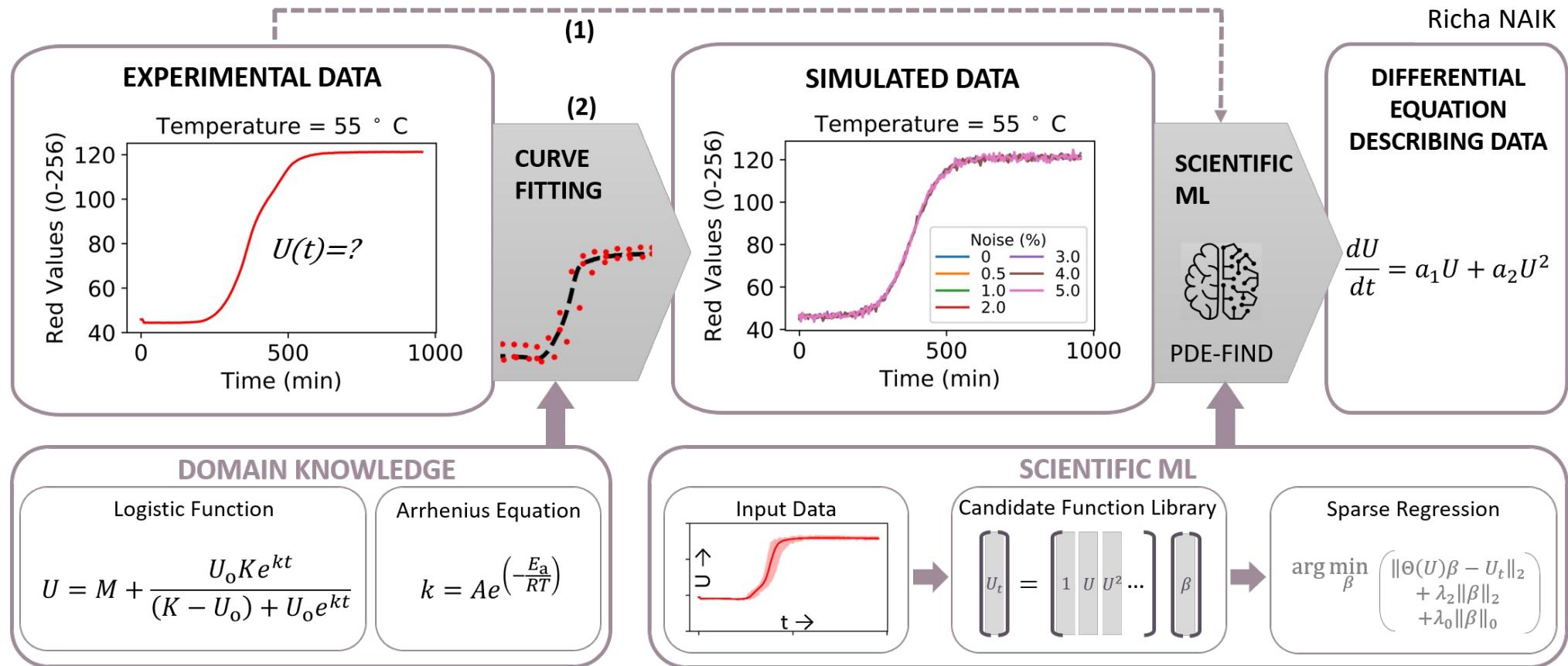


The pandemic gave us an opportunity to cement new collaborations. Christoph Brabec' team in Germany made devices, which confirmed that our ML-guided composition was considerably more stable than the prior SOTA composition, indicating that this film-based proxy measurement translates into full devices.



Richa NAIK

Extract Knowledge: Infer Root Cause(s) of Degradation



R.R. Naik *et al.*, *npj Computational Materials* **8**, 72 (2022)

<https://doi.org/10.1038/s41524-022-00751-5>

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We used sparse regression to identify the equations governing MAPI degradation.



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Human Interpretation

$$\frac{dU}{dt} = a_0 + a_1 U + a_2 U^2$$

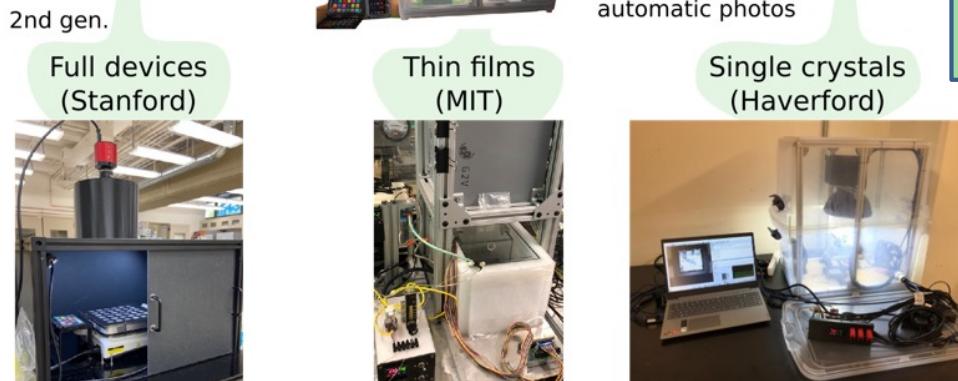
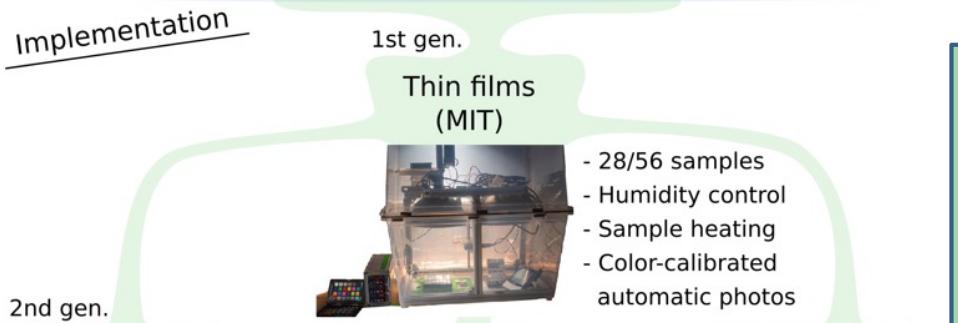
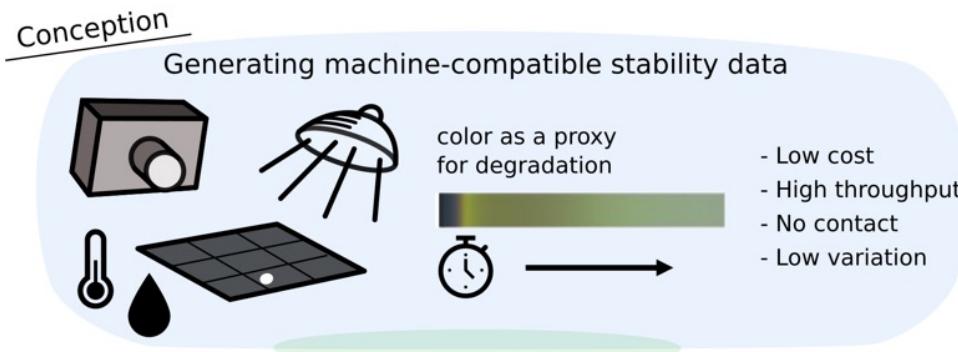


The identified equation corresponds to an autocatalytic reaction — meaning that once the reaction starts, it is hard to stop. The trick is to avoid nucleation sites that can promote decomposition, suggesting greater film (+precursor) purity is important. Interestingly, OLEDs underwent similar learning, to achieve commercial displays.

OLEDs suggest a pathway for perovskites:

- Hermetically seal the device
- Avoid impurities (*i.e.*, nucleation points for phase transformation) in the sealed devices.

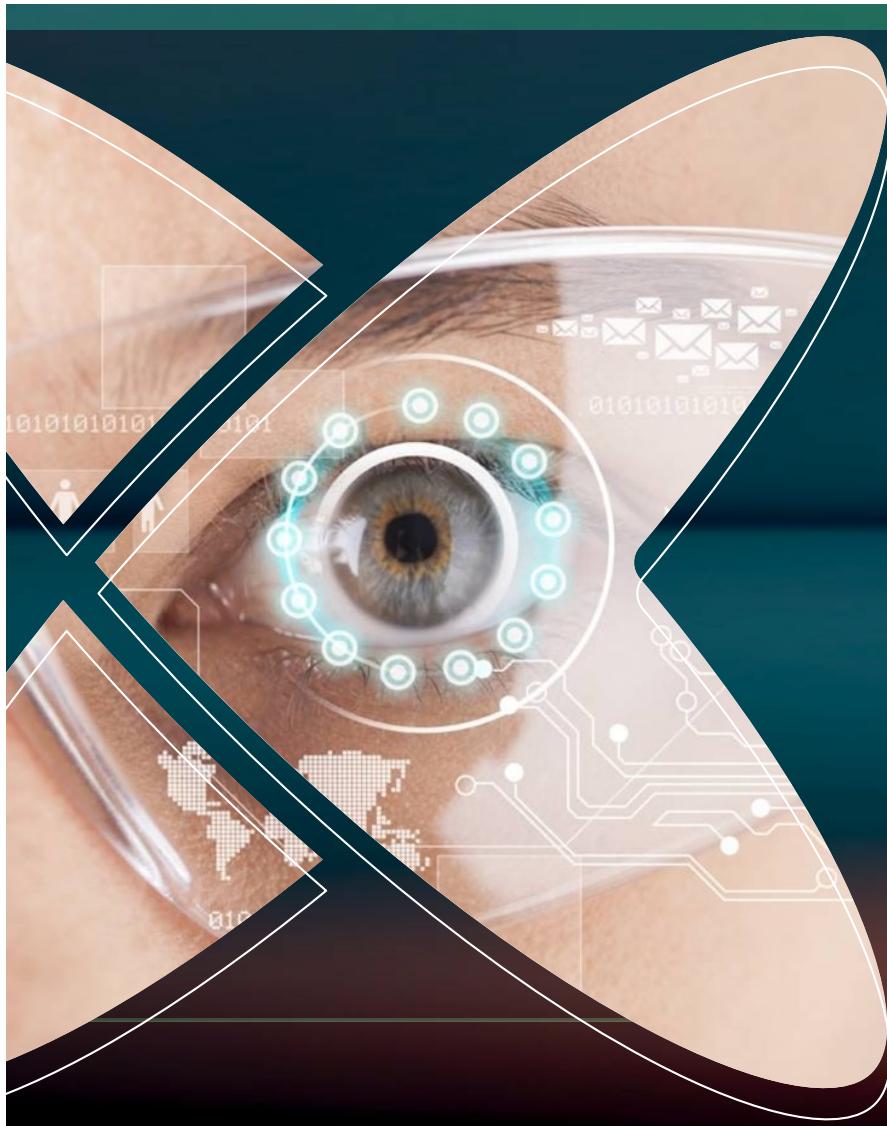
Open Science



After sharing our environmental chamber design with two other labs, we're open sourcing the entire "generative design" of hardware models: Bills of materials, assembly instructions, GitHub code, and manuscript.

Publication forthcoming...

- For glove box
 - Full enclosure
 - Large image area
- 1 Sun illumination
 - Tunable spectrum
 - Condensation-free
- Variable sample size
 - Small samples (1cm^2)
 - Detail 3



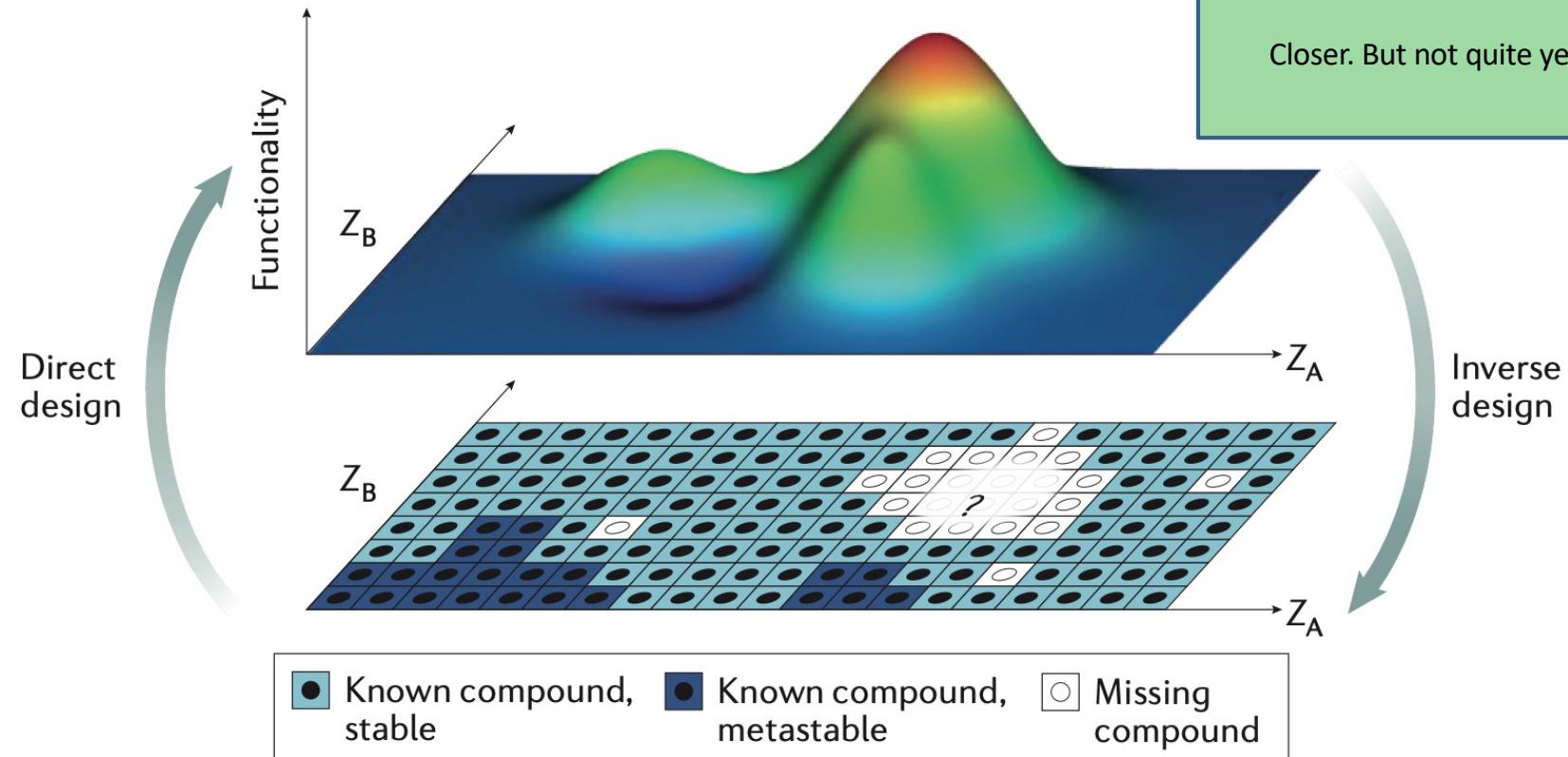
Several of the co-authors of the papers you've just seen formed Xinterra, a startup dedicated to accelerating the development of new materials for sustainability. Co-founder Kedar Hippalgaonkar will share an update at the Acceleration Consortium meeting in Toronto Aug 30 – Sept 2, 2022.



**Faster and
efficient
materials
development**



Inverse Design: Are We There Yet?



A. Zunger, *Nat. Rev. Chem.* **2**, 121 (2018)
<https://doi.org/10.1038/s41570-018-0121>

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“Green Spaces” in Inverse Design

- Materials representations that unlock simpler algorithms
- Improved “refinements” to existing models
- Transformers
- Synthesizability estimators
- Knowledge inference & uncertainty estimation
- Ethics & Dual Use

Here are some promising research areas.

Lab Logo: “A leaf blended with an integrated circuit”

 Hugging Face
DALL-E mini by craiyon.com



...I think with the right use cases, they can be transformative. We used this to inspire our new lab logo. We then passed these inspirations to a professional logo designer, and got...

Lab Logo: “A leaf blended with an integrated circuit”

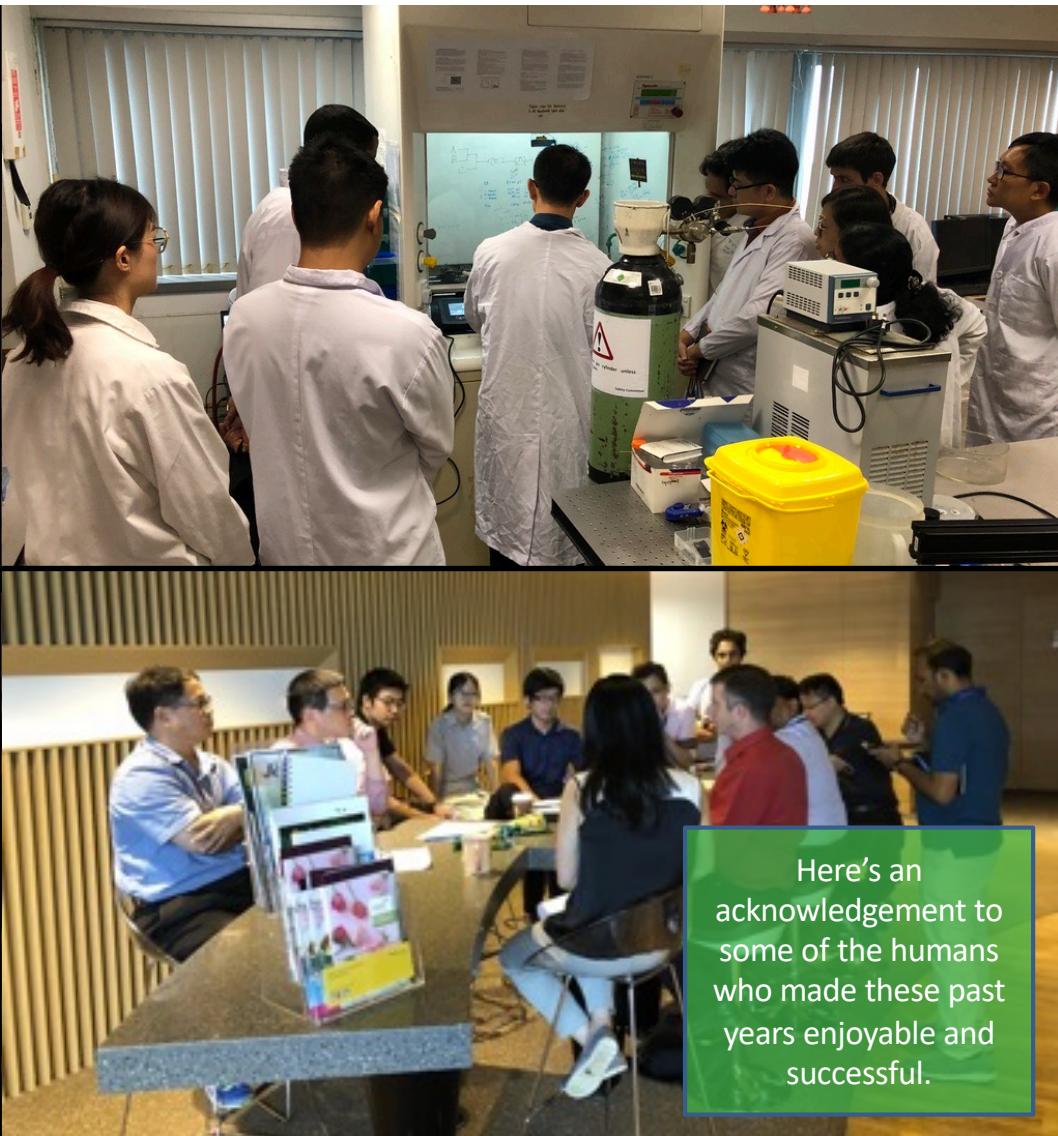
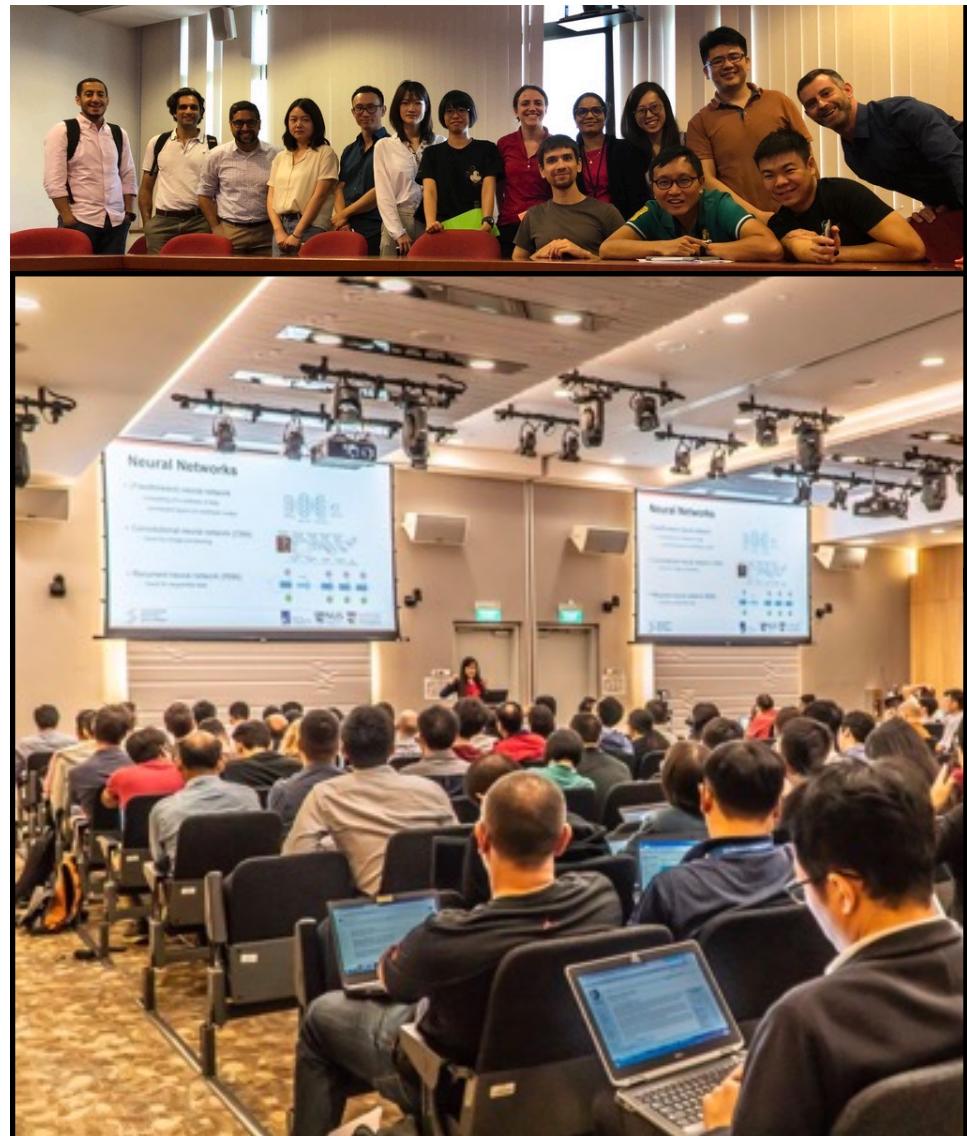


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Arjie Lupac,
entrepreneur

...this. I think
combining human and
machine intelligence
might still be the way
to go in this field for
some time.



Here's an acknowledgement to some of the humans who made these past years enjoyable and successful.



Here's an acknowledgement to some of the humans who made these past years enjoyable and successful.

Goal:

**You give me a set of properties,
I give you a material.**

We're not there yet,
but we've made
progress.

Perhaps you can help
make this possible.



Use your power,
responsibly.



Photo by <http://www.zainalandzainal.com/>



<https://twitter.com/toniobuonassisi>



<https://www.LinkedIn.com/today/author/tonio-buonassisi-7854543>



<https://pv-lab.github.io/perovskit/>