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SAND2022-15915C

CHALLENGING THE LIMITS SURROUNDING THE ADOPTION OF AI-GUIDED MANUFACTURING FOR MATERIALS RELIABILITY

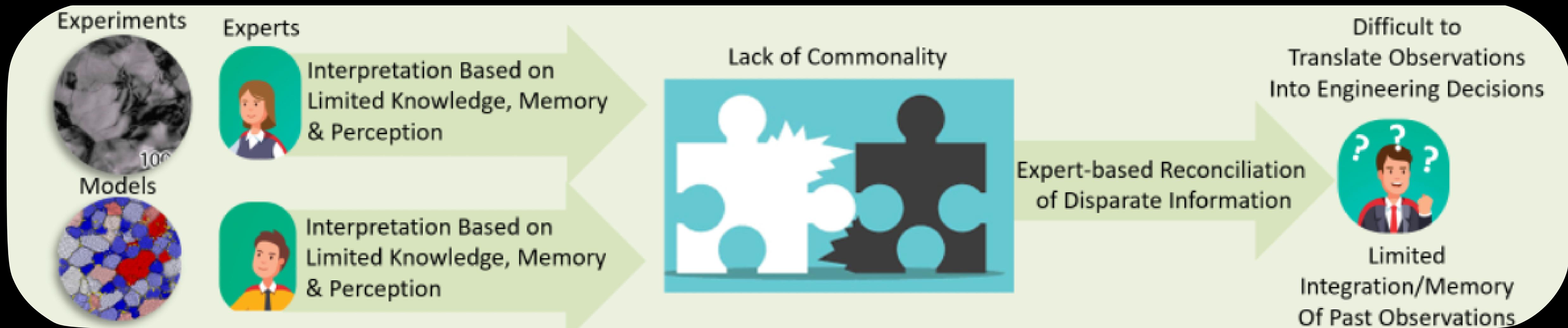
RÉMI DINGREVILLE (RDINGRE@SANDIA.GOV)
SANDIA NATIONAL LABORATORIES



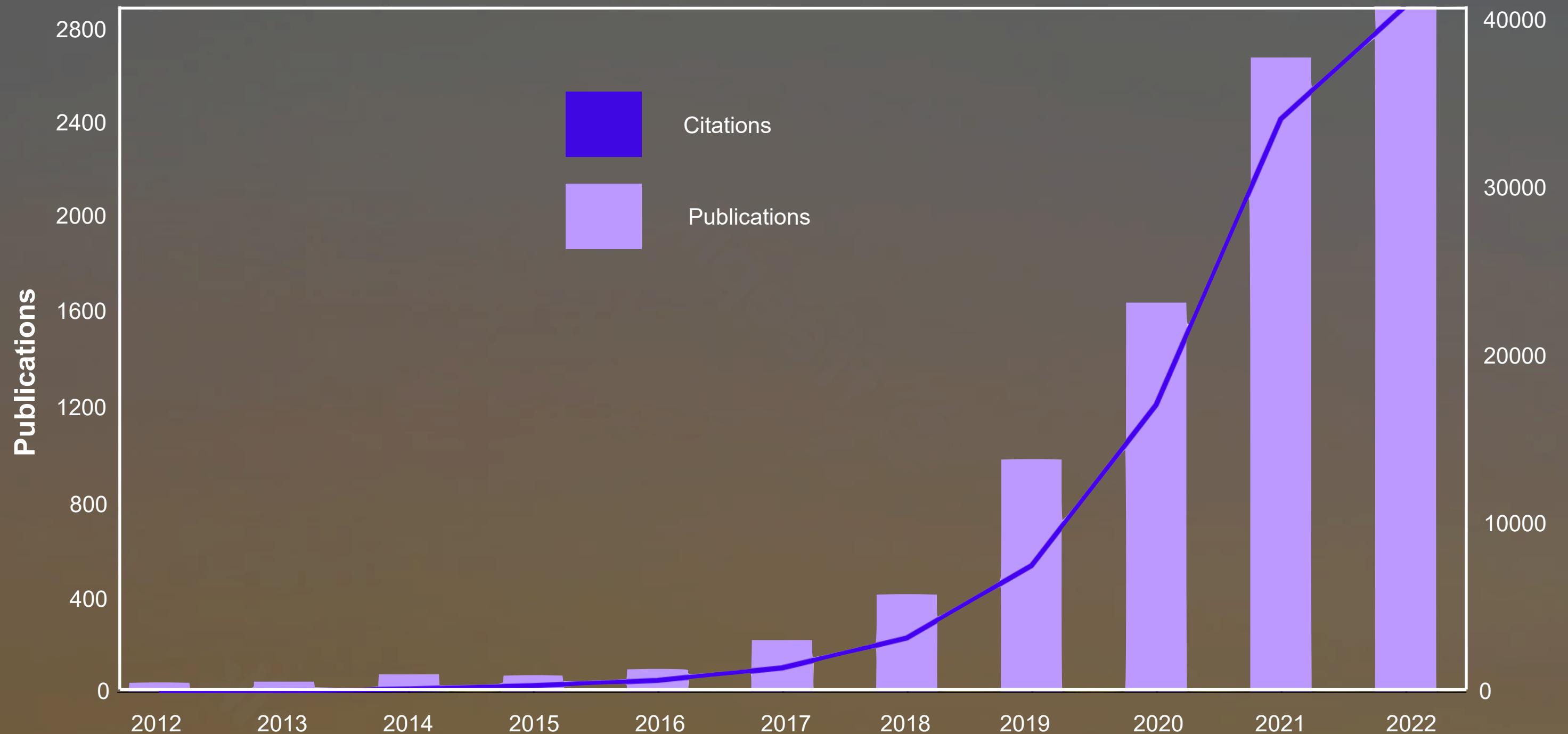
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OUR CURRENT APPROACH TO MATERIALS RELIABILITY IS SLOW, INCONSISTENT, COSTLY AND LIMITED BY OUR OWN COGNITION

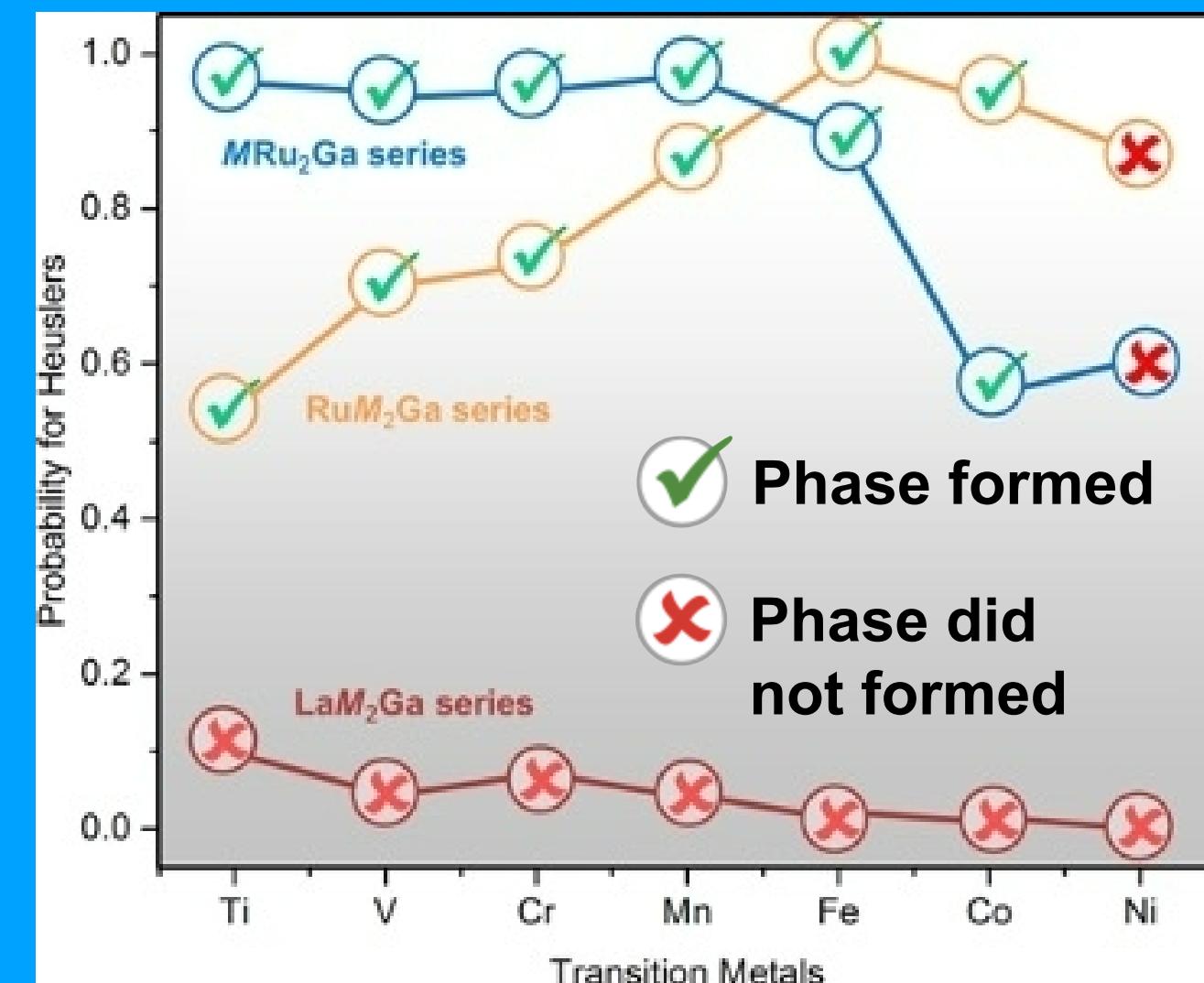


OVER THE PAST DECADE, AI HAS INFILTRATED MSE AND MANUFACTURING PROCESSES



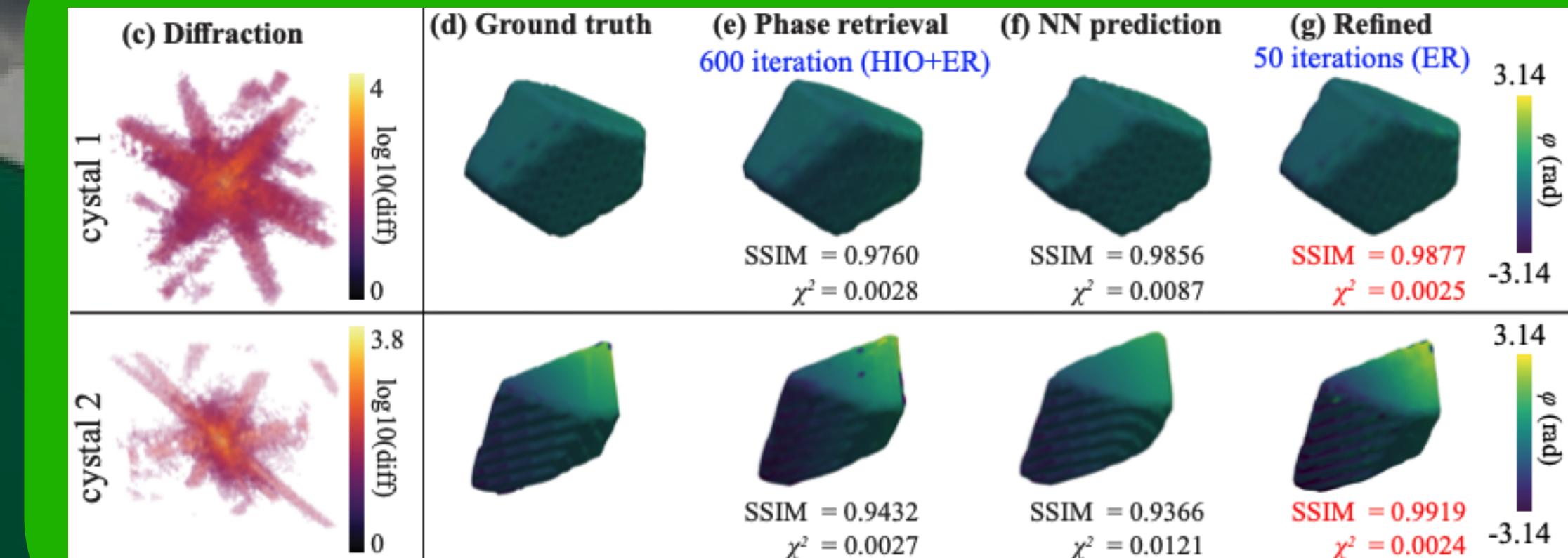
THE EFFECTIVE USE OF AI PROMISES TO IMPROVE OUR ABILITY TO Do “BUSINESS” DIFFERENTLY

MATERIALS DISCOVERY



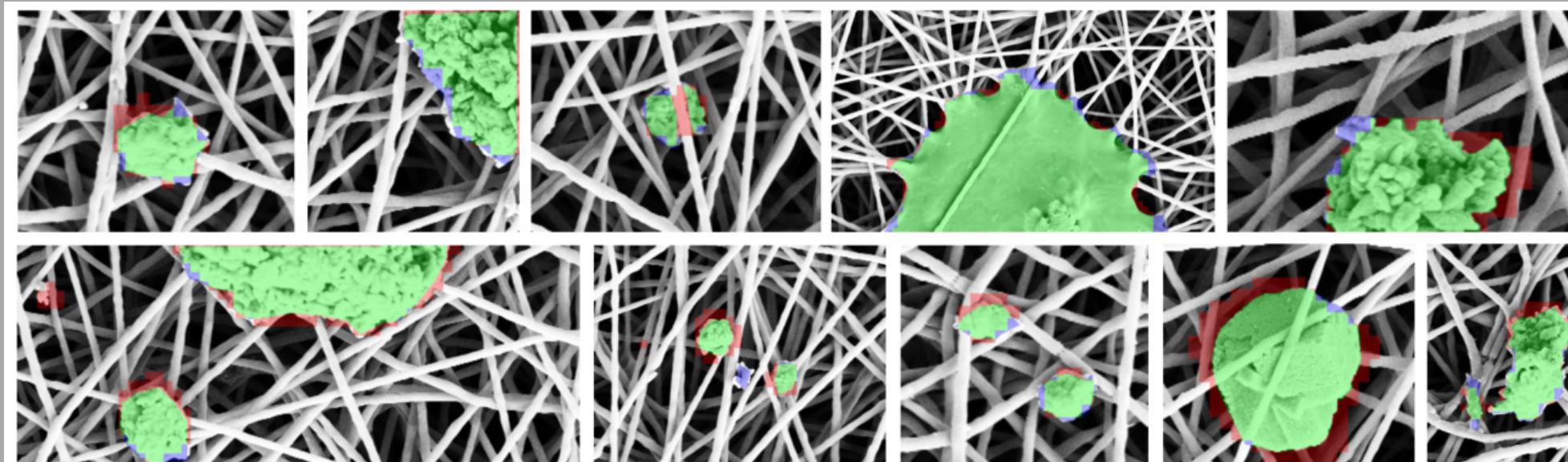
Oliynyk et al., *Chem. Mater.*, 2016

MATERIALS FINGERPRINTING



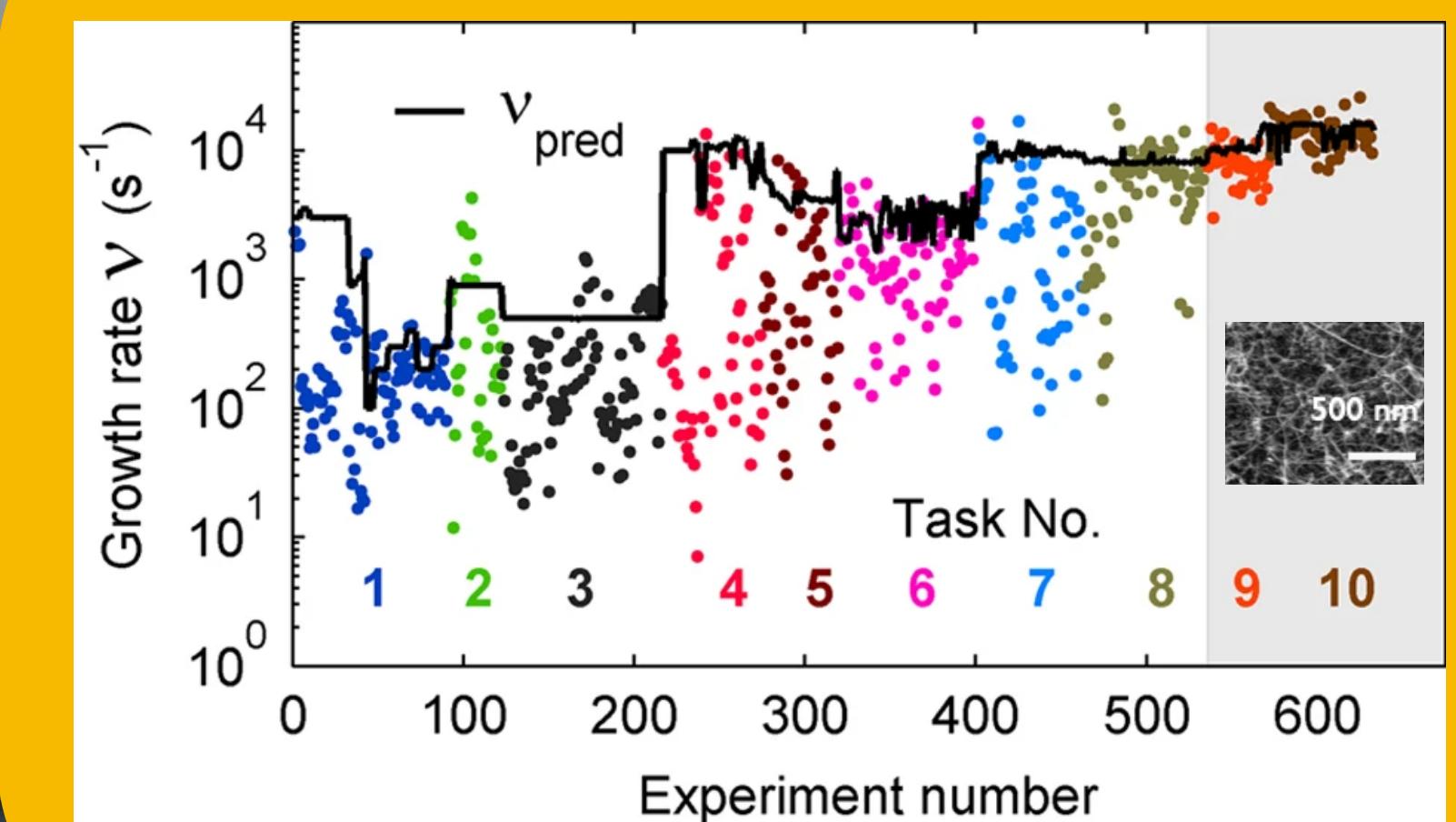
Yao et al., *arXiv*, 2022

ANOMALY DETECTION



Napoletano et al., *Sensors*, 2018

AUTONOMOUS SYNTHESIS

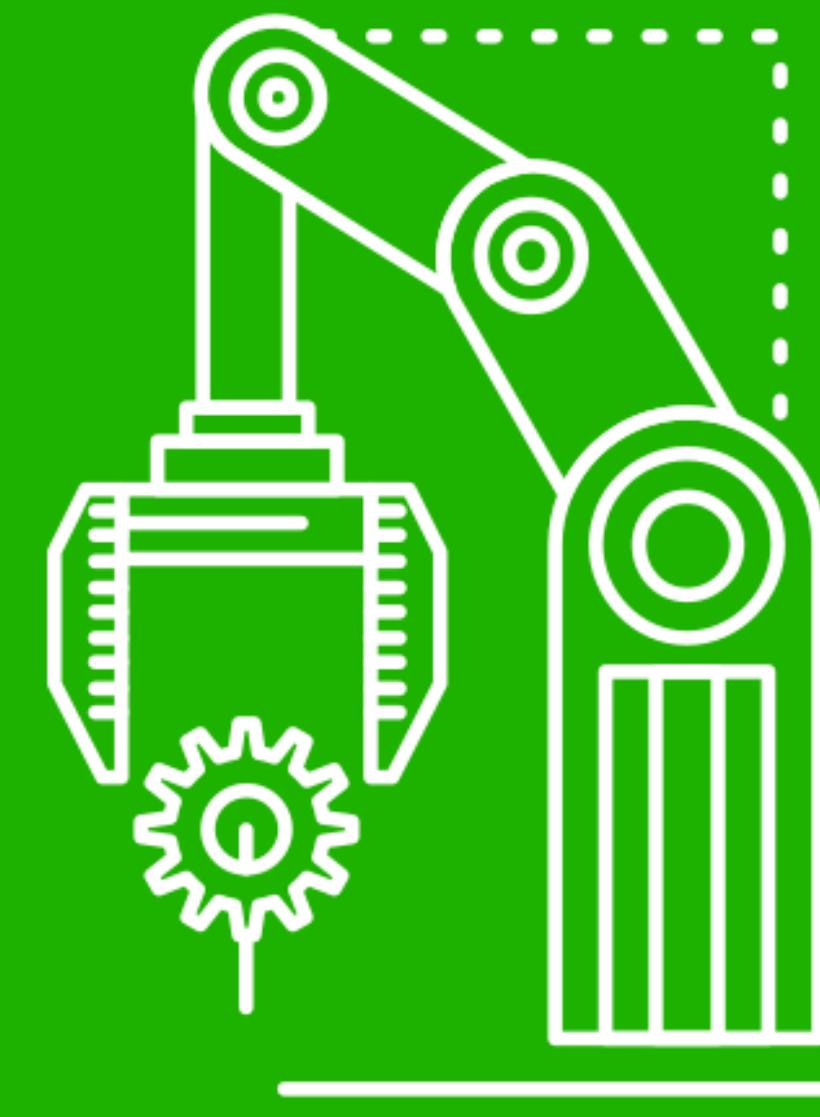


Nikolaev et al., *npj Comput. Mater.*, 2016

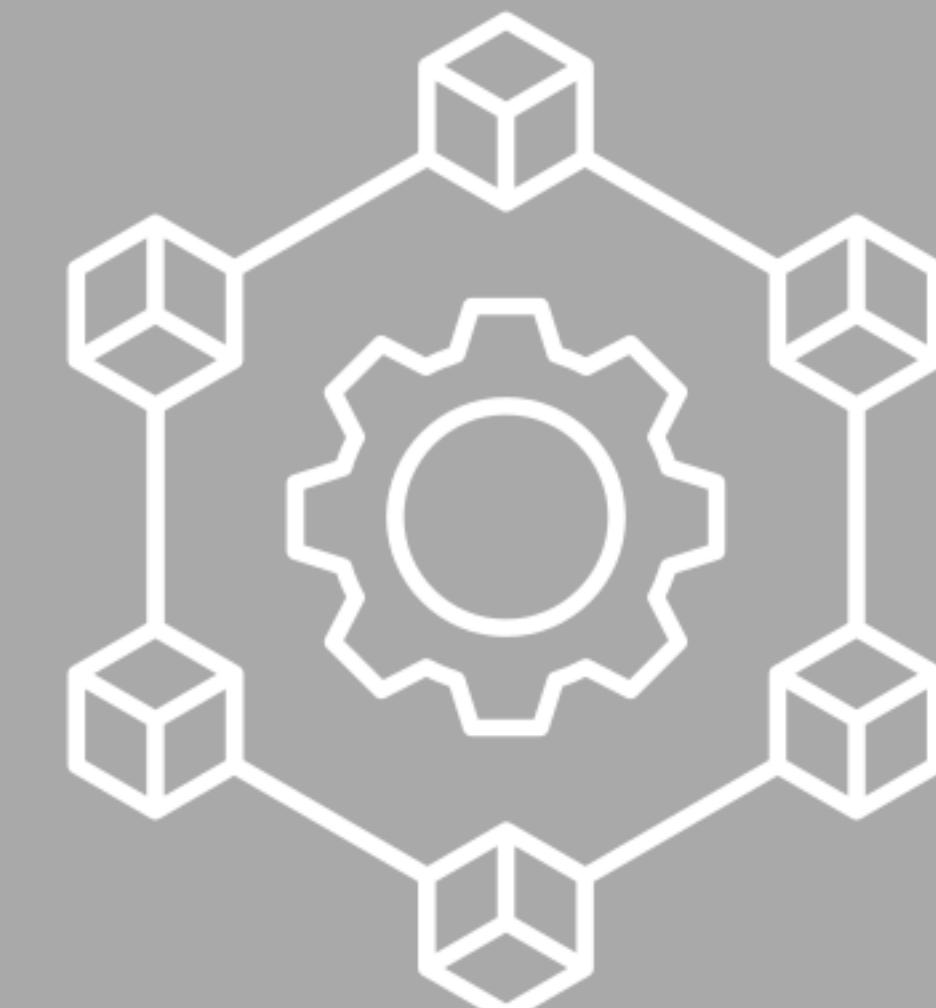
NEED FOR GREATER
FLEXIBILITY, RELIABILITY, AND
TRUSTWORTHINESS
OF AI ALGORITHMS



EXPERIMENTS ARE DIFFICULT
TO AUTOMATE



LACK OF AI-RELATED DATA
INFRASTRUCTURE AND BEST
PRACTICE

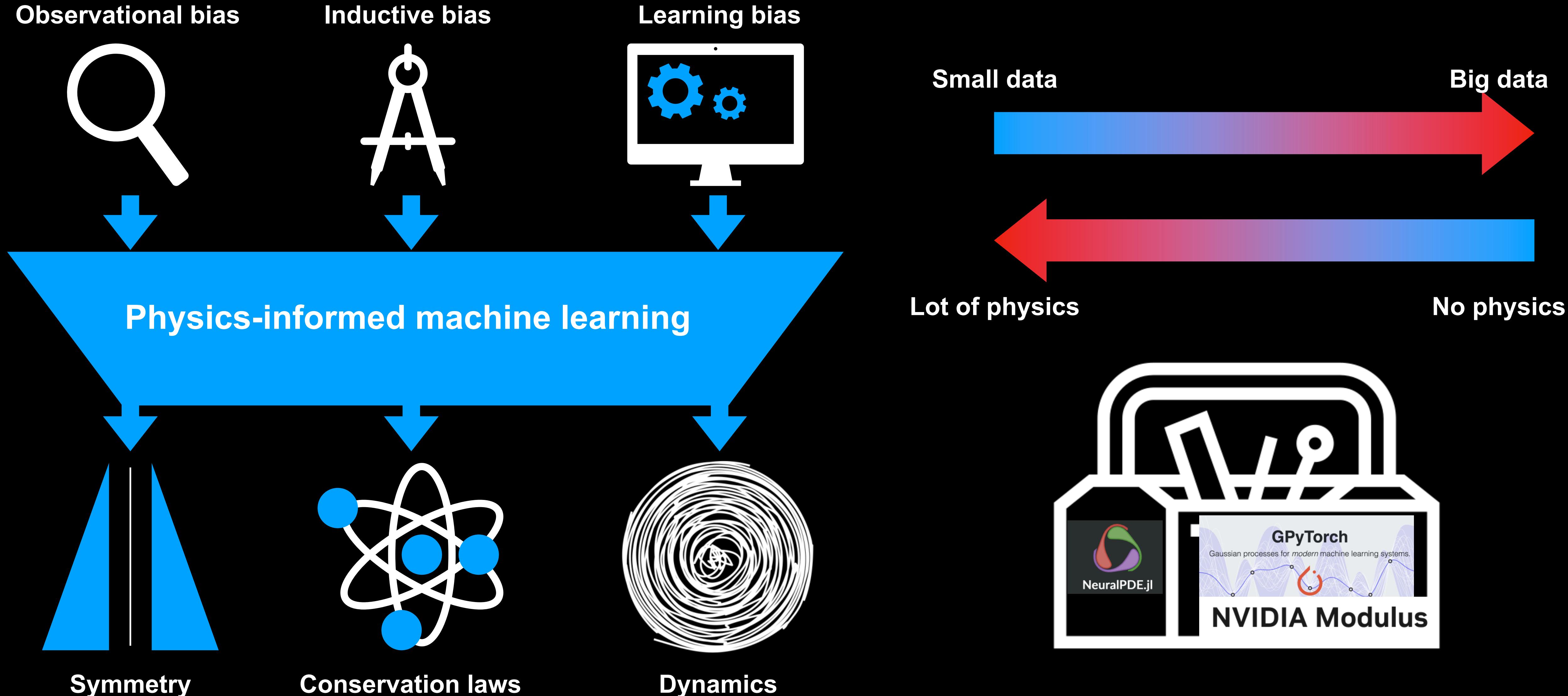


LACK OF LARGE SCALE
INVESTMENTS IN AI FOR
MATERIALS AND MATERIALS
PROCESSING





WE NEED TO INTEGRATE PHYSICAL LAWS AND DOMAIN KNOWLEDGE BY TEACHING ML MODELS THE GOVERNING PHYSICAL RULES

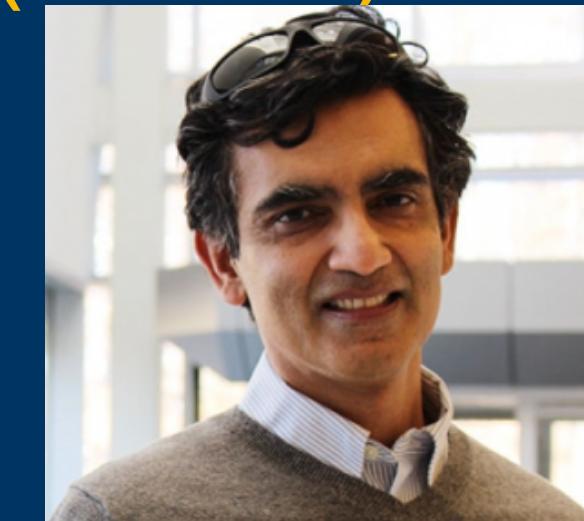
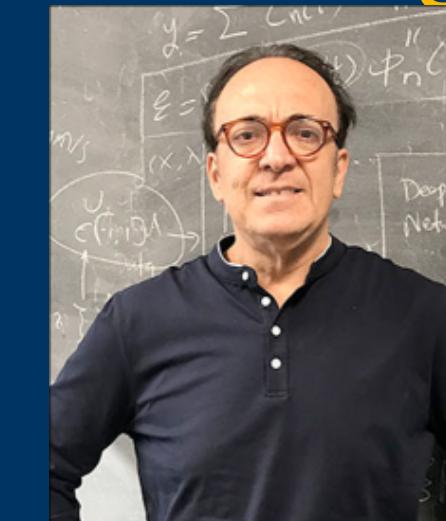




Track for *BeyondFingerprinting* Day 2: Wed. 9:15AM—11:30AM MT

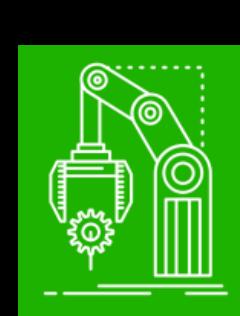
Kardianakis (Brown), Arroyave (TAMU),

Dingreville (SNL), Martinez (SNL), Garikipati (UMich)



- Will we ever develop models that have robust extrapolation performance into untested regimes?
- Are PINNs predictions meaningless without SME or experimental validation?
- What is the path forward to develop off-the-shelf ML models that can replace known and trusted analyses and characterization capabilities?





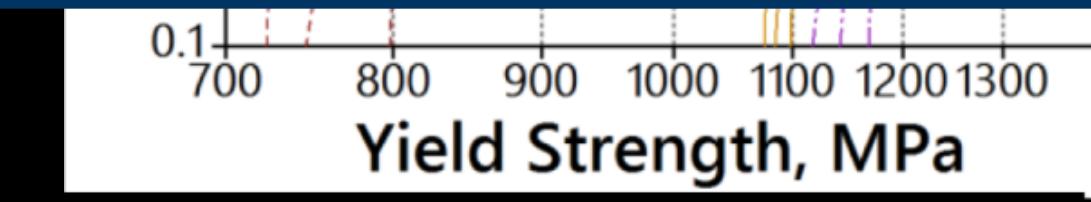
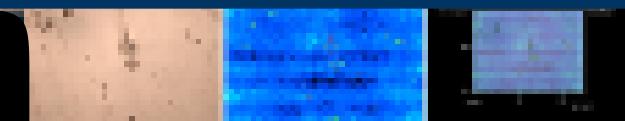
WHERE CAN WE REPLACE COSTLY “GOLD-STANDARD” CHARACTERIZATION MEASUREMENTS
WITH **Low-Cost/Unconventional Methods** THAT CONTAINS APPROPRIATE SIGNATURES?

Track for *BeyondFingerprinting* Day 2: Wed. 12:10PM—2:25PM MT

Taheri (JHU), Kalinin (UTK/Amazon),
Ophus (LBNL), Takeuchi (U Maryland), Fowler (SNL)



- Will materials scientists ever be able to trust surrogate, low-cost/unconventional measurements?
- Data-rich/data-poor dilemma: will materials science ever produce meaningful data sufficient to support AI/ML analysis?
- Will data produced by AI/ML analysis lose interpretability for SME?



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Is AI/ML ALGORITHMS AND AUTOMATION ALL WE NEED?

AUTONOMOUSLY

Track for *BeyondFingerprinting* Day 2: Wed. 2:25PM—4:55PM MT

Hattrick-Simpers (U Toronto), Stebner (GT),
Warren (NIST), Saal (Citrine), Mehta (SLAC)



- How should we measure payoffs from ML investments, and when can we expect these payoffs to emerge?
- Will these approaches bridge from lab-scale to manufactured products with improved performance and reliability?
- Will human interpretation of ML-generated data always be a bottleneck during materials discovery?

BEYONDFINGERPRINTING TRACK DAY 2: AI WILL NOT REPLACE MATERIAL SCIENTISTS BUT MATERIALS SCIENTISTS WHO USE AI WILL REPLACE THOSE WHO DON'T



Computer scientist

Program manager