

The modern-day blacksmith

Gareth Conduit

Model **sparse** datasets

Exploit **property-property** relationships

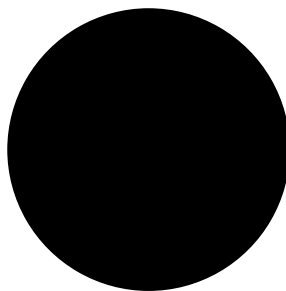
Merge data, computer simulations, and physical laws

Reduce costly experiments to **accelerate** discovery

Commercialized as Alchemite™ by **Intellegens**

Black box machine learning for materials design

Composition



Properties

Defects



Fatigue



Strength



Train the machine learning

3870454990176143
6412046921823707
6488783419689686
1181558158737756
4102468240322648
3464176636980663
7857681349204530
2240819727856471
9839630878154322
1166912246415911

Composition



6488348704023749
8012754834215237
2272034535605225
8174681581845845
6224043080254684
1132084110885053
4535407535613613
8023131978697758
5488943723909634
8211320709698126
2677601866125614
0045818967704685
6985375313052341
2052413617013298
5560923129005660

Properties

Defects

Fatigue

Strength



Machine learning predicts material properties

Composition



Properties

Defects



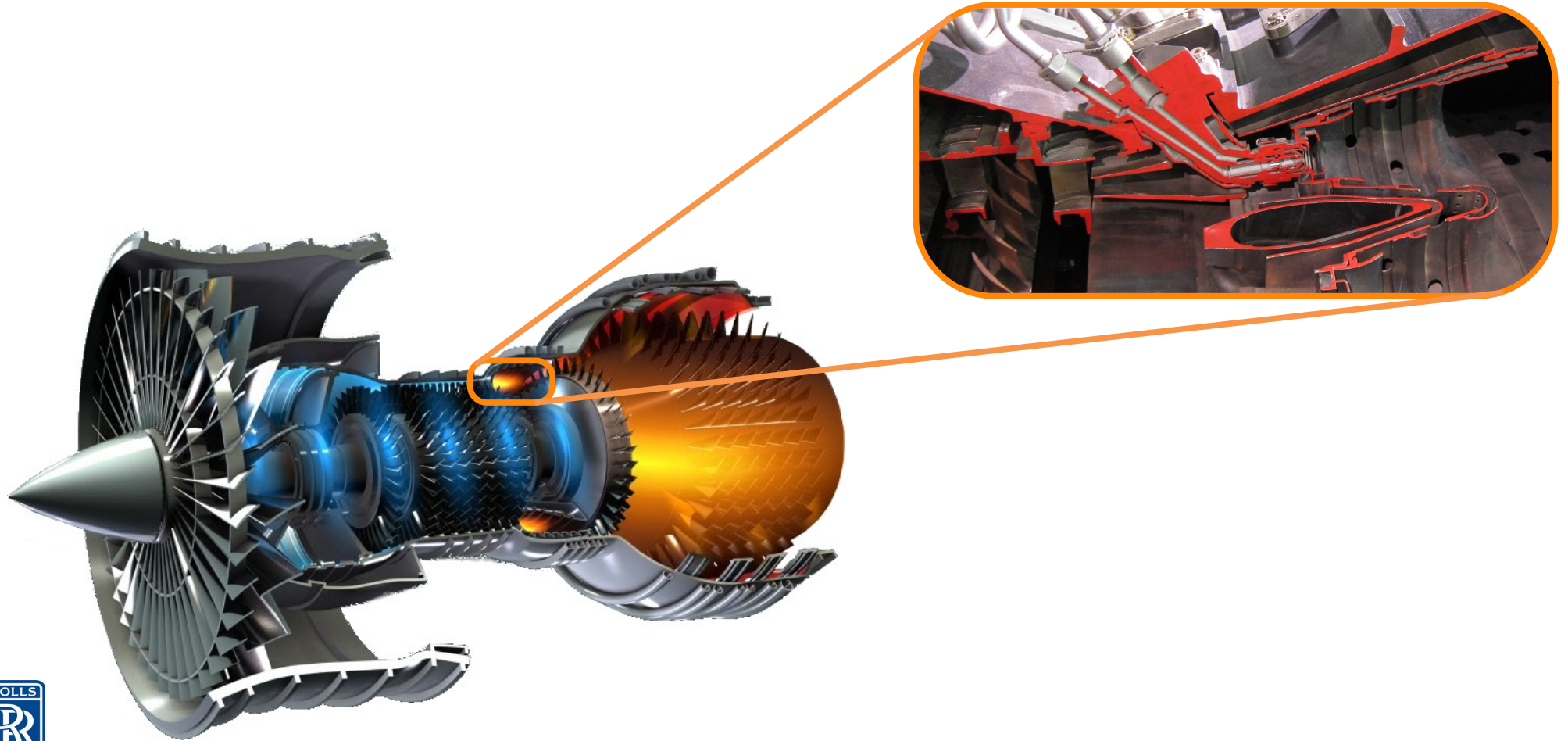
Fatigue



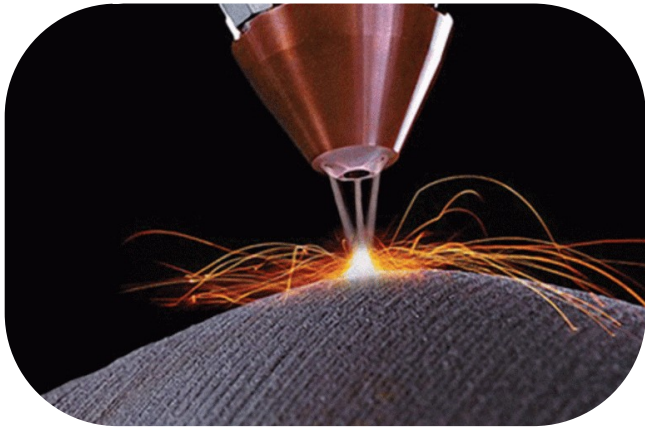
Strength



Combustor in a jet engine



Defects form during printing



Laser

Data available to model defect density

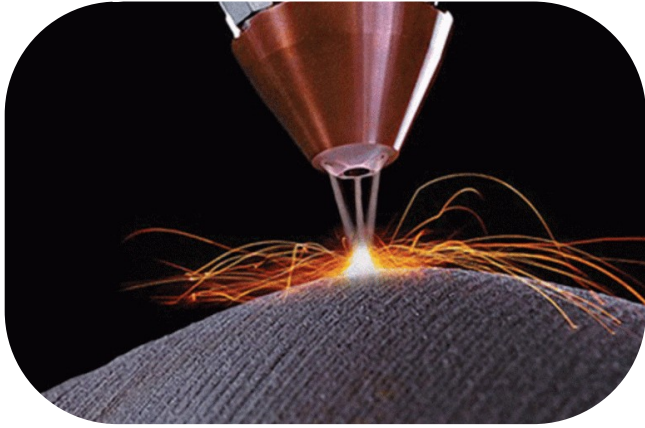


Composition and heat treatment space **30** dimensions

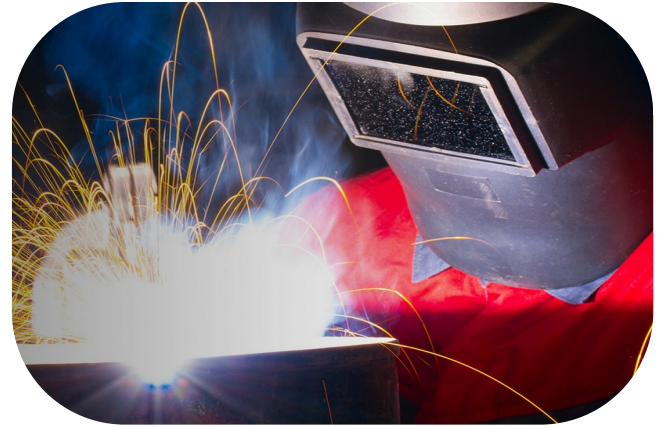
Requires **31** points to fit a hyperplane

Just **10** data entries available to model defect density

Ability for printing and welding are strongly correlated



Laser



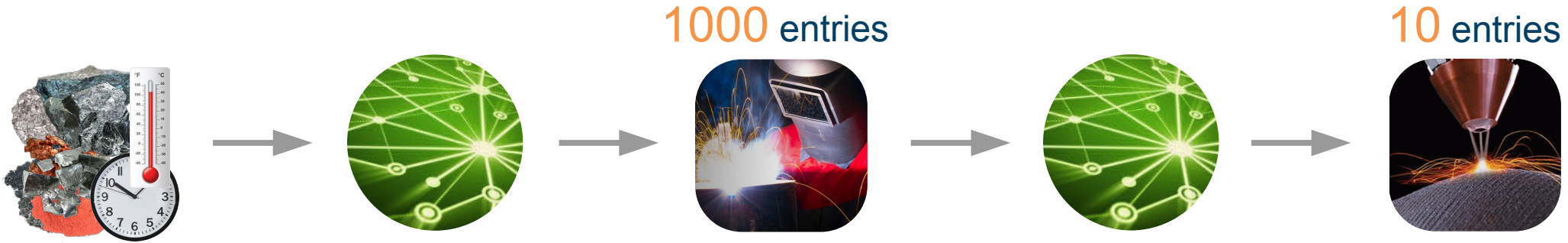
Electricity

First predict weldability



Use **1000** weldability entries to understand complex composition → weldability model

Use weldability to predict defects formed



Use **1000** weldability entries to understand complex composition → weldability model

10 defects entries capture the simple weldability → defect relationship

Two interpolations give composition → defects **extrapolation**



Use CALPHAD to predict strength



Use **100,000** CALPHAD results to model complex composition → phase behavior

500 strength entries capture the phase behavior → strength relationship

Two interpolations aid the composition → strength **extrapolation**



Target properties

Elemental cost	< 25 \$kg ⁻¹
Density	< 8500 kgm ⁻³
γ' content	< 25 wt%
Oxidation resistance	< 0.3 mgcm ⁻²
Defects	< 0.15% defects
Phase stability	> 99.0 wt%
γ' solvus	> 1000°C
Thermal resistance	> 0.04 KΩ ⁻¹ m ⁻³
Yield stress at 900°C	> 200 MPa
Tensile strength at 900°C	> 300 MPa
Tensile elongation at 700°C	> 8%
1000hr stress rupture at 800°C	> 100 MPa
Fatigue life at 500 MPa, 700°C	> 10 ⁵ cycles

Composition and processing variables

Cr 19%



Co 4%



Mo 4.9%



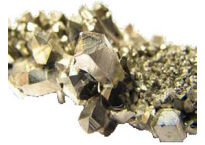
W 1.2%



Zr 0.05%



Nb 3%



Al 2.9%



C 0.04%



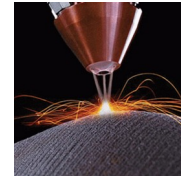
B 0.01%



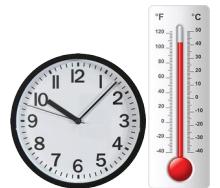
Ni



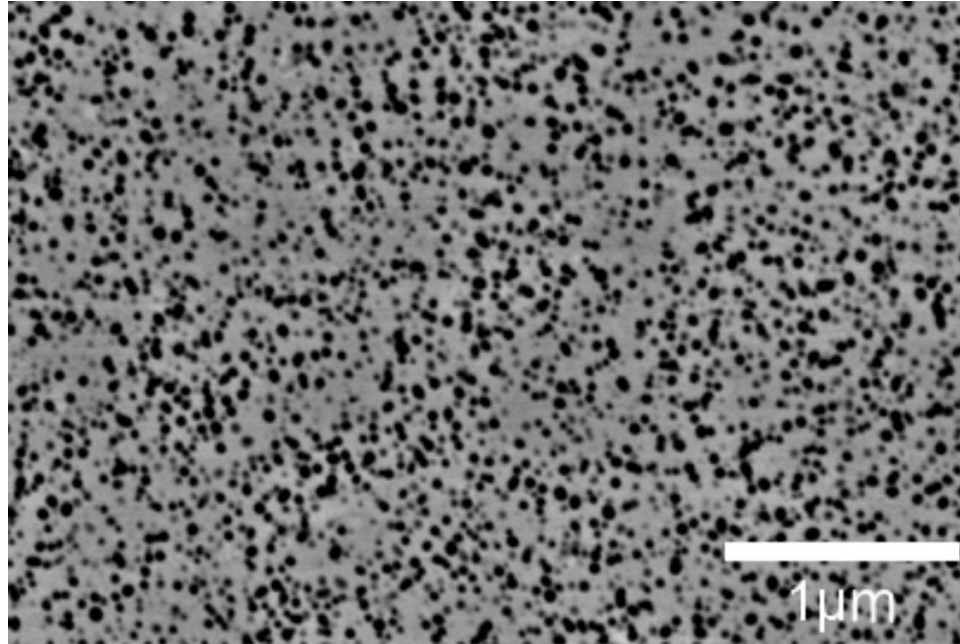
Expose 0.8



T_{HT} 1300°C

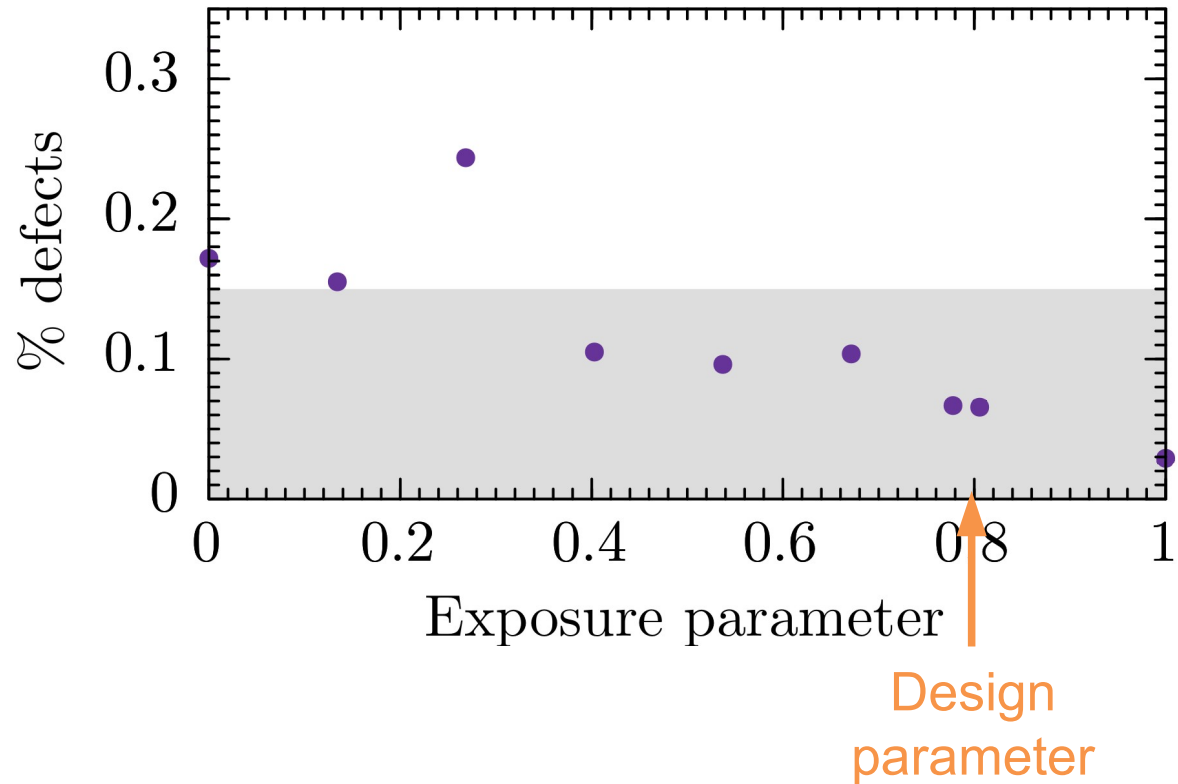


Microstructure



Probabilistic neural network identification of an alloy for direct laser deposition
Materials & Design 168, 107644 (2019)

Testing the defect density

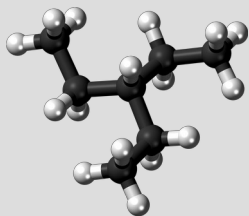




Journal of Computer-Aided
Molecular Design **35**, 112501140 (2021)



NASA Technical
Memorandum
20220008637 (2022)



nature
machine intelligence

REVIEW ARTICLE

<https://doi.org/10.1038/s42256-020-0156-7>



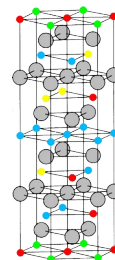
Predicting the state of charge and health of batteries using data-driven machine learning

Man-Fai Ng¹, Jin Zhao², Qingyu Yan², Gareth J. Conduit³ and Zhi Wei Seh⁴



Fluid Phase Equilibria
501, 112259 (2019)

Journal of Chemical Physics
153, 014102 (2020)



Johnson Matthey
Technology Review
66, 130 (2022)



Development of methodology



2013

Multiple
properties for
Rolls Royce
engines

Development of methodology



2013

2014

Multiple properties for Rolls Royce engines

Property-property correlations with Rolls Royce and BP

Development of methodology



*Concurrent
materials design*



2013

2014

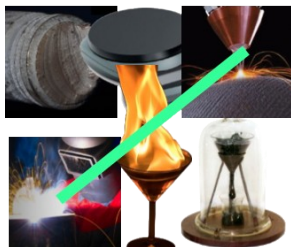
2015

Multiple
properties for
Rolls Royce
engines

Property-
property
correlations
with Rolls
Royce and BP

Royal Society
University
Research
Fellowship

Development of methodology



*Concurrent
materials design*



2013

2014

2015

2016

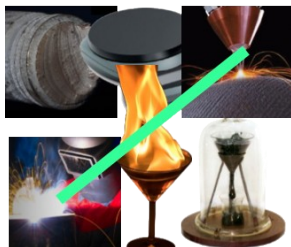
Multiple
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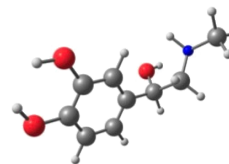
Royal Society
University
Research
Fellowship

Experiment-
simulation
correlations
with Samsung

Development of methodology



Concurrent materials design



2013

2014

2015

2016

2017

Multiple properties for Rolls Royce engines

Property-property correlations with Rolls Royce and BP

Royal Society University Research Fellowship

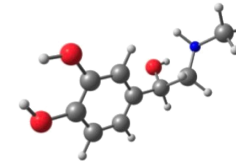
Experiment-simulation correlations with Samsung

Drug discovery study with etherapeutics

Development of methodology



Concurrent materials design



2013

Multiple properties for Rolls Royce engines

2014

Property-property correlations with Rolls Royce and BP

2015

Royal Society University Research Fellowship

2016

Experiment-simulation correlations with Samsung

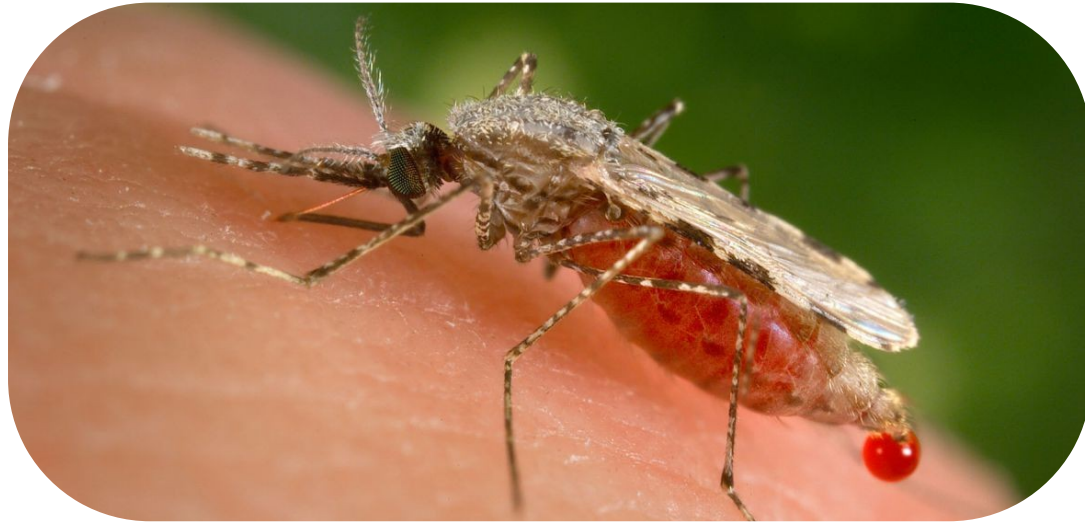
2017

Drug discovery study with e-therapeutics

2018

Founding of Intellegens

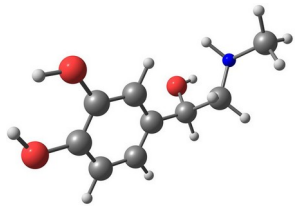
Open Source Malaria contest



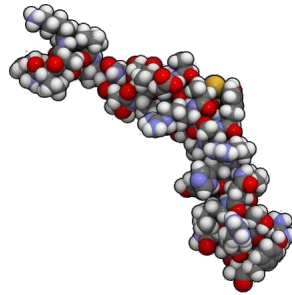
OPEN SOURCE MALARIA

Looking for New Medicines

Action of a drug



Drug

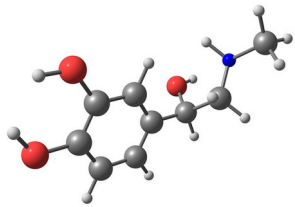


Protein

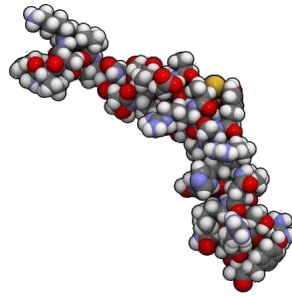


Effect

Action of a drug



Drug

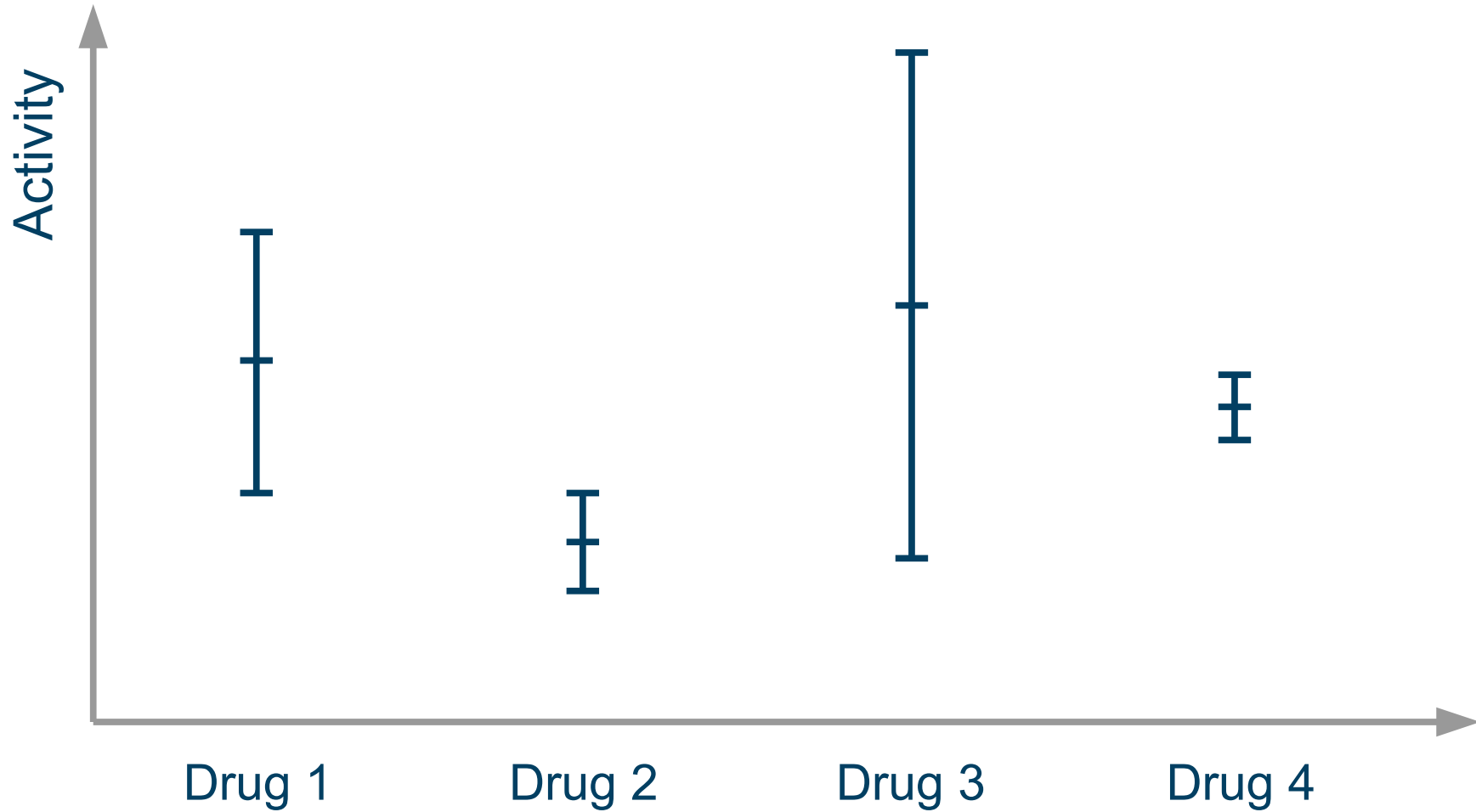


Protein

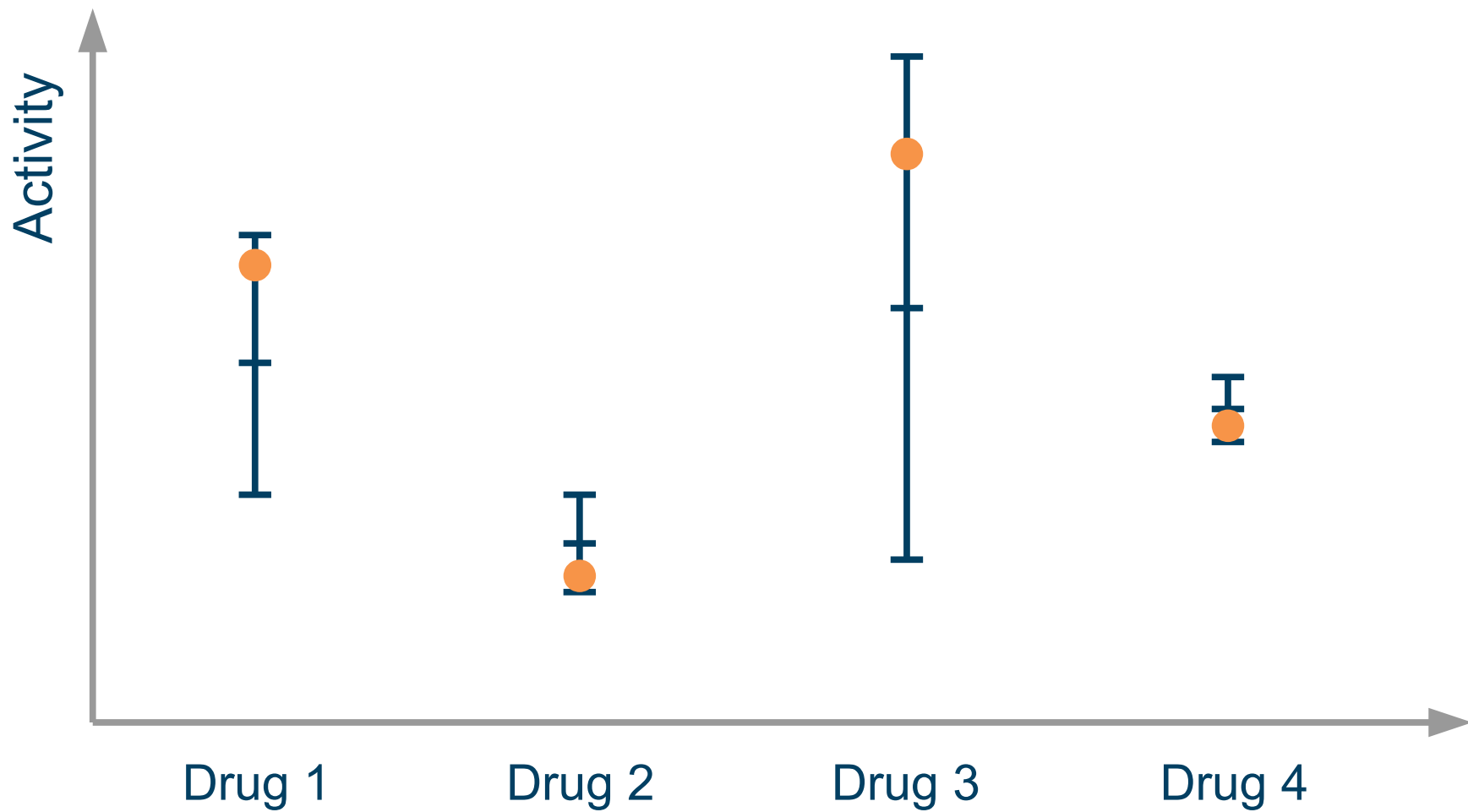


Effect

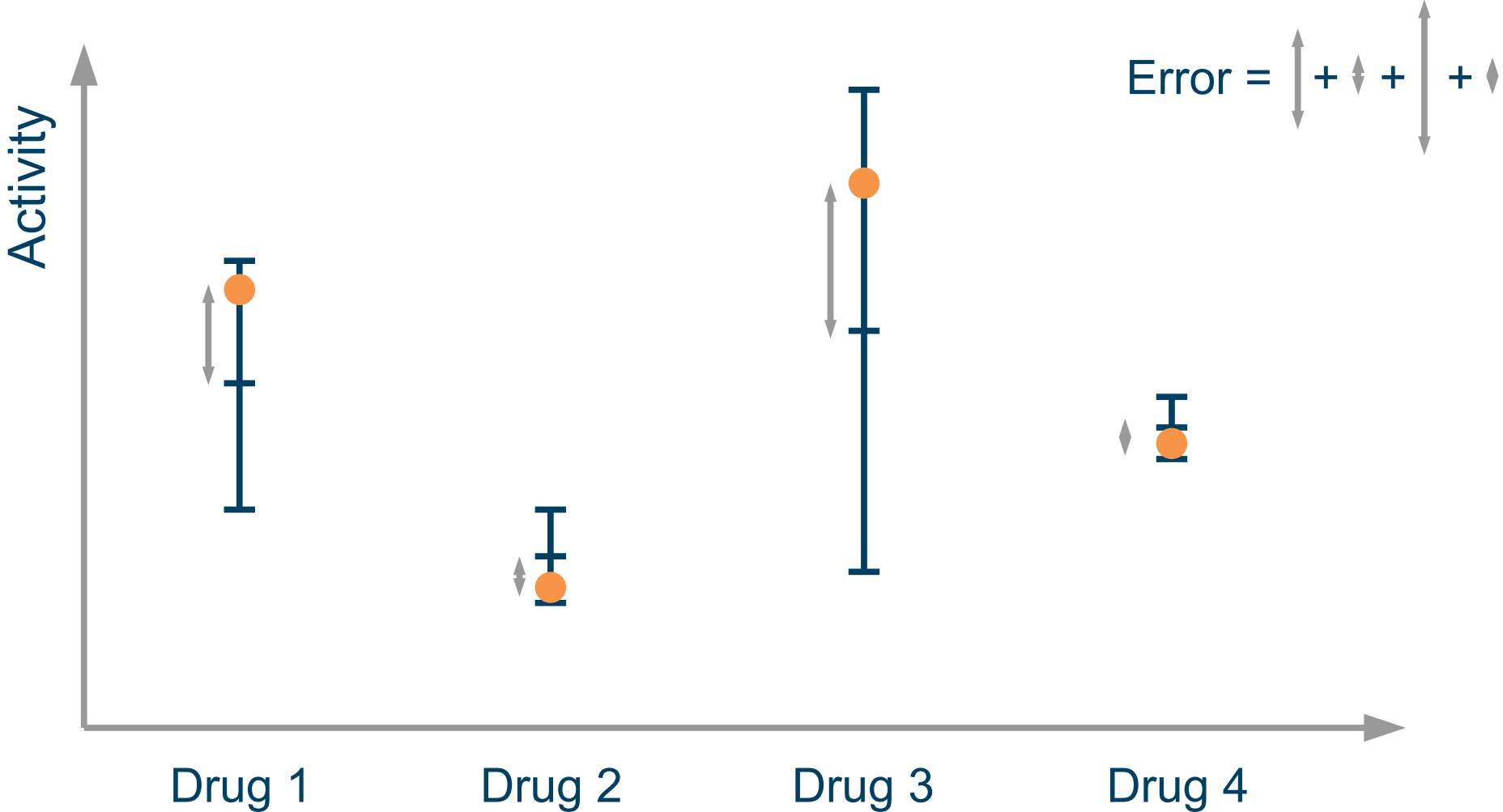
Predictions have an uncertainty



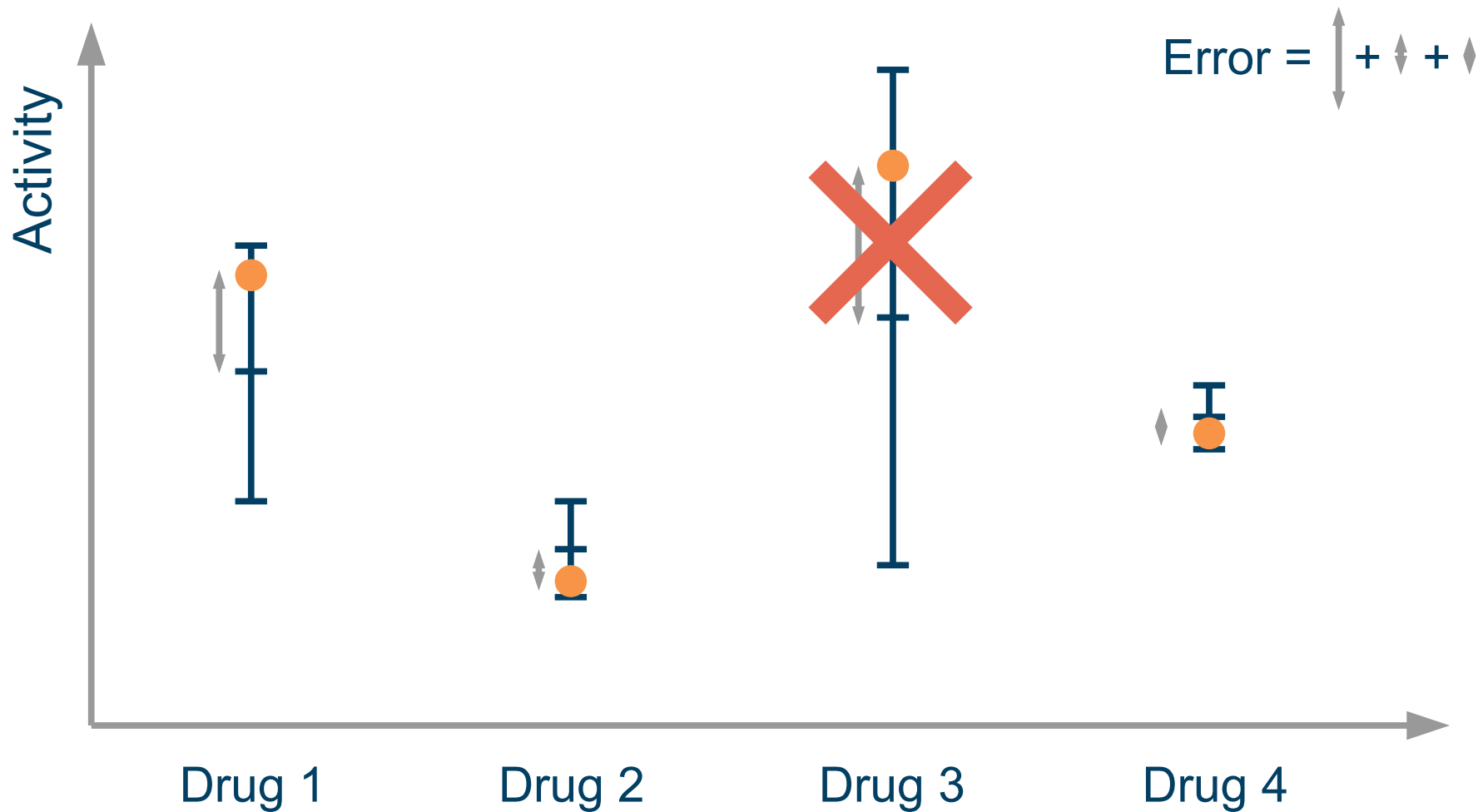
Validation data typically within one standard deviation



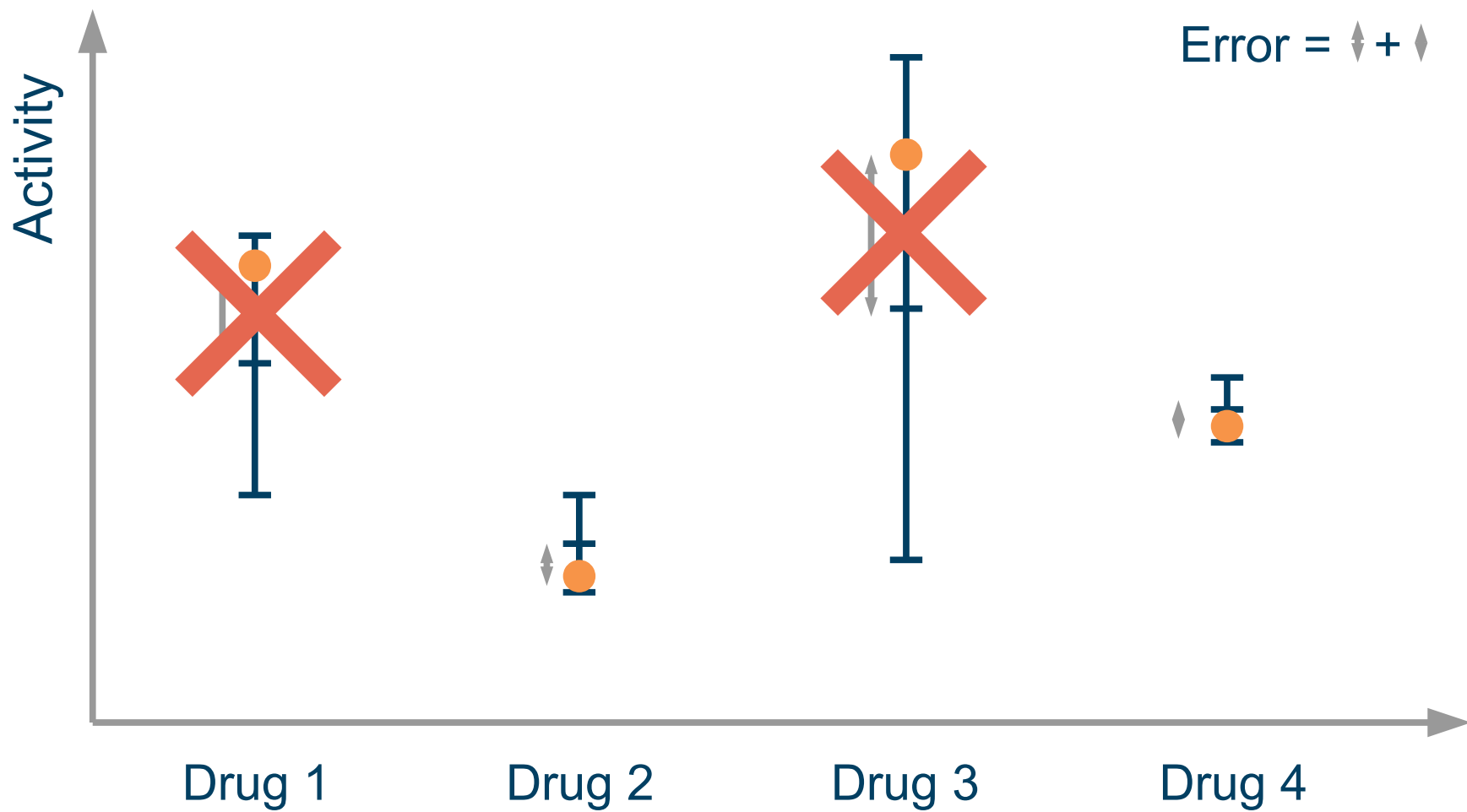
Accuracy R^2 metric calculated with difference from mean



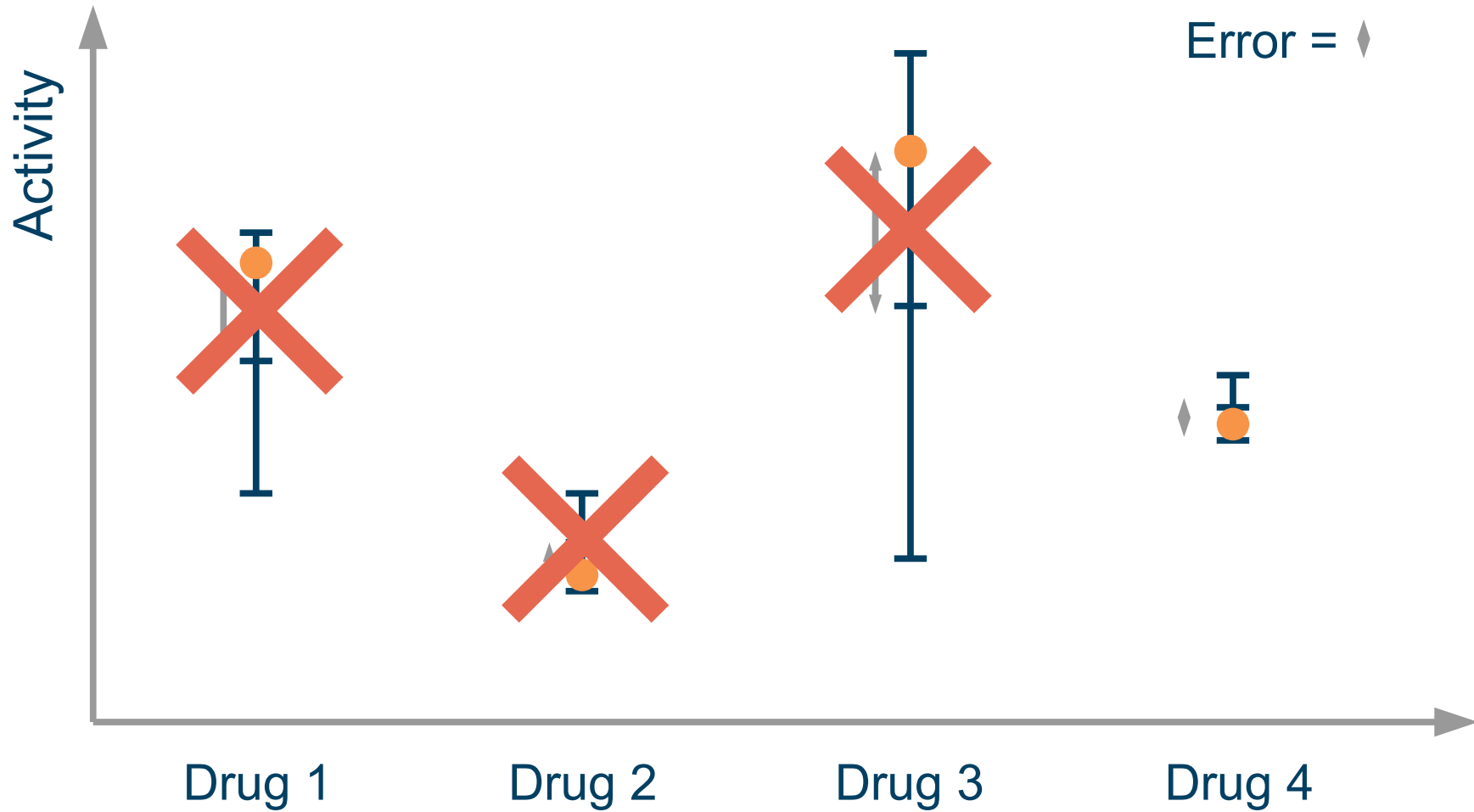
Impute 75% of data with smallest uncertainty



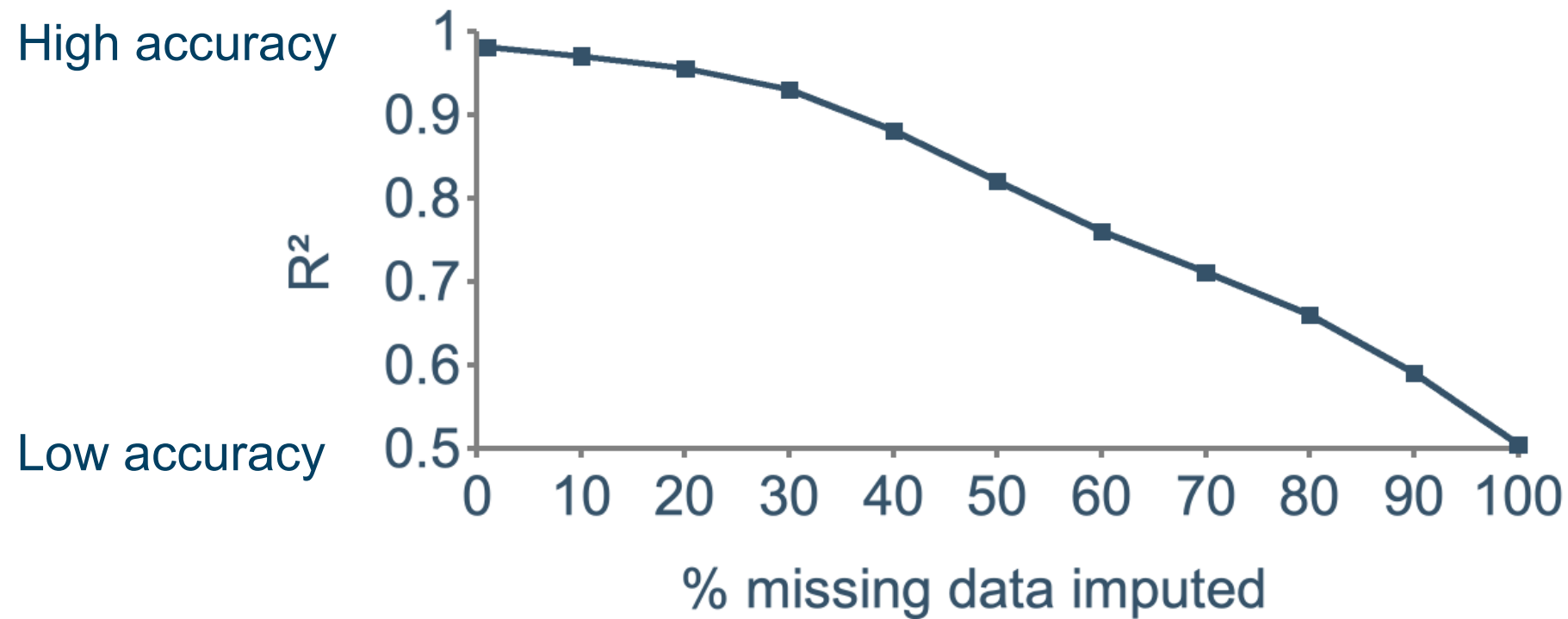
Impute 50% of data with smallest uncertainty



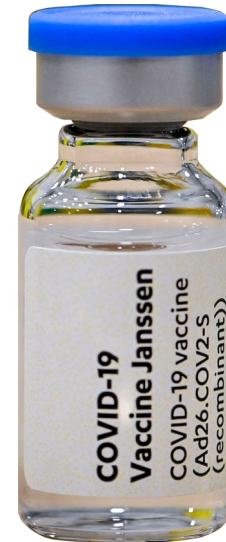
Impute 25% of data with smallest uncertainty



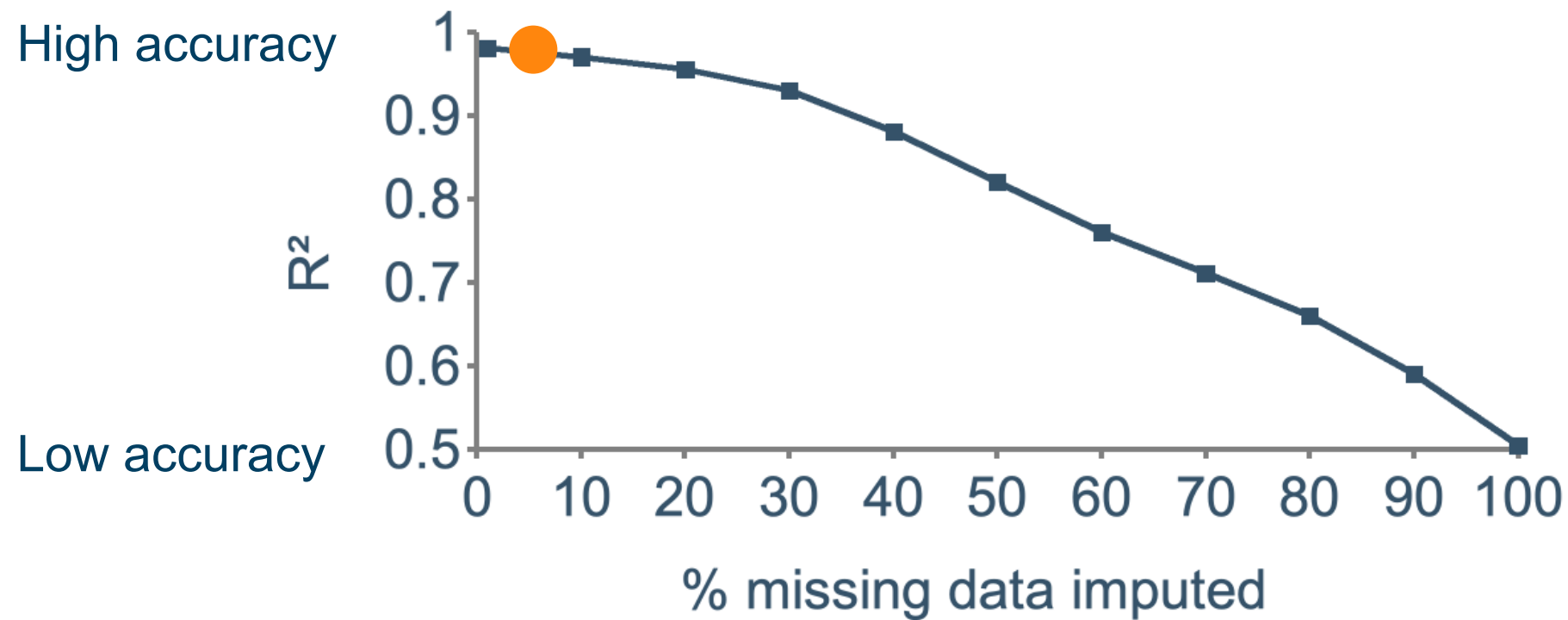
Improved performance by exploiting uncertainty



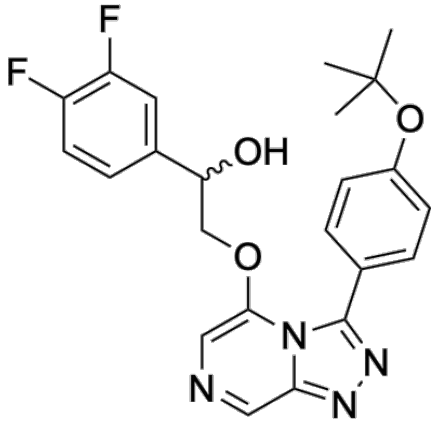
Different drugs can treat the same ailment



Focus on compounds with low uncertainty



Open Source Malaria experimental validation

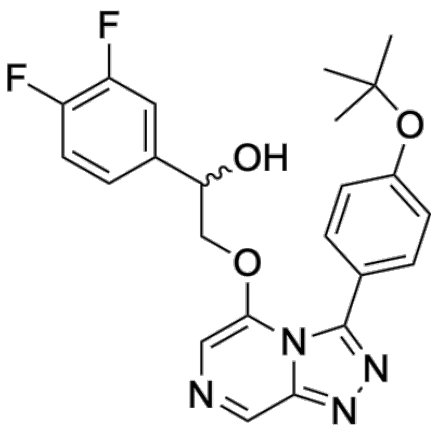


Optibrium & Intellegens

0.647 μM

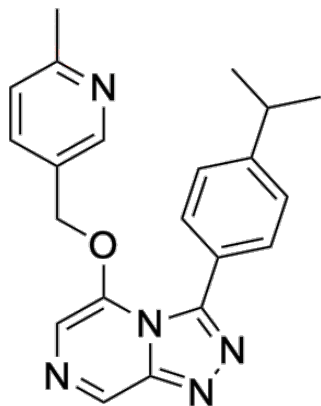
Journal of Medicinal Chemistry 64, 16450 (2021)

Open Source Malaria other compounds



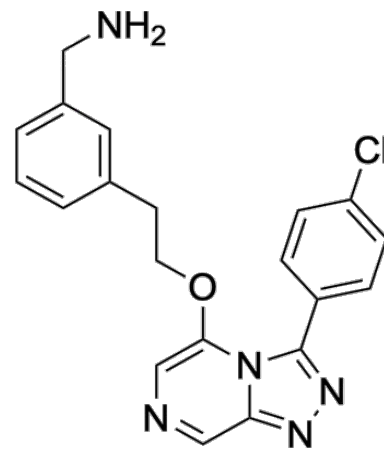
Optibrium & Intellegens

0.647 μM



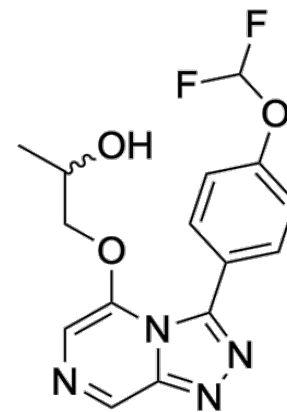
Davy Guan

>25 μM



Exscientia

10.9 μM



Molomics

>25 μM

Commercialization

 therapeutics



2018

Transfer
contracts from
University

Commercialization

e-therapeutics



2018

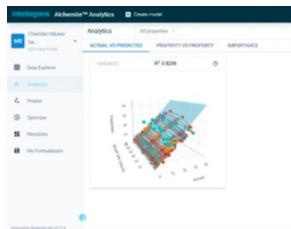
2019

Transfer
contracts from
University

Consultancy
work

Commercialization

e-therapeutics



2018

2019

2020

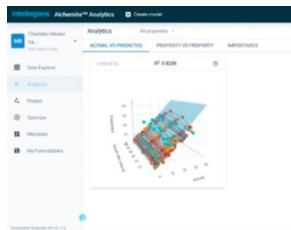
Transfer contracts from University

Consultancy work

Launch Alchemite Analytics™ product

Commercialization

e-therapeutics



optibrium



2018

2019

2020

2021

Transfer
contracts from
University

Consultancy
work

Launch
Alchemite
Analytics™
product

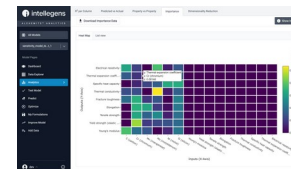
Launch
Cerella™
product with
Optibrium

Commercialization

e-therapeutics



optibrium



2018

2019

2020

2021

2021

Transfer contracts from University

Consultancy work

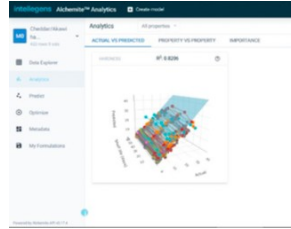
Launch Alchemite Analytics™ product

Launch Cerella™ product with Optibrium

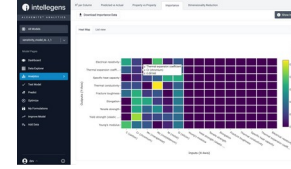
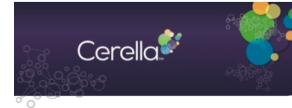
Enterprise licenses & many case studies

Commercialization

e-therapeutics



optibrium



ANSYS / GRANTA



2018

2019

2020

2021

2021

2022

Transfer contracts from University

Consultancy work

Launch Alchemite Analytics™ product

Launch Cerella™ product with Optibrium

Enterprise licenses & many case studies

Launch product with ANSYS Granta

Merge computer simulations with experimental data and exploit **property-property** relationships to circumvent **missing data**

Designed and **experimentally verified** alloy for direct laser deposition

Exploited **uncertainty** to predict most probable drug

Generic approach applied to materials, chemicals, pharmaceuticals, and beyond

Commercialized as Alchemite™ by **Intellegens**