

intellegens

DATA - DRIVEN DISCOVERY



7 examples of how materials & chemicals companies innovate using Alchemite machine learning



1. Alloys in aerospace
2. Batteries
3. Antimalarial drug discovery
4. 3D printing
5. Data validation
6. Forging hammers
7. Alkanes for lubricants



Alloys in aerospace





Challenge

Alloys must have good performance at **high temperatures**

Commercial superalloys do not have the **optimal balance of properties**

Long and **iterative** research and development process is expensive



Solution

Machine learning tool **Alchemite**

Augment **experimental data** with **thermodynamic predictions**

Incorporate **uncertainty** to design alloys with the highest likelihood to **meet design targets**

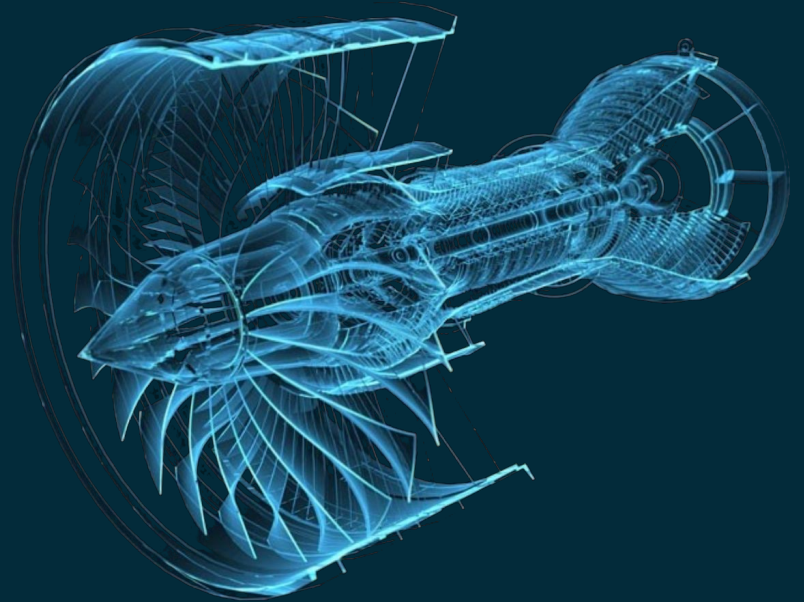


Outcome

New nickel-base superalloy
that met **11** targets

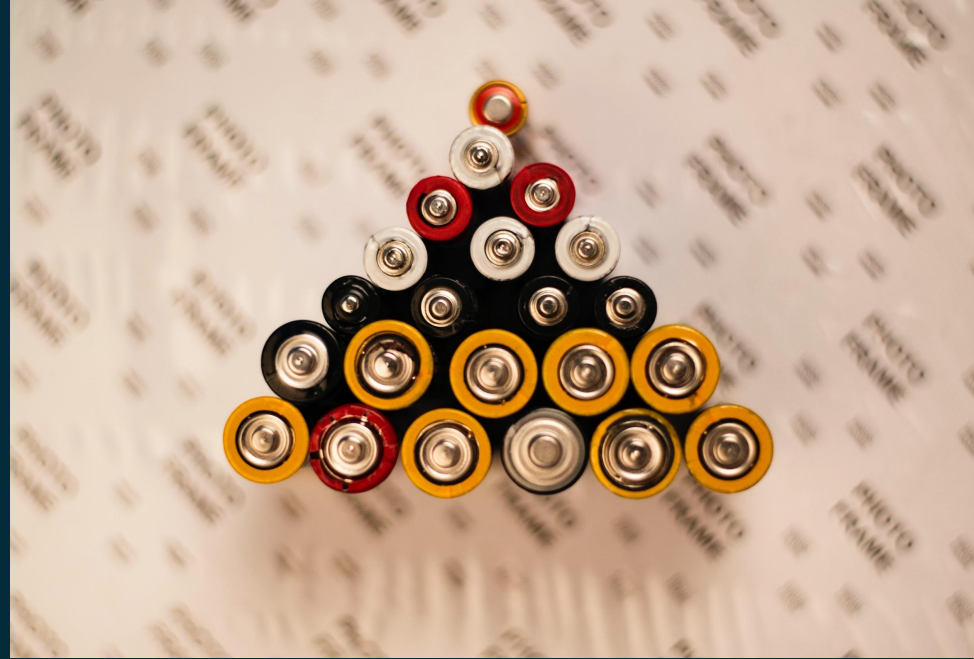
Properties were
experimentally verified

Reduced development **cost**
by £10m





Batteries





Challenge

Performance, cost, and safety of batteries will drive the development of **electric vehicles**

Materials and **battery management** system can both be improved by machine learning

Further research into battery chemistries requires **more complicated** battery dynamics



Solution

Machine learning to design the **cathode**, **anode**, and **electrolyte**

Predictive models for **battery management** systems to alleviate range anxiety





Outcome

Ran virtual experiments to **focus the search** for new materials

Reduced development and fabrication **costs** while improving key battery metrics

Predict remaining useful life, state of health, and state of charge



Antimalarial drug discovery





Challenge

Identify a new molecule against a novel **antimalarial** mechanism of action

Available data covers multiple assays, but only a **small fraction** have been measured





Solution

In blind test Alchemite **accurately predicted** activity of compounds against novel mechanisms of action

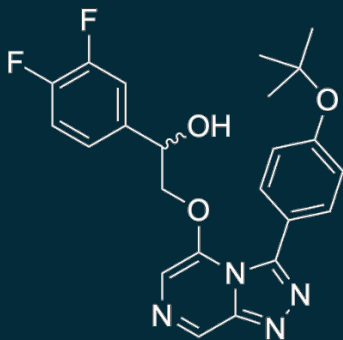
Outperforms conventional QSAR and other machine learning models



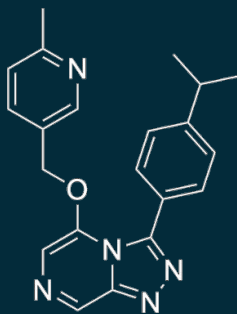
Outcome

Alchemite confidently predicted a compound generated automatically by Optibrium's StarDrop™ software

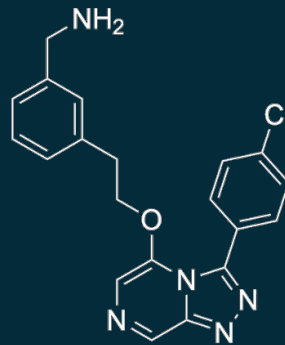
Compound synthesised and tested by OSM group, only entry that demonstrated **potency against the target**



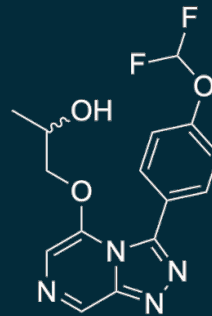
Optibrium/Intellegens
0.647 μM



Davy Guan
>25 μM



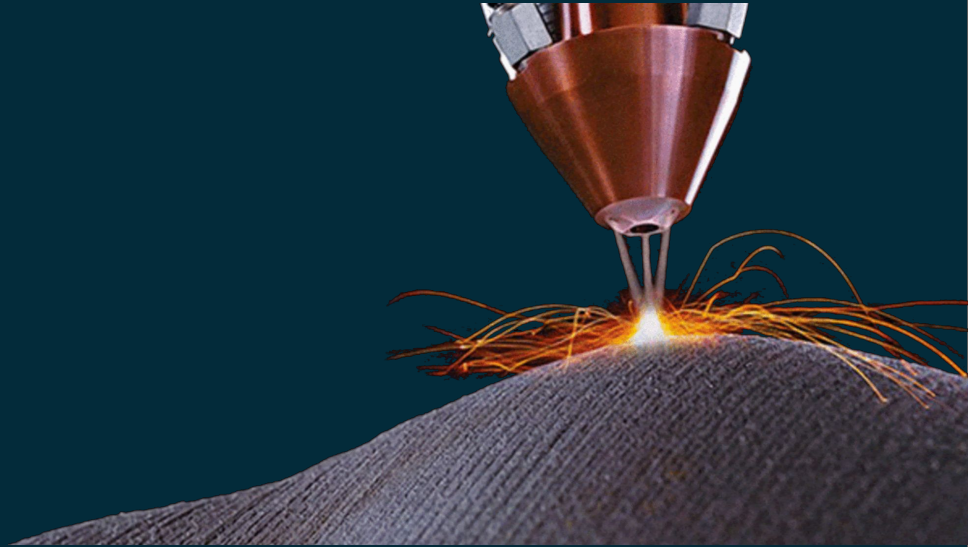
Exscientia
10.9 μM



Molomics
>25 μM



3D printing





Challenge

Additive manufacturing is a **new process** that requires **new materials**

Ability to print materials is **poorly understood**

Direct laser deposition (a 3D printing method) has only been applied to **ten** nickel alloys



Solution

Alchemite exploited **property-property** relationships to capture of new insights

Juxtaposed historical **welding** and sparse **direct laser deposition** data

Optimized alloy for direct laser deposition and also fulfil other target properties



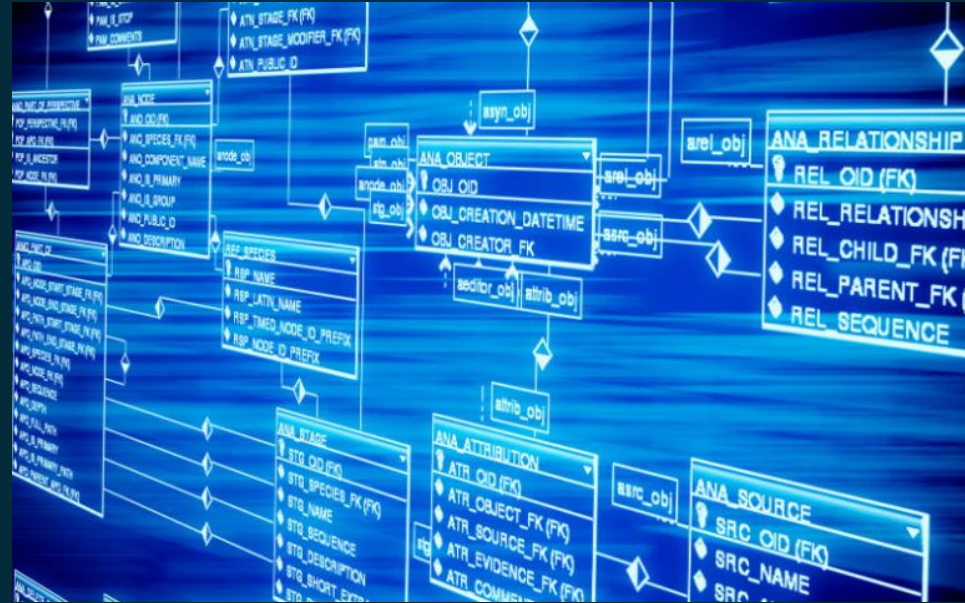
Outcome

Alloy for direct laser deposition **experimentally verified**

Alchemite **saved 15 years** worth of development



Data validation





Challenge

Companies that sell curated databases must have efficiently **identified** and **fixed** all erroneous entries

Errors in central database misguides research programs and will lead to **inaccurate** machine learning

Triage data introduced to a central database



Solution

Alchemite models all available data

Automatically identifies **unexpected** values to **focus attention** where additional investigation is most effective



Outcome

Identified errors including typos, errors, or genuine outliers

Automate tool to triage submitted data

intellegens Alchemite™ Analytics Create model demo

example_steels 205 rows 12 cols

Data Explorer SHOW PREDICTIONS HIDE OUTLIERS

	C	Mn	Si	Cr	Ni	Mo	V	Co	Heat treatment/C	YS/MPa	UTS/MPa	Elong/%
⊗	0.02	0.1	2.02	0.01	18.4	0.02	0.01	0.01	816	1461	1502	
⊗	0.02	0.1	2.02	0.01	18.4	0.02	0.01	0.01	816	1137	1185	
⊗	0.02	0.11	3.18	0.01	18.3	0.02	0.01	0.01	816	1461	1502	
⊗	0.02	0.11	3.18	0.01	18.3	0.02	0.01	0.01	816	1378	1461	
⊗	0.35	0.07	4.75	0.01	18.5	0.02	0.01	0.01	816	1495	1826	
⊗	0.35	0.07	4.75	0.01	18.5	0.02	0.01	0.01	816		2074	
⊗	0.01	2.5	0.02	0.01	18	0.02	0.01	0.01	816	1461	1502	
⊗	0.01	2.5	0.02	0.01	18	0.02	0.01	0.01	816	1481	1543	
⊗	0.02	0.01	1.93	13.9	6	0.02	0.01	6	1038		1213	
⊗	0.02	0.01	1.82	13.9	5.1	0.02	0.01	8.9	1038			17
⊗	0.02	0.01	1.86	15.9	4	0.02	0.01	9.2	1038			17
⊗	0.01	0.01	1.72	15.7	4.1	0.02	0.01	7.4	1038			16
⊗	0.01	0.01	1.99	17.5	2.1	0.02	0.01	11.8	1038			10
⊗	0.16	0.16	0.05	1.97	10.05	1	0.01	13.88	899			12
⊗	0.16	0.06	0.04	1.95	10.15	0.98	0.01	13.8	899			15
⊗	0.16	0.06	0.04	1.95	10.15	0.98	0.01	13.8	816			13.3
⊗	0.17	0.16	0.05	2.97	10.02	1.21	0.01	13.74	816			15.5
⊗	0.16	0.06	0.04	1.95	10.15	0.98	0.01	13.8	899			15.4
⊗	0.16	0.16	0.05	1.97	10.05	1	0.01	13.88	899			15.5
⊗	0.17	0.16	0.05	2.97	10.02	1.21	0.01	13.74	816			16
⊗	0.12	0.18	0.06	2.01	10.07	1	0.01	12.1	816	1452	1503	16.5
⊗	0.12	0.18	0.06	2.01	10.07	1	0.01	12.1	816	1415	1503	16.5

Powered by Alchemite API v0.15.2



Forging hammers





Challenge

Forging hammers wear and degrade through repeated use, can they be made more durable?

Machine learning explores **new alloys** to improve strength at elevated temperature



Solution

Augment **experimental data** with **thermodynamic models**

Incorporate **uncertainty** to propose molybdenum-base alloys with the highest likelihood to meet **design targets**

Alchemite suggests new precipitate chemistry with improved properties



Outcome

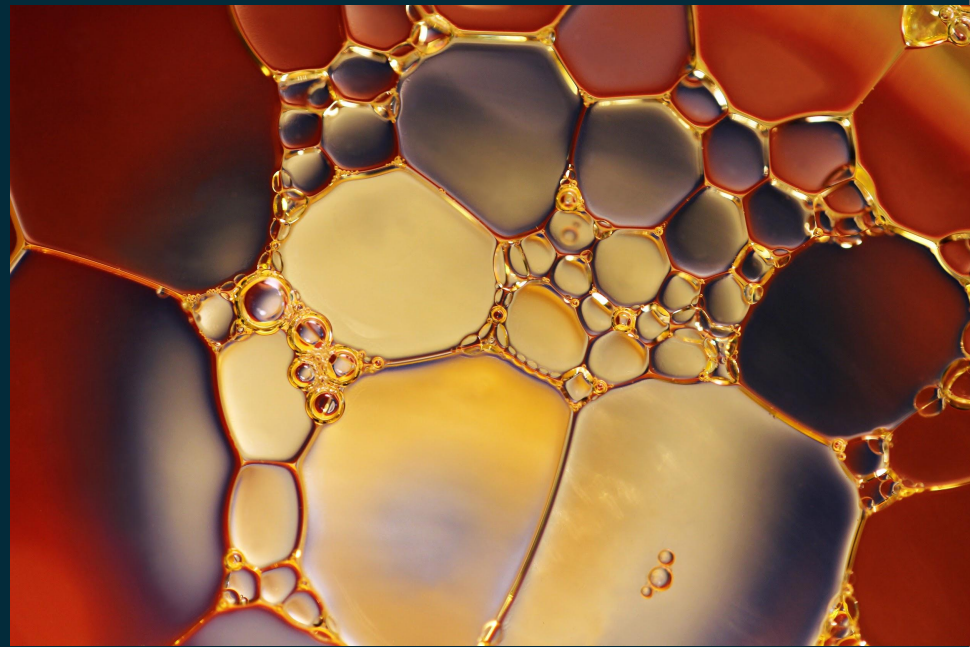
New molybdenum-base alloys found with **5-fold** reduction in experimental time and cost

Improve molybdenum-base hammers **lifetime by 30%**, reducing deployment cost and factory downtime





Alkanes for lubricants





Challenge

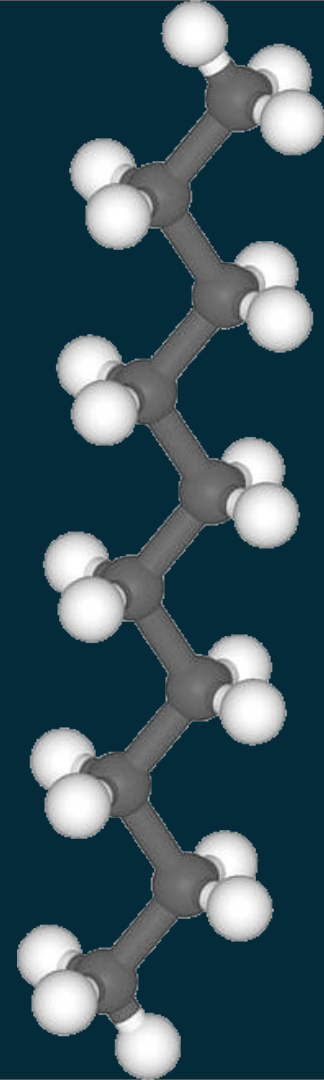
Lubricants need to be optimized for new applications such as electric vehicles

Lubricants contain **hydrocarbon** molecules

Predicting the properties of hydrocarbons facilitates the development of base oils

Solution

Exploit **property-property** correlations augmented by computer simulations to predict physical properties





Outcome

Alchemite accelerated the identification of optimal hydrocarbons **tenfold**

Alchemite estimated intractable properties including **shear viscosity** five-times more accurately than other methods



Summary

Alchemite adds value to **sparse** data problems by exploiting **property-property** correlations

Juxtapose results from **different methods**

Significant saving in **reduced cost** and **accelerated** discovery

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