

Materials discovery with artificial intelligence

Gareth Conduit

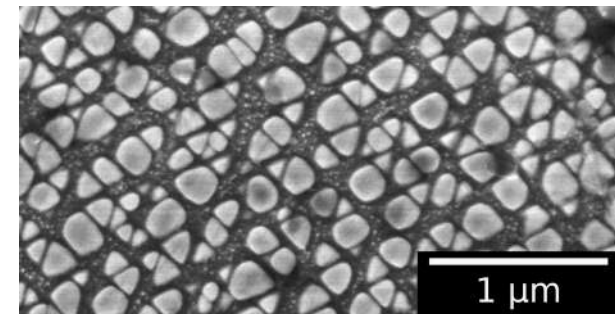
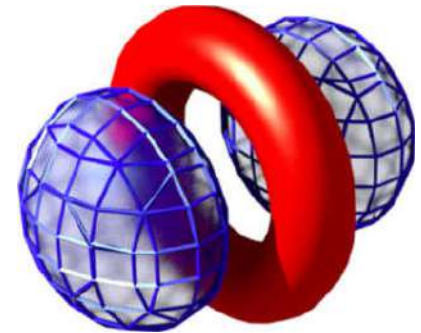
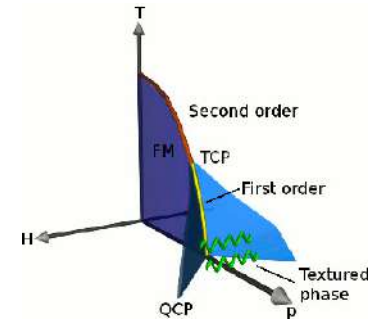
TCM Group, Department of Physics

Career

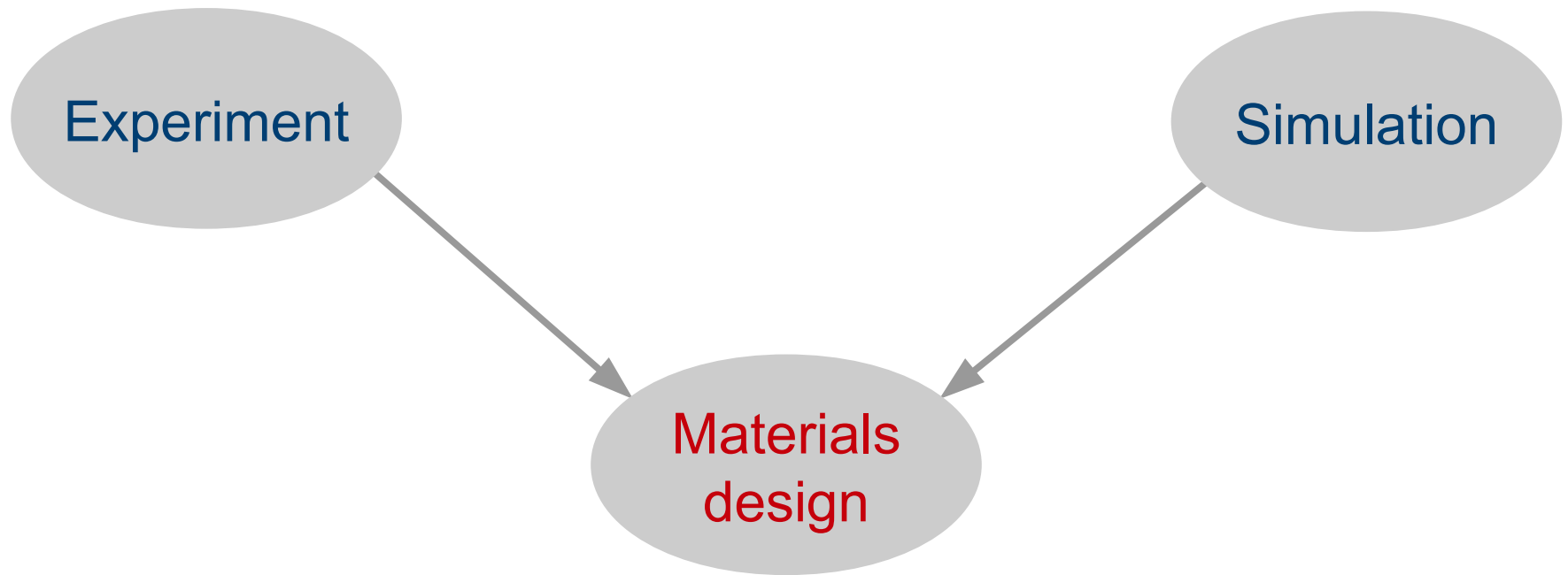
Cambridge
PhD: Ferromagnets

Weizmann Institute
Royal Commission for the
Exhibition of 1851

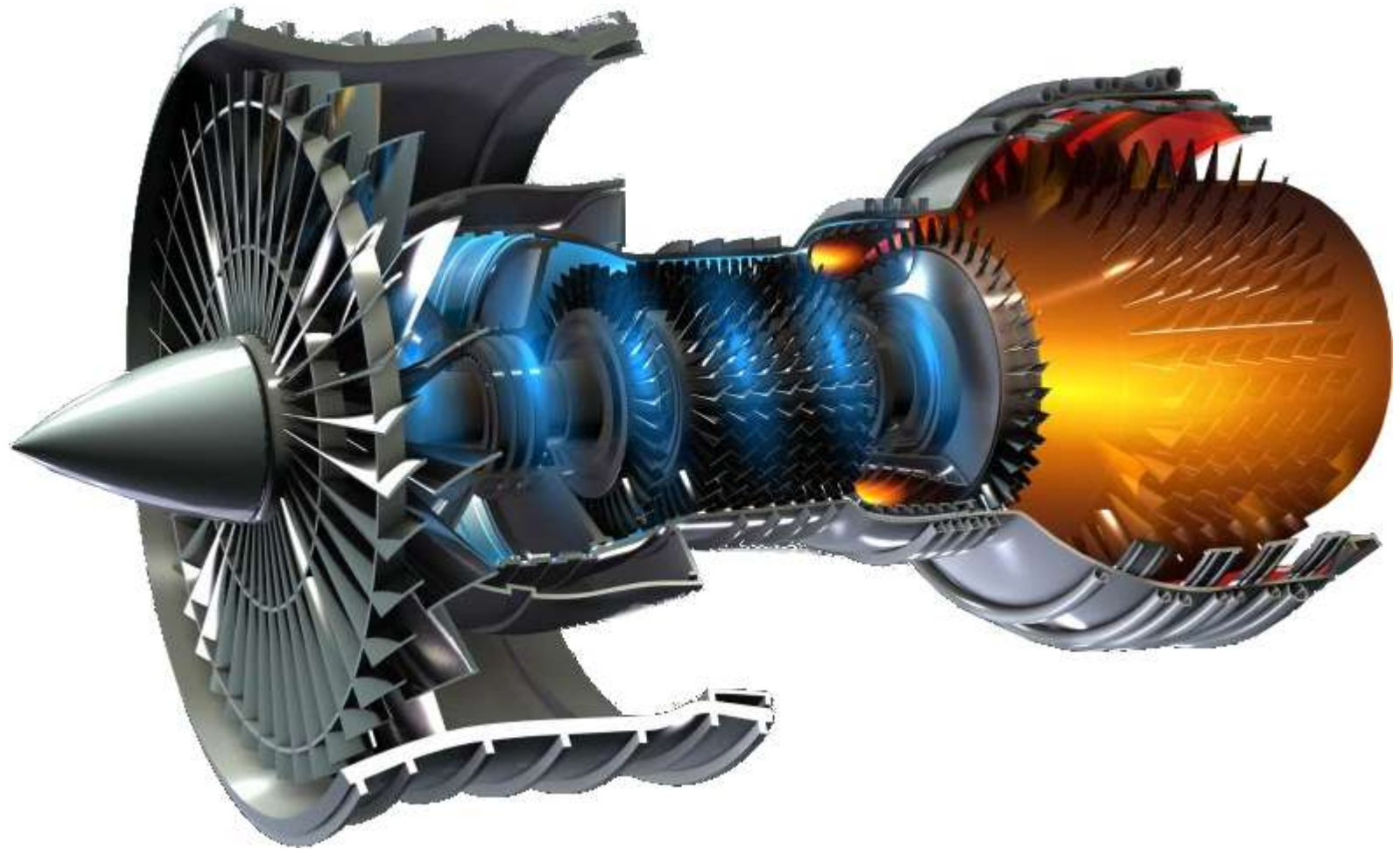
Cambridge
Royal Society University
Research Fellowship



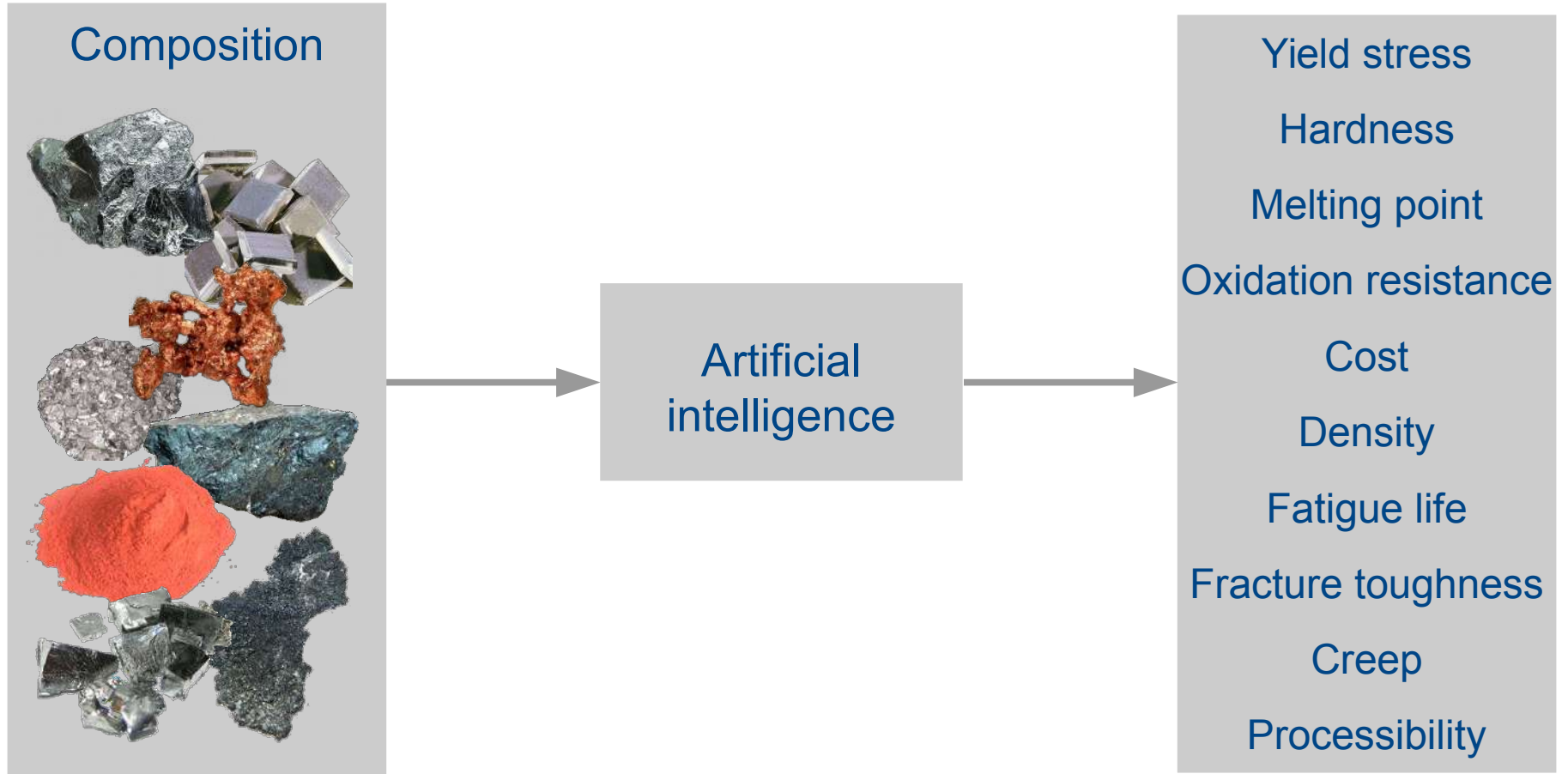
Approaches to materials design



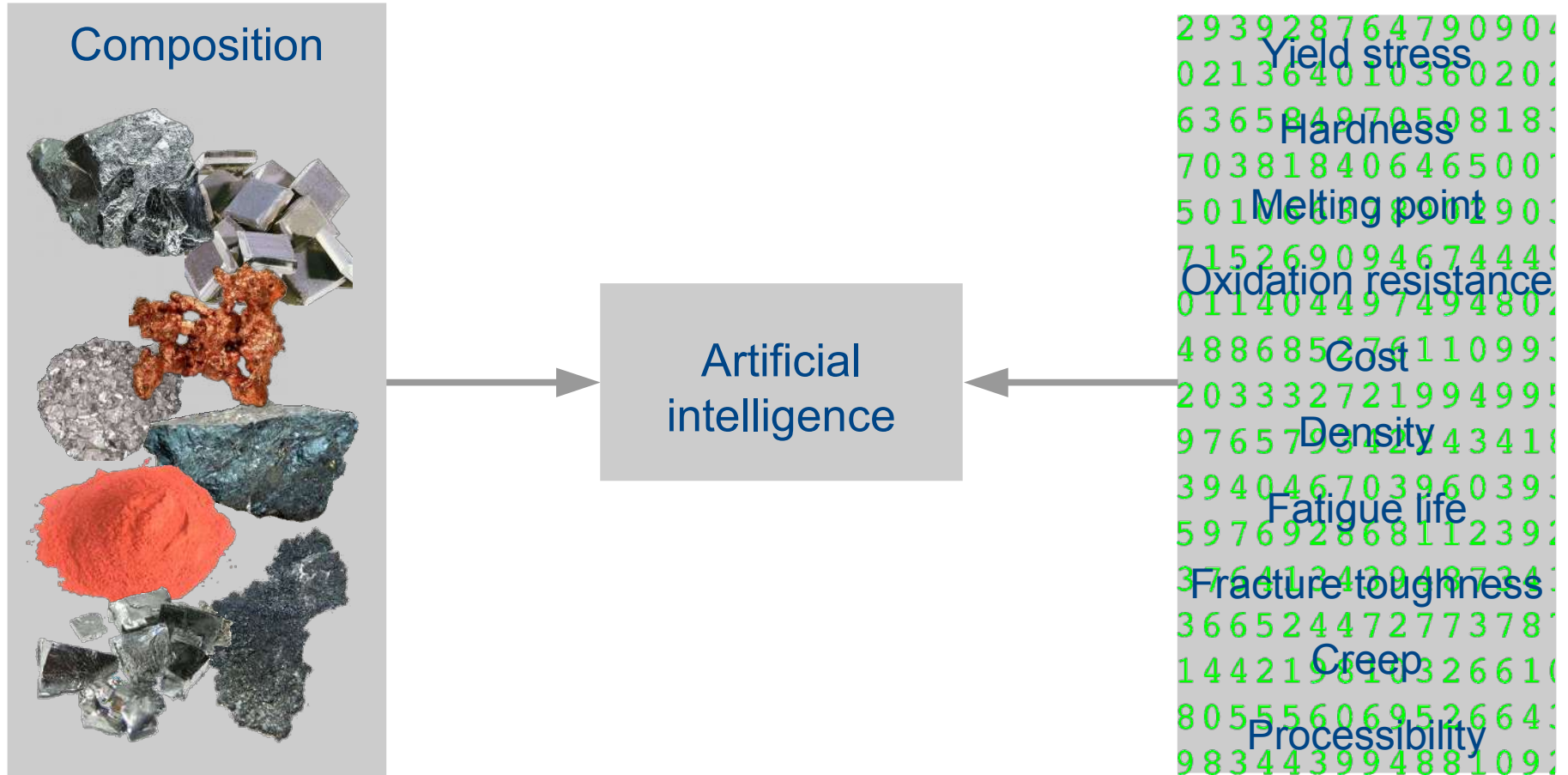
Schematic of a jet engine



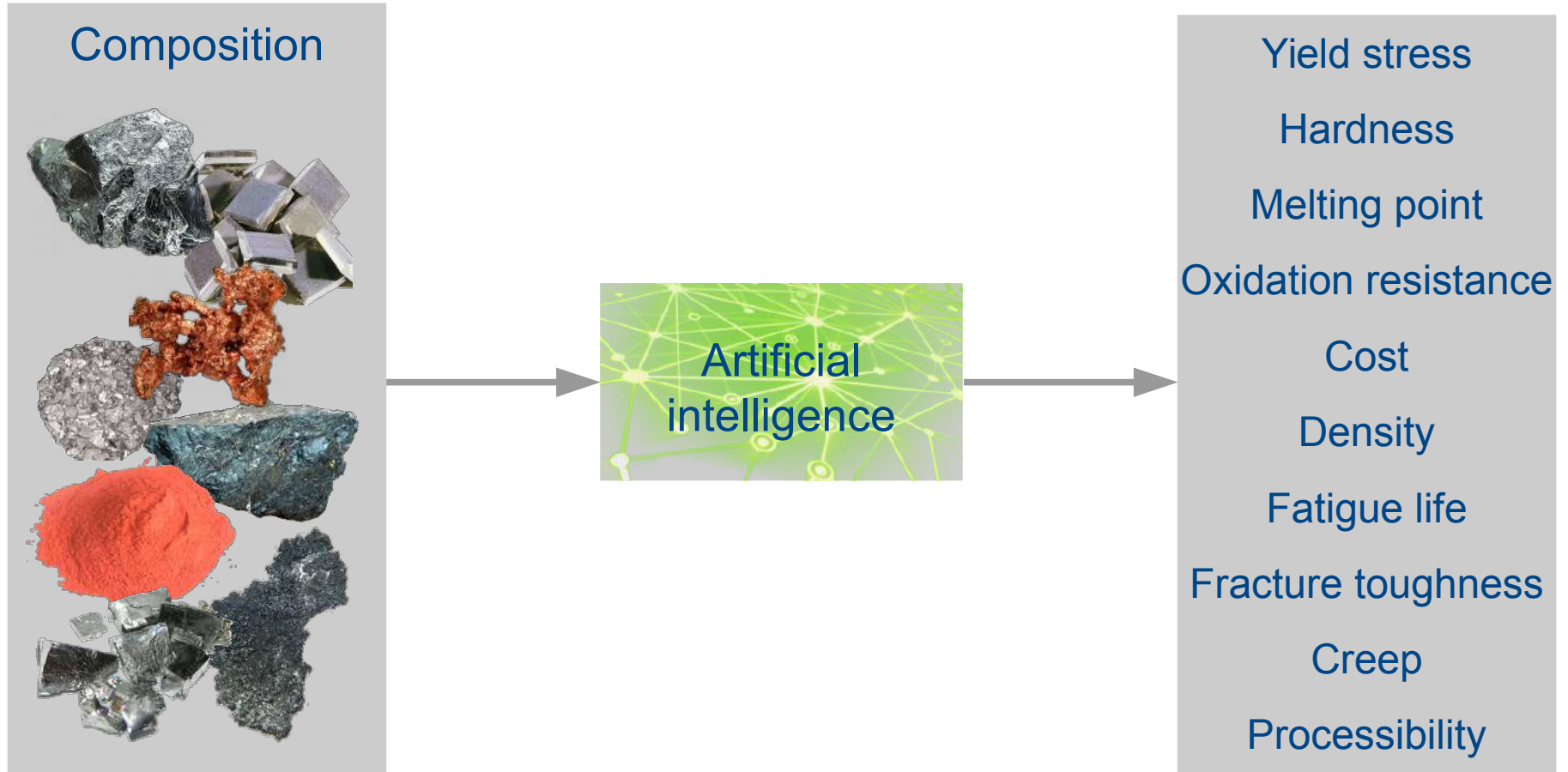
Artificial intelligence



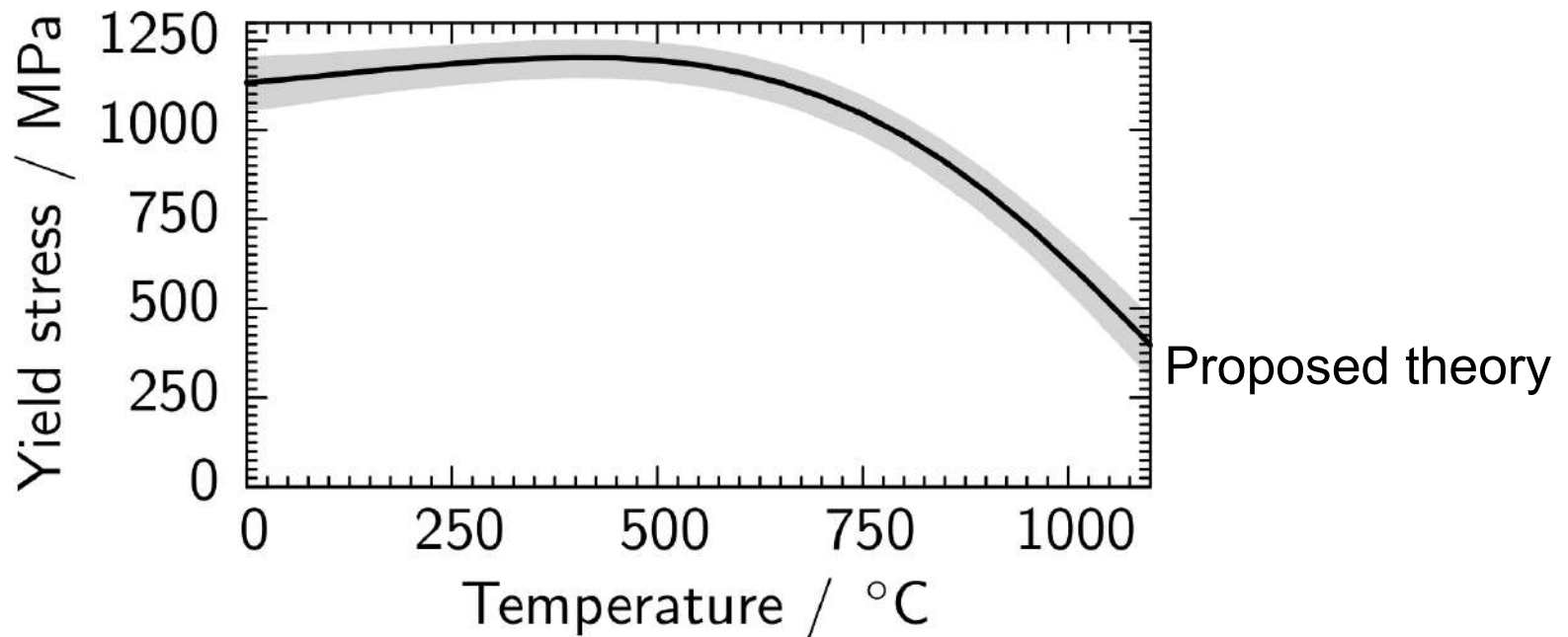
Artificial intelligence



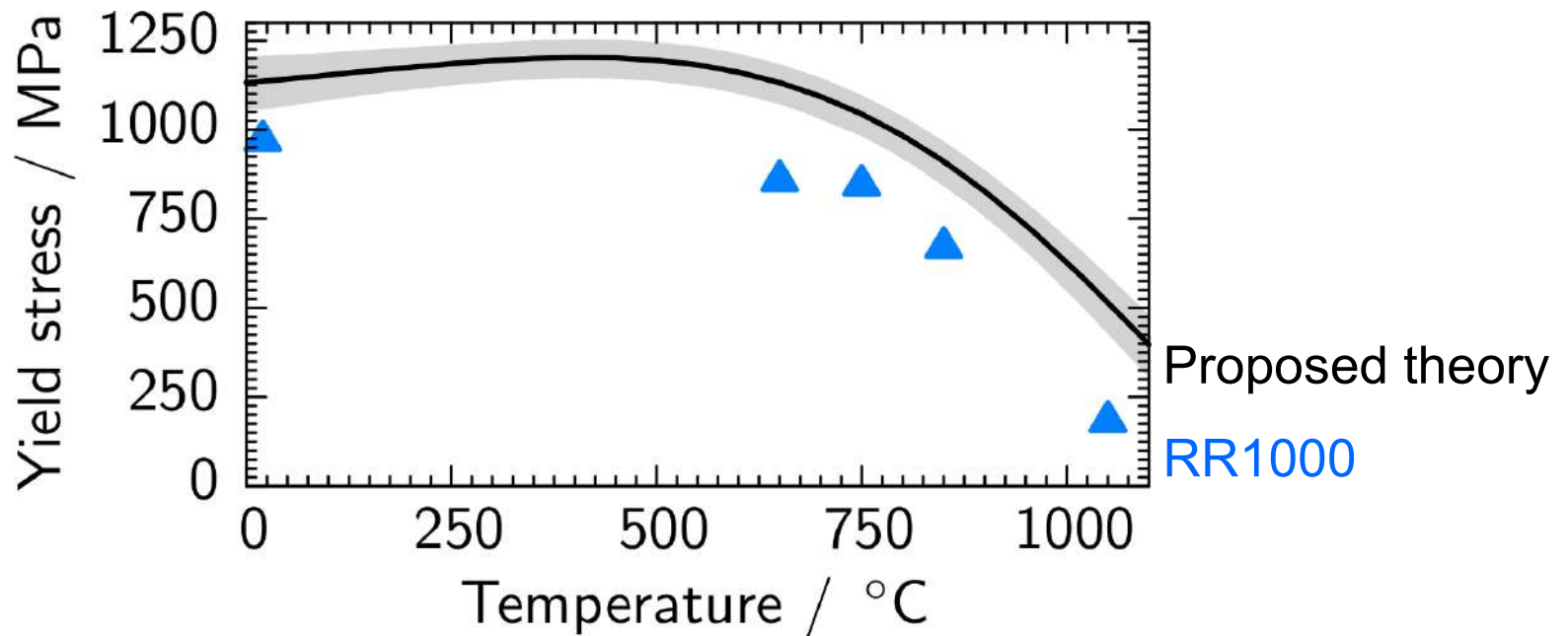
Artificial intelligence



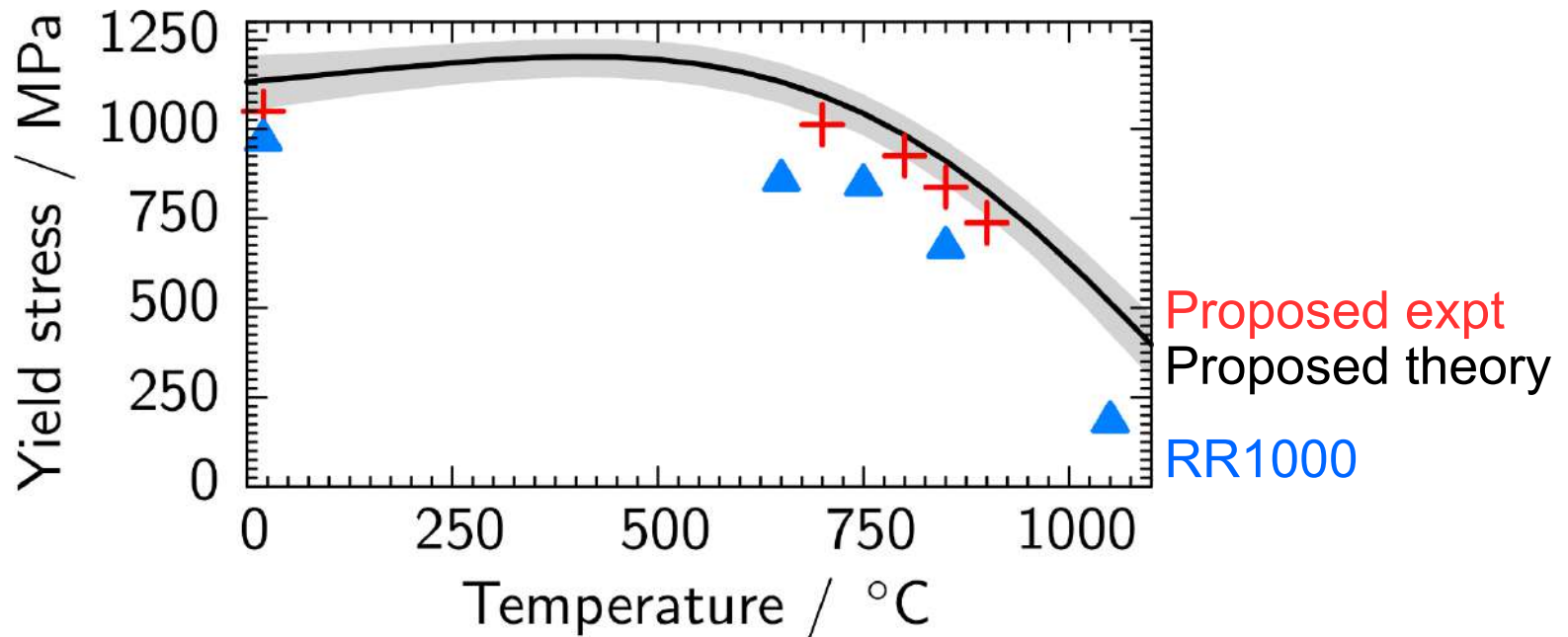
Testing the yield stress



Testing the yield stress



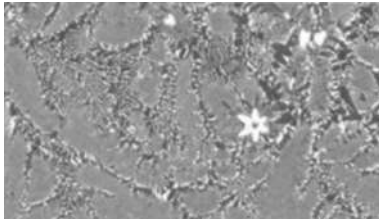
Testing the yield stress



Alloys discovered

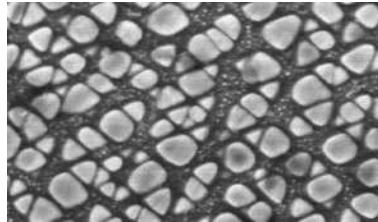
Cr-Cr₂Ta alloys

Intermetallics, 48, 62



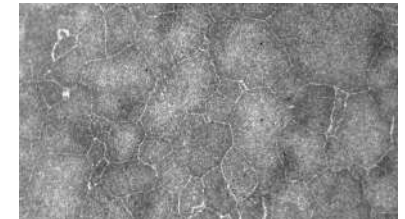
Combustor alloy

GB1408536



RR1000 grain growth

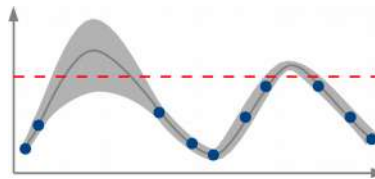
Acta Materialia, 61, 3378



Discovery algorithm

EP14153898

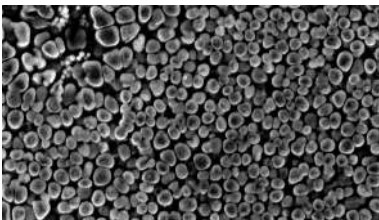
US 2014/177578



Ni disc alloy

EP14157622

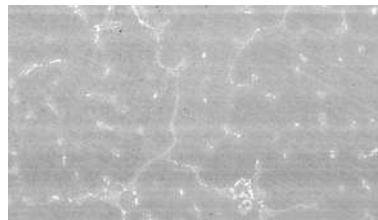
US 2013/0052077 A2



Mo-Hf forging alloy

EP14161255

US 2014/223465



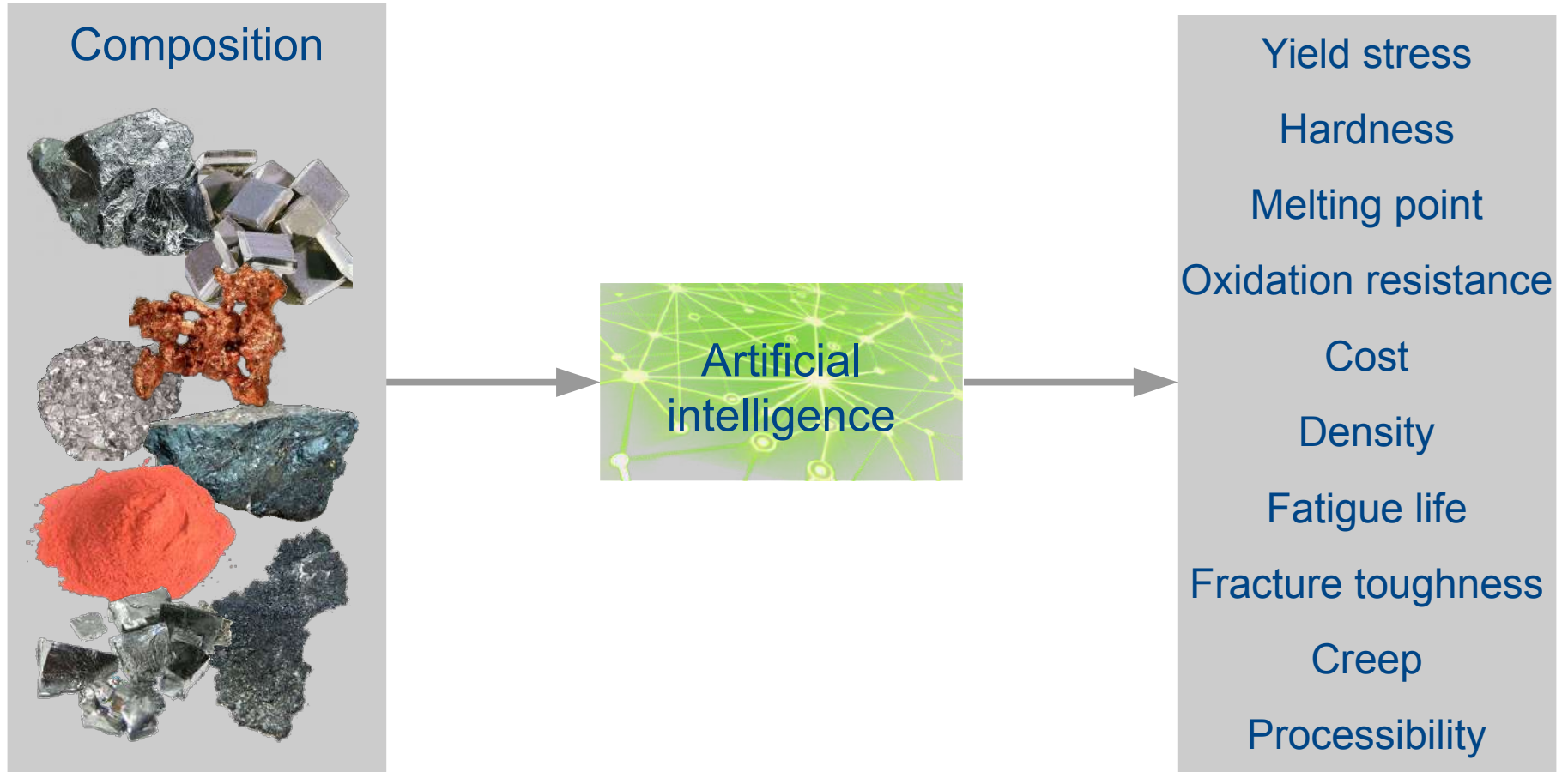
Mo-Nb forging alloy

EP14161529

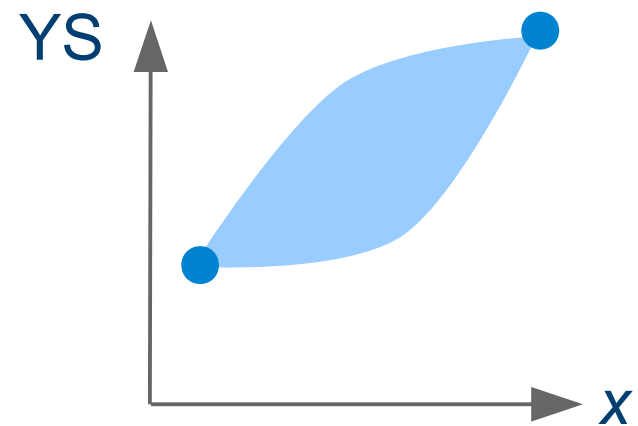
US 2014/224885



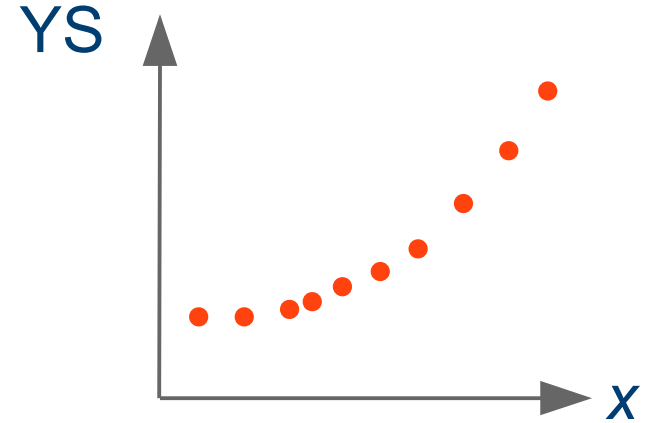
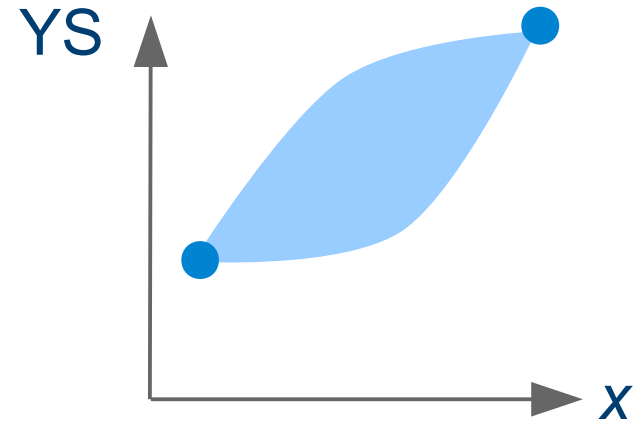
Artificial intelligence



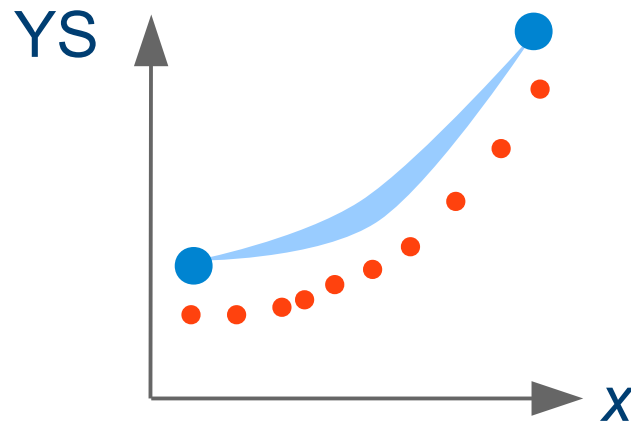
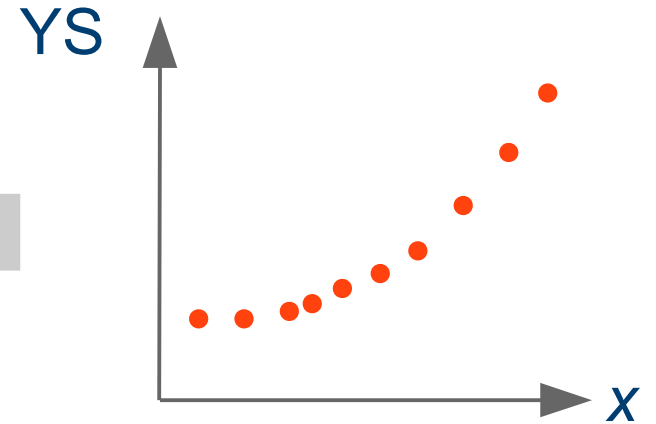
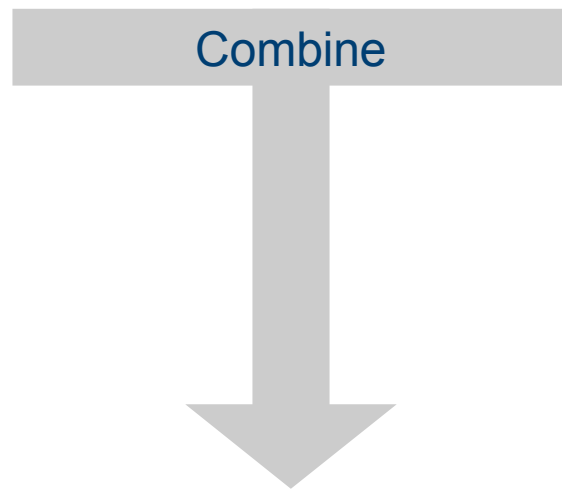
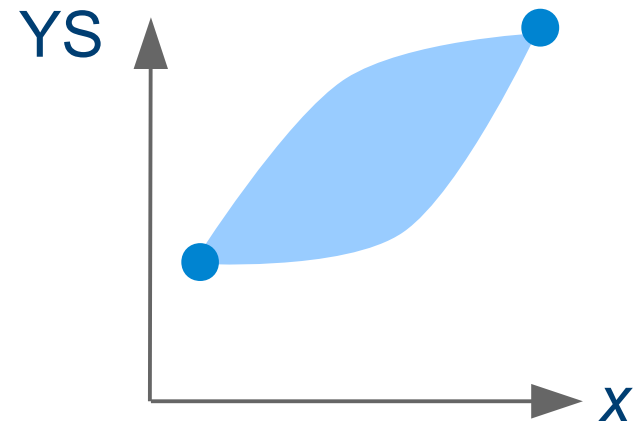
Merging simulation and experiment



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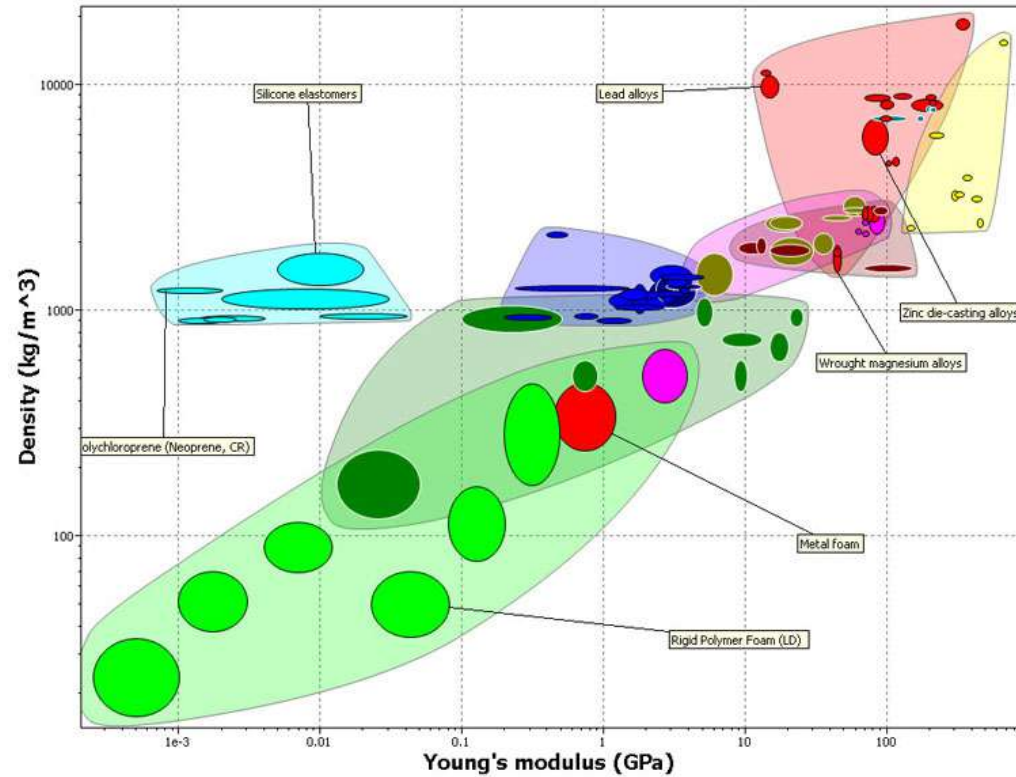
Exploiting experiment and simulations



Lithium cathode materials



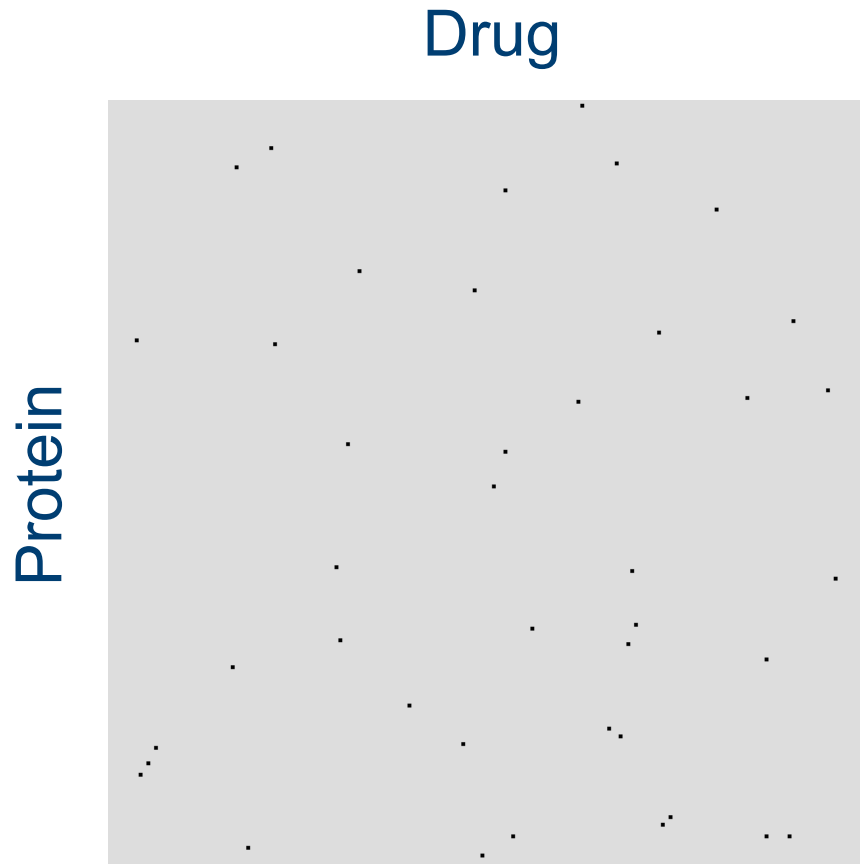
Database verification



Found 156 erroneous points confirmed against primary sources

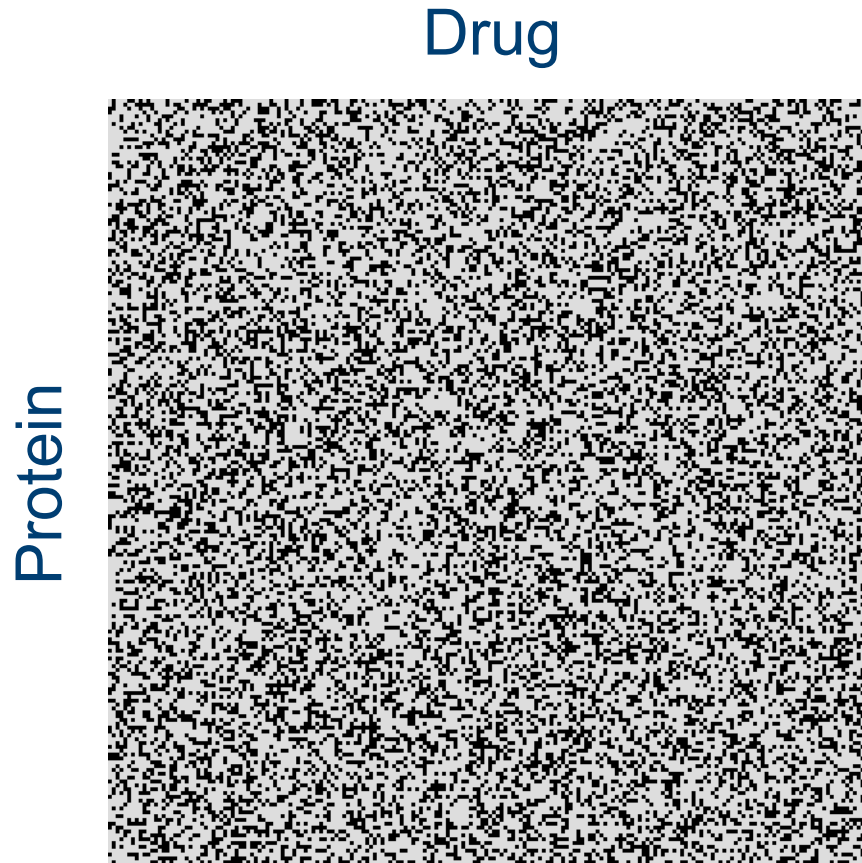
Protein activity data

ChEMBL database has protein activity for 0.1% of compounds



Protein activity data

Filled in 32% of the data points with 99.3% accuracy



Summary

Used artificial intelligence to discover materials

Can handle fragmented data

Merge simulations and experiments into holistic design tool

Proposed four new alloys, experimentally verified