

Data validation and imputation with artificial intelligence

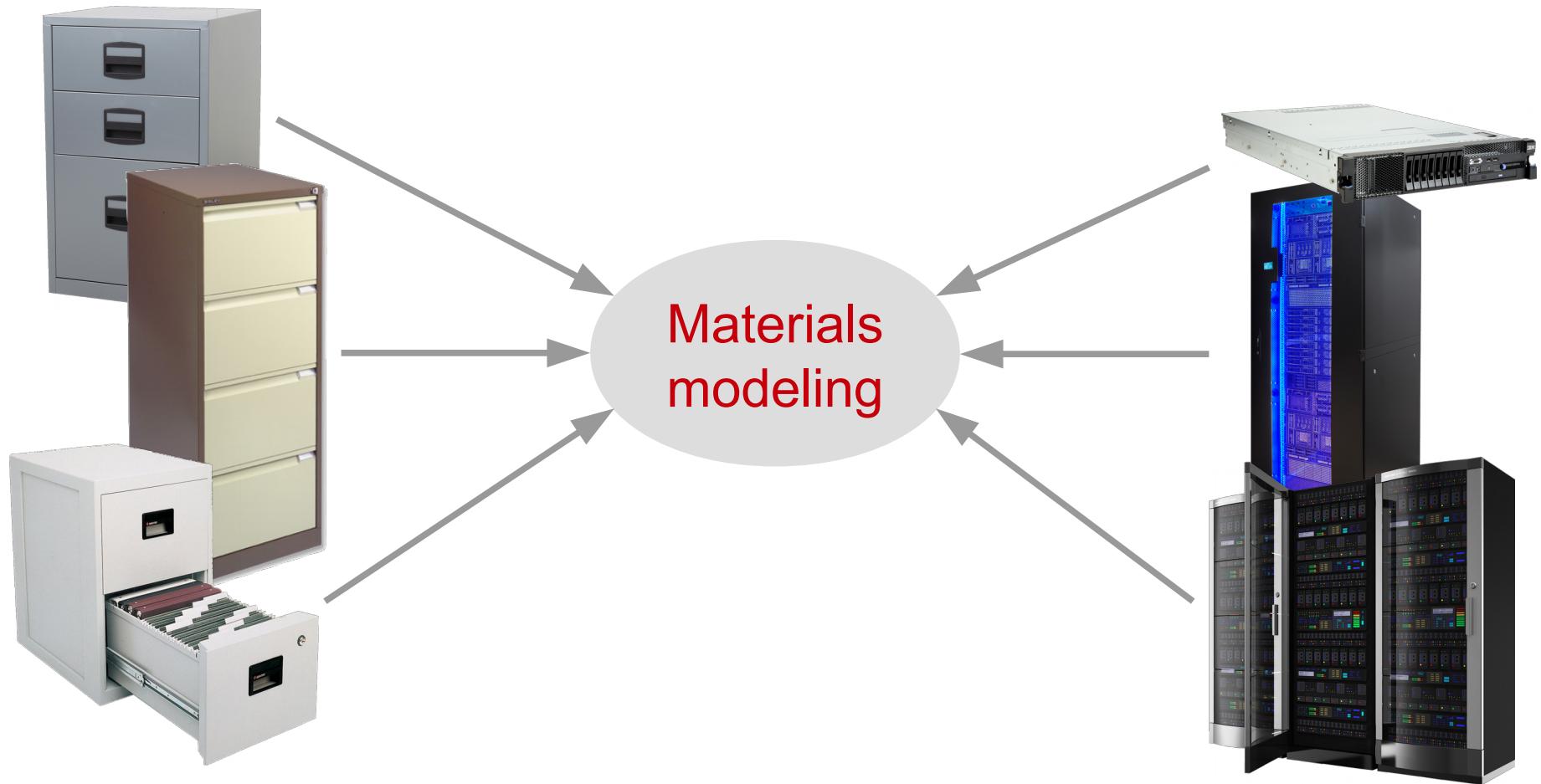
Gareth Conduit

TCM Group, Department of Physics

Exploiting materials databases

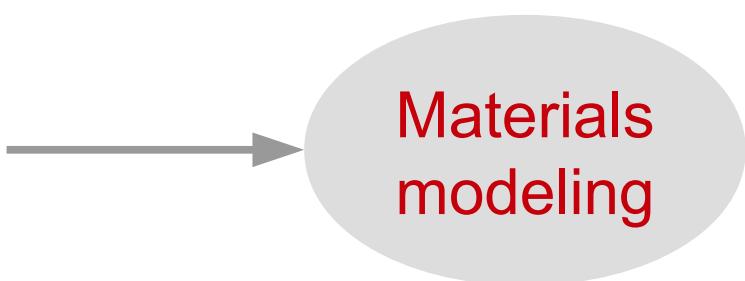
Experiment

Simulation



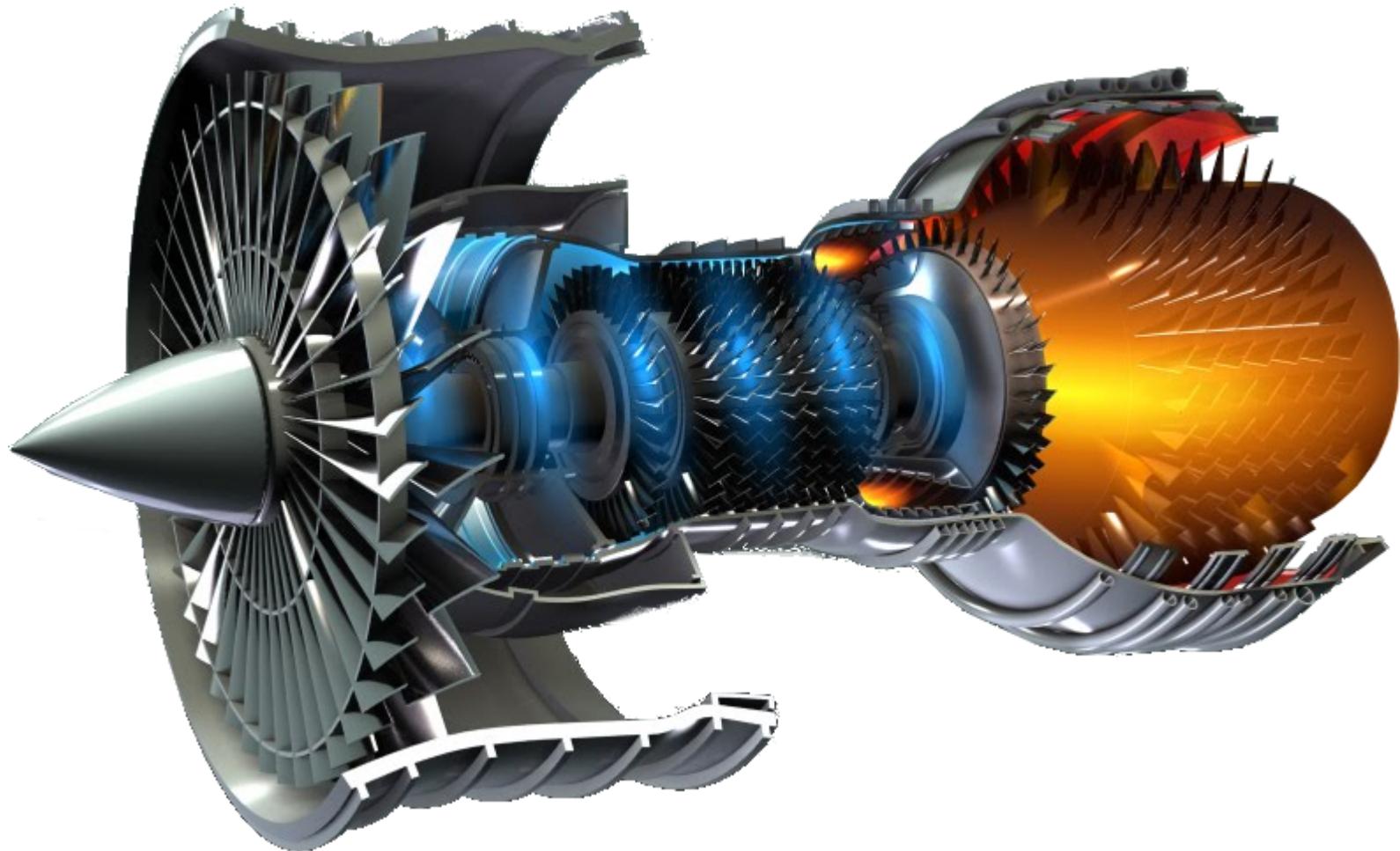
Exploiting materials databases

Experiment

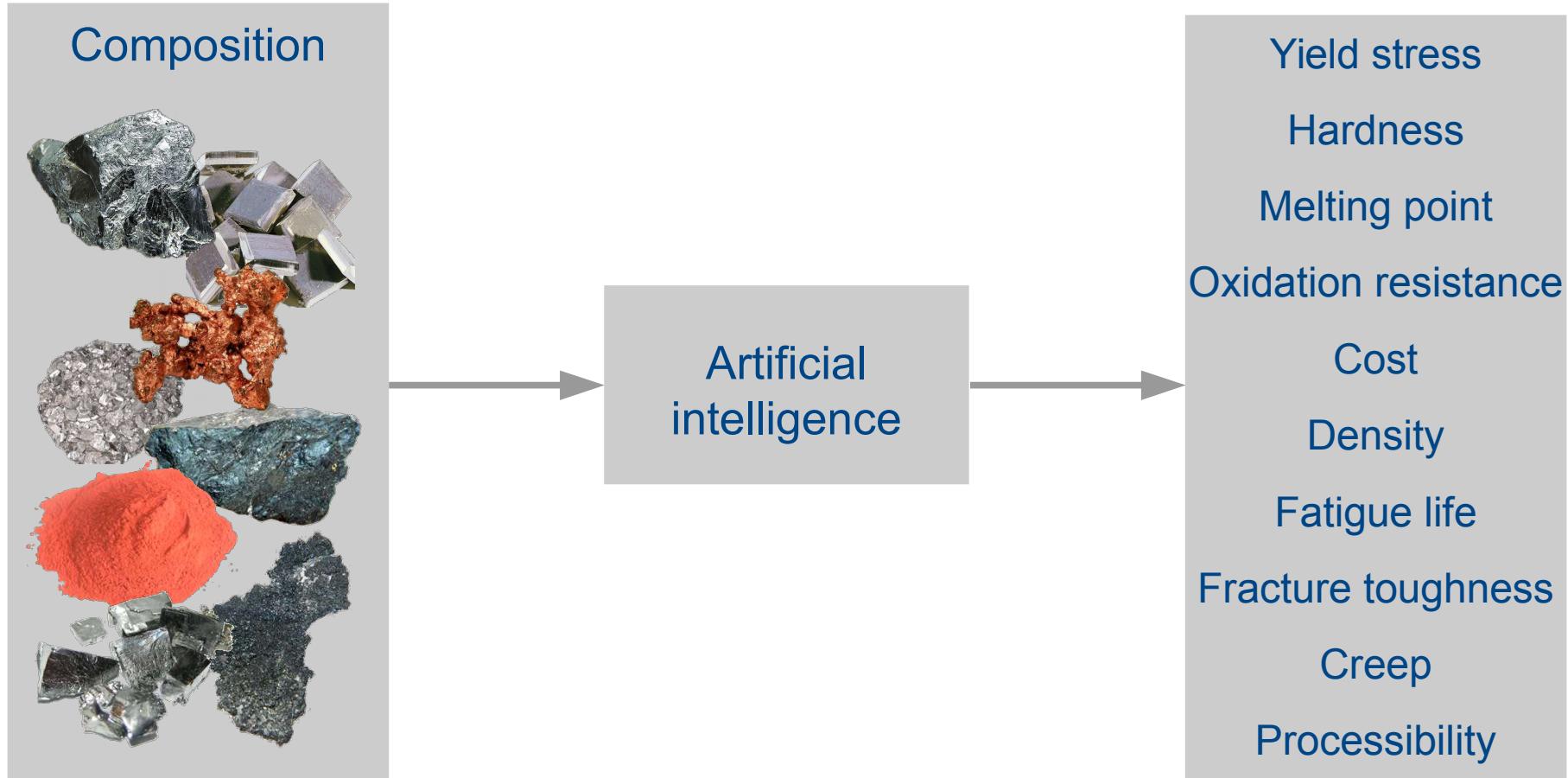


Materials
modeling

Schematic of a jet engine



Artificial intelligence



Artificial intelligence



Artificial
intelligence

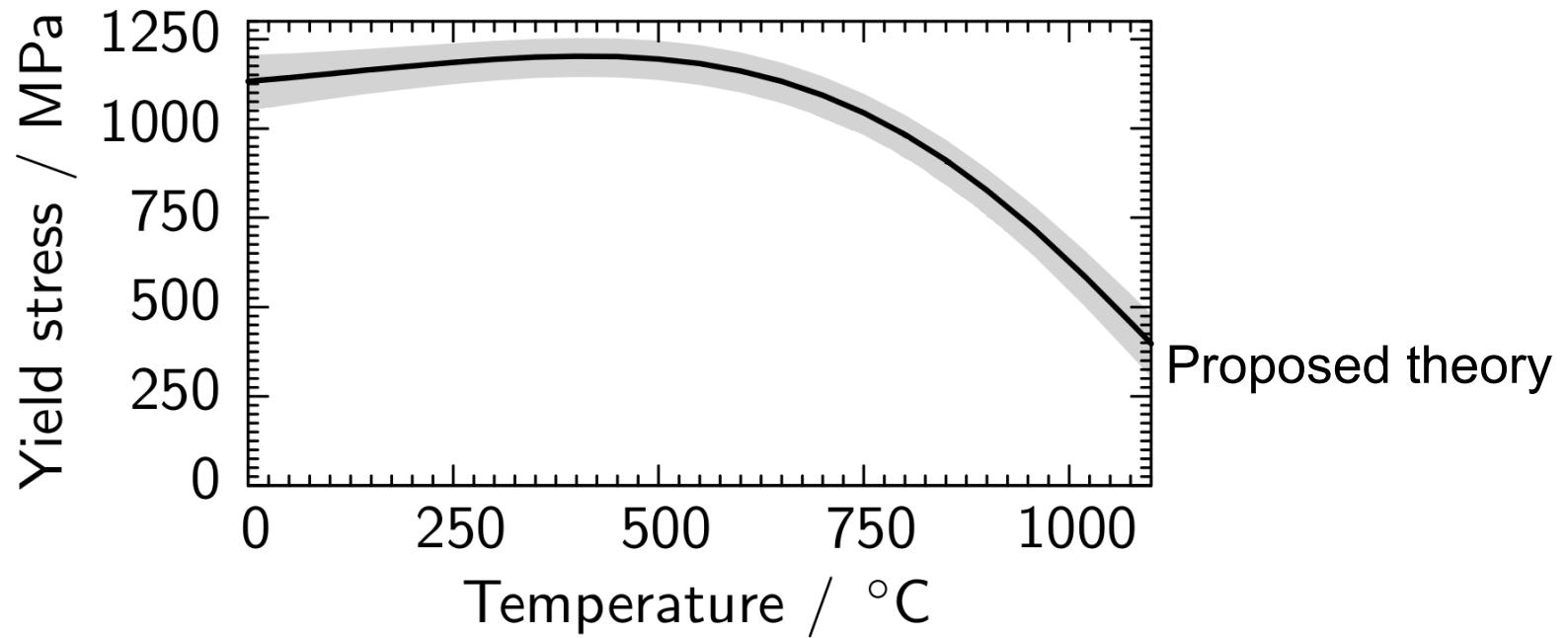
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7	0	3	8	1	8	4	0	6	4	6	5	0	0	7
5	0	1	6	3	6	4	9	7	0	5	0	2	9	0
7	1	5	2	6	9	0	9	4	6	7	4	4	9	4
0	1	1	4	0	4	4	9	7	4	9	4	8	0	2
4	8	8	6	8	5	7	3	1	1	0	9	9	3	1
2	0	3	3	3	2	7	2	1	9	9	4	9	9	5
9	7	6	5	7	9	9	4	2	1	4	3	4	1	8
3	9	4	0	4	6	7	0	3	9	6	0	3	9	1
5	9	7	6	9	2	8	6	8	1	1	2	3	9	2
3	7	6	4	1	3	4	2	0	4	8	7	3	4	1
3	6	6	5	2	4	4	7	2	7	7	3	7	8	1
1	4	4	2	1	9	8	1	3	2	6	6	1	0	6
8	0	5	5	5	6	0	6	9	5	2	6	6	4	3
9	8	3	4	4	3	9	9	4	8	8	1	0	9	2

Artificial intelligence

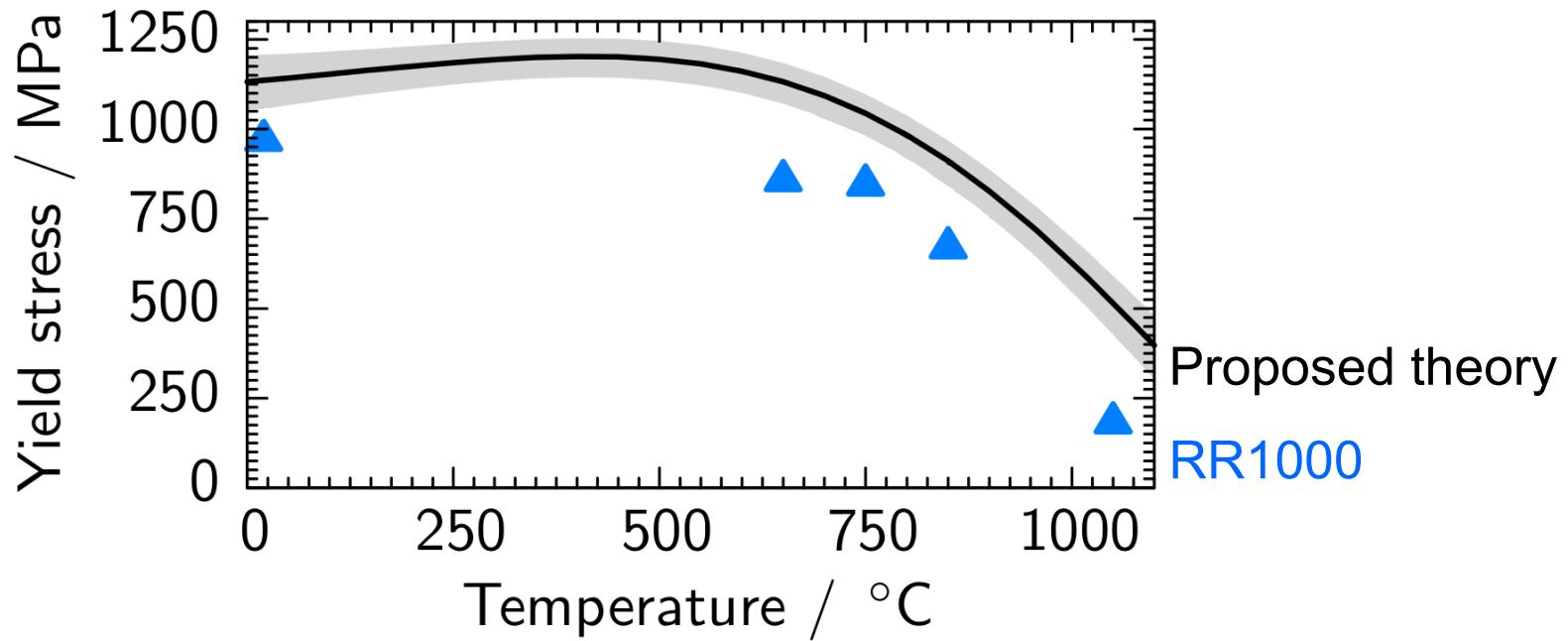


- Yield stress
- Hardness
- Melting point
- Oxidation resistance
- Cost
- Density
- Fatigue life
- Fracture toughness
- Creep
- Processability

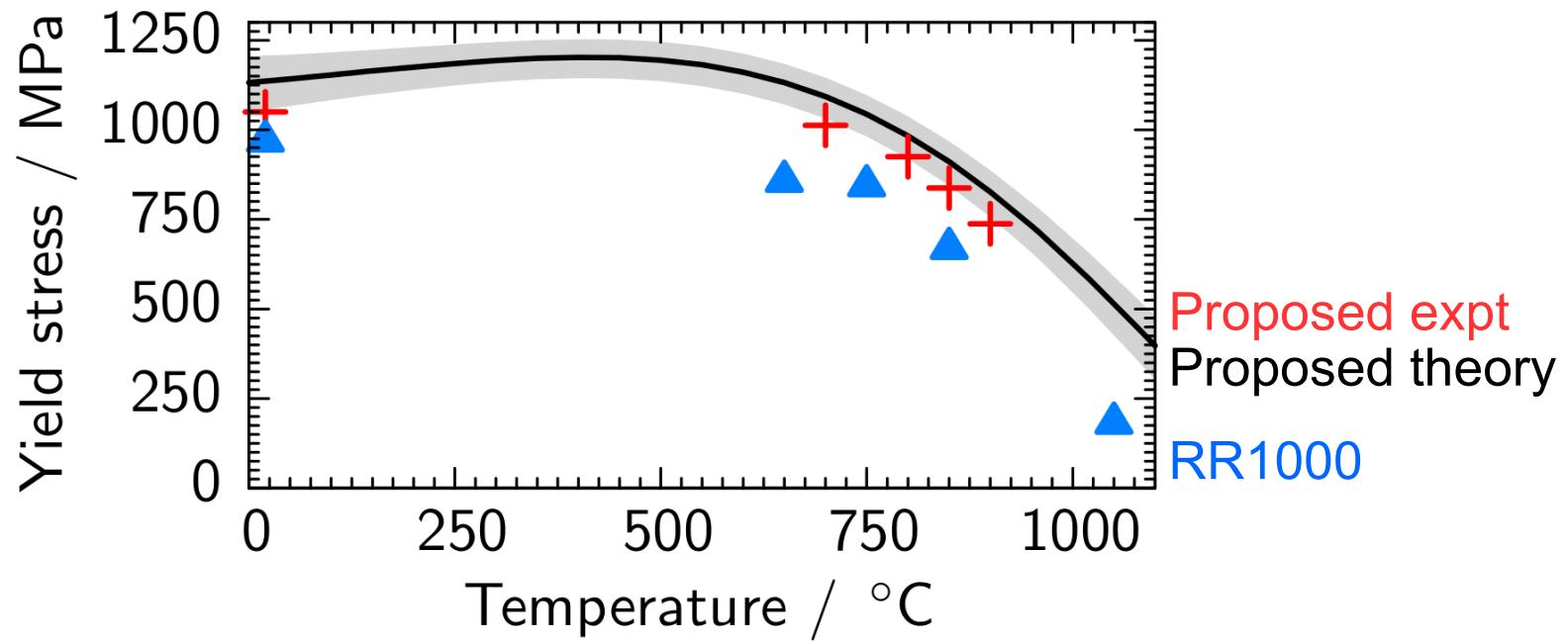
Testing the yield stress



Testing the yield stress



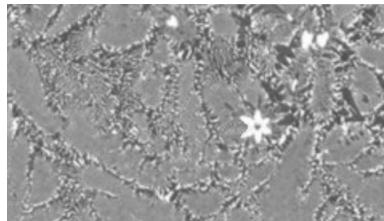
Testing the yield stress



Alloys discovered

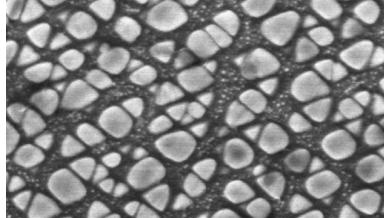
Cr-Cr₂Ta alloys

Intermetallics, 48, 62



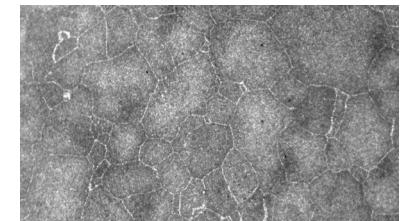
Combustor alloy

GB1408536



RR1000 grain growth

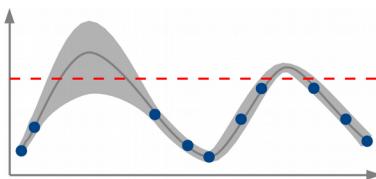
Acta Materialia, 61, 3378



Discovery algorithm

EP14153898

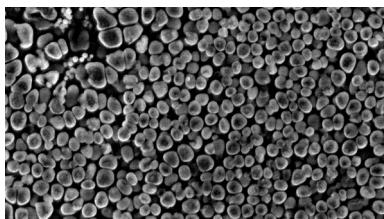
US 2014/177578



Ni disc alloy

EP14157622

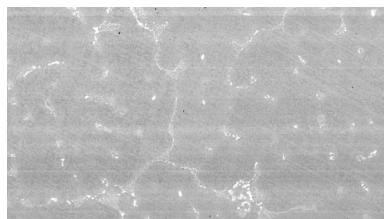
US 2013/0052077 A2



Mo-Hf forging alloy

EP14161255

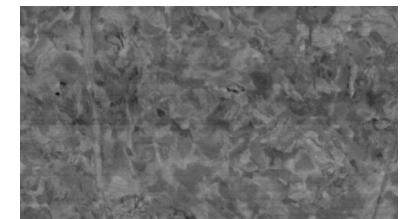
US 2014/223465



Mo-Nb forging alloy

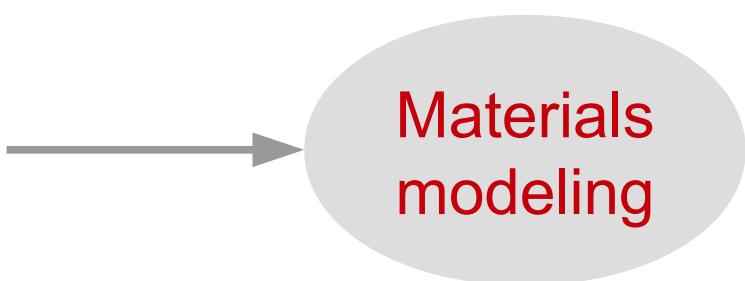
EP14161529

US 2014/224885



Exploiting materials databases

Experiment

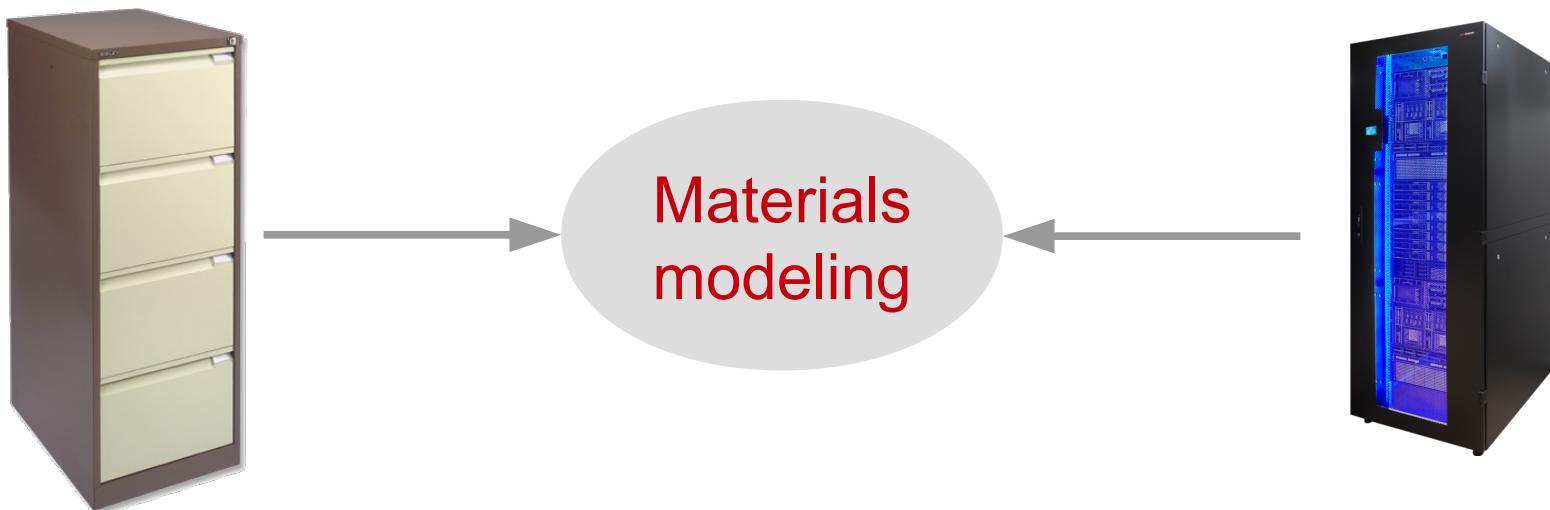


Materials
modeling

Exploiting materials databases

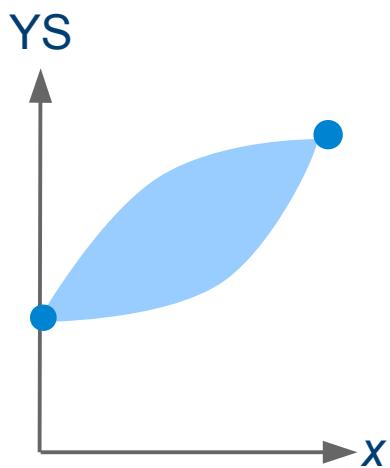
Experiment

Simulation



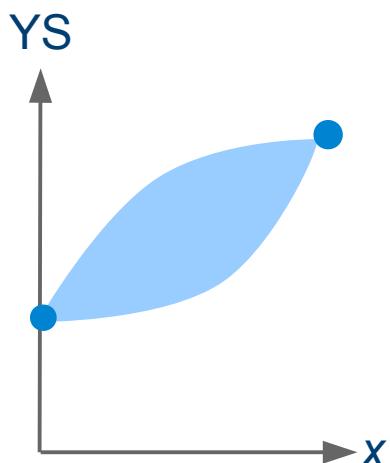
Merging simulation and experiment

Experiment

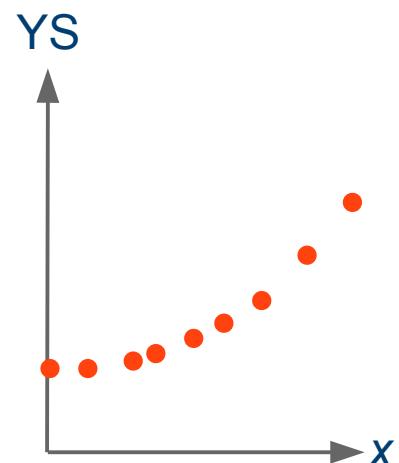


Merging simulation and experiment

Experiment

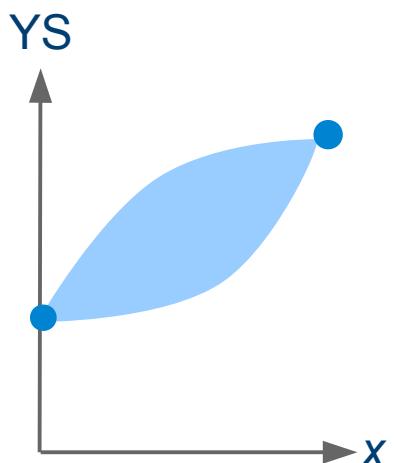


Simulation

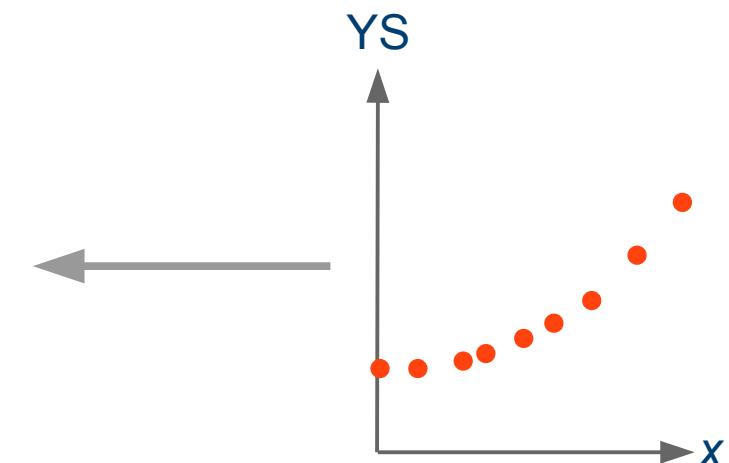


Merging simulation and experiment

Experiment



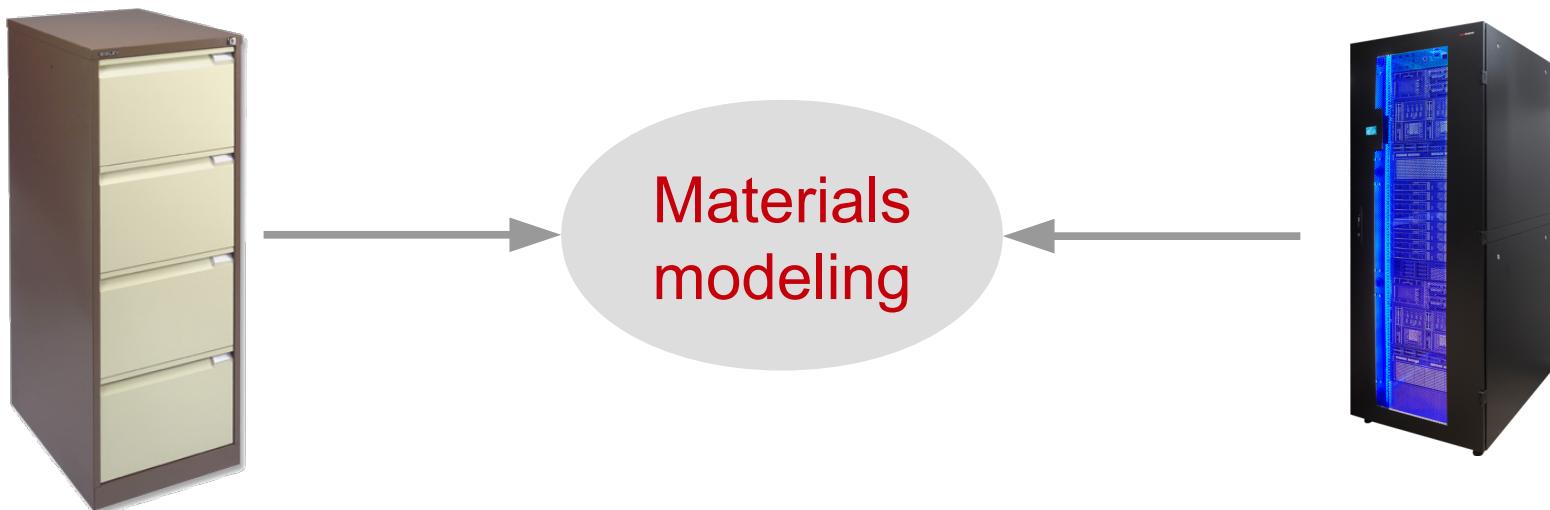
Simulation



Exploiting materials databases

Experiment

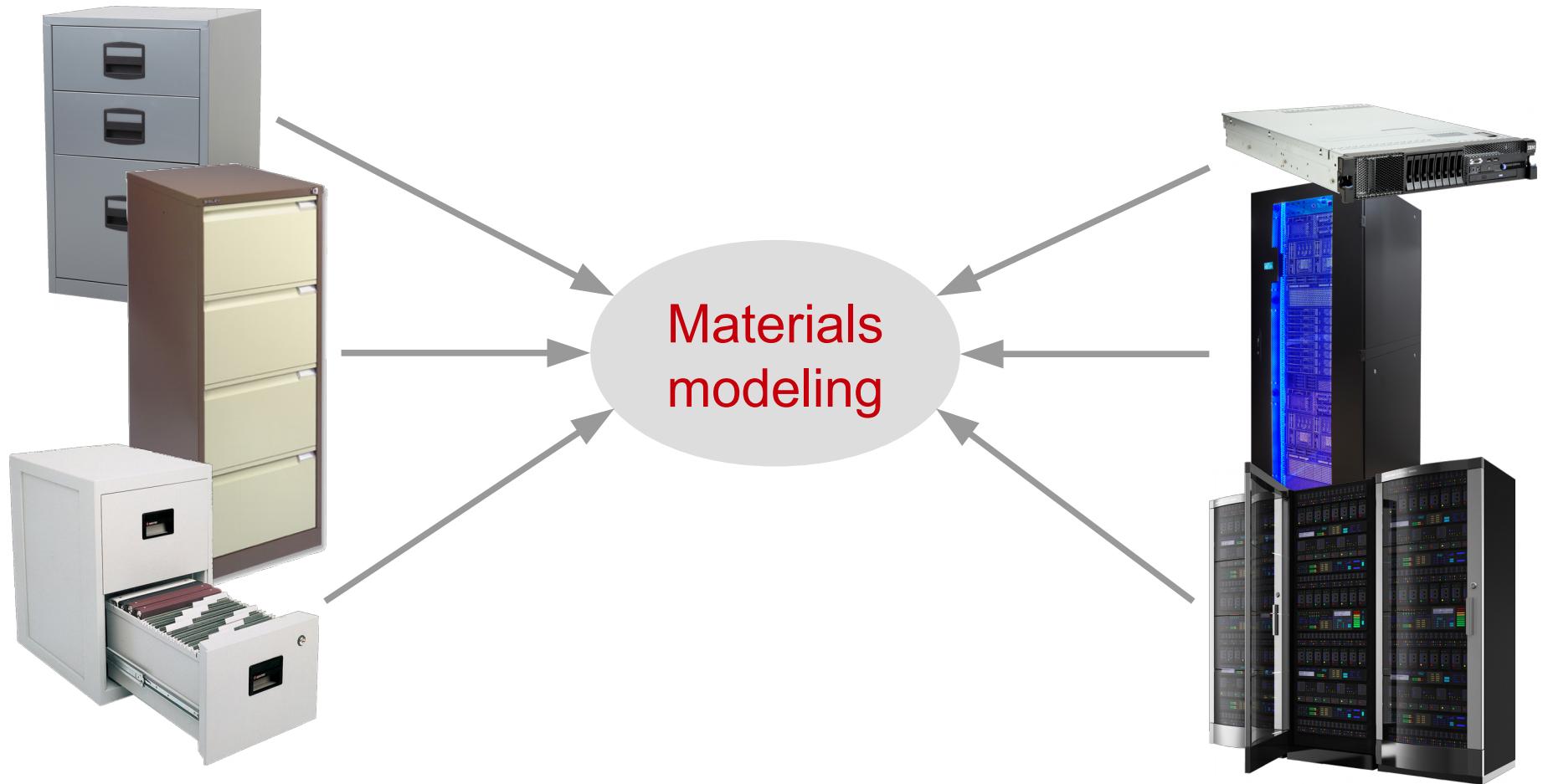
Simulation



Exploiting materials databases

Experiment

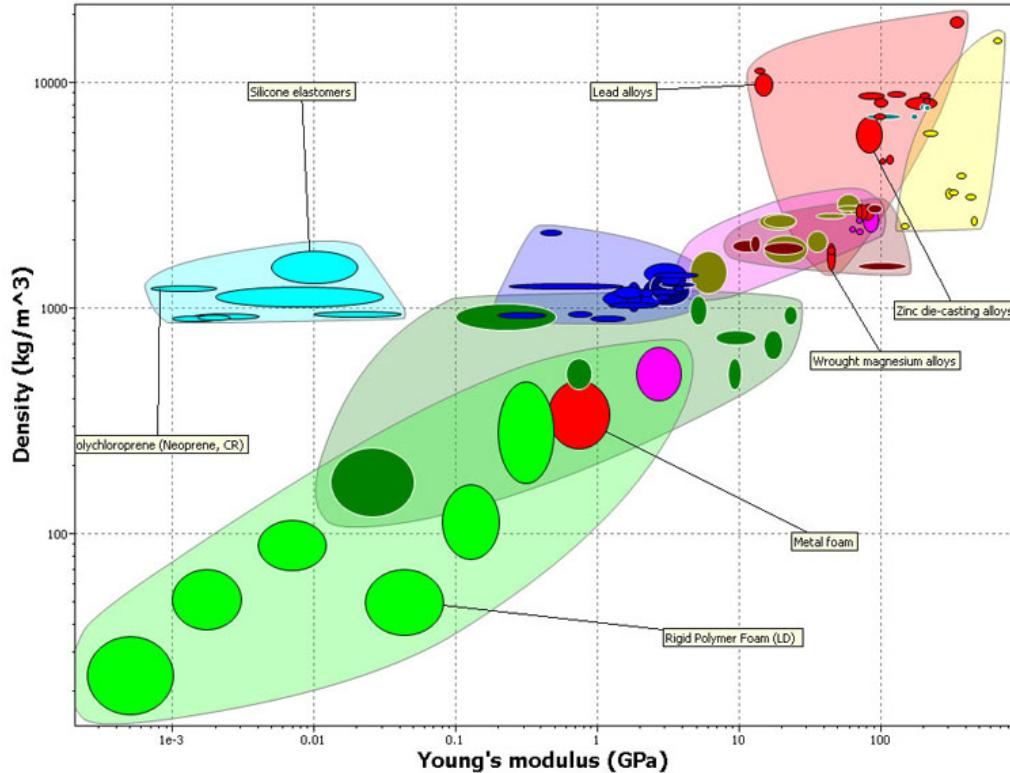
Simulation



Fragmented databases

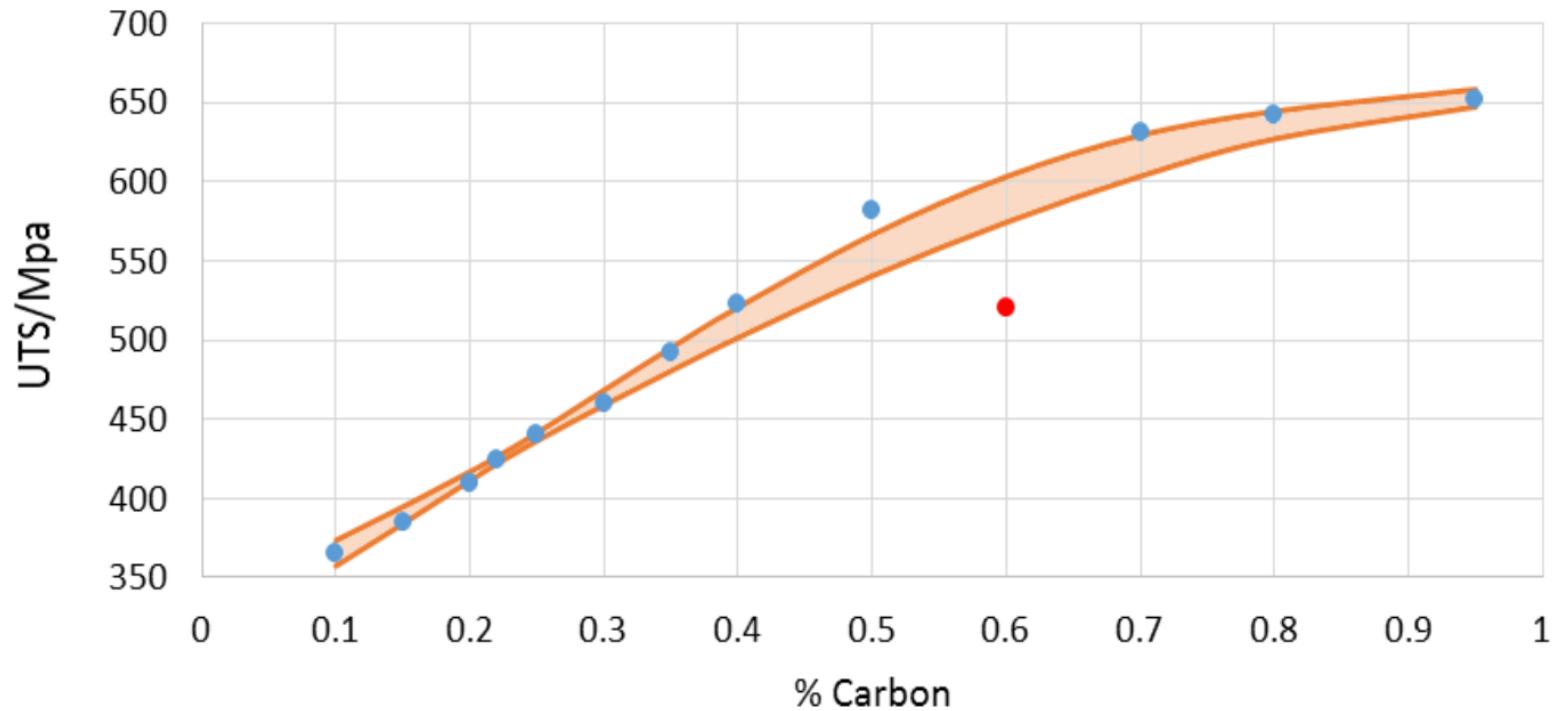
Composition	Computed YS	YS	Hardness	UTS	Charpy	Compressive
✓	✓	✓	✓	✗	✗	✗
✓	✓	✗	✓	✓	✗	✓
✓	✓	✗	✗	✓	✓	✓
✓	✓	✓	✓	✗	✓	✗
✓	✓	✗	✗	✓	✗	✗
✓	✓	✗	✓	✗	✓	✗
✗	✗	✓	✗	✓	✓	✗
✓	✓	✗	✓	✗	✗	✓
✓	✓	✓	✗	✗	✓	✗

Database verification



10,000,000 entries, 20% filled

Erroneous points: 1D example

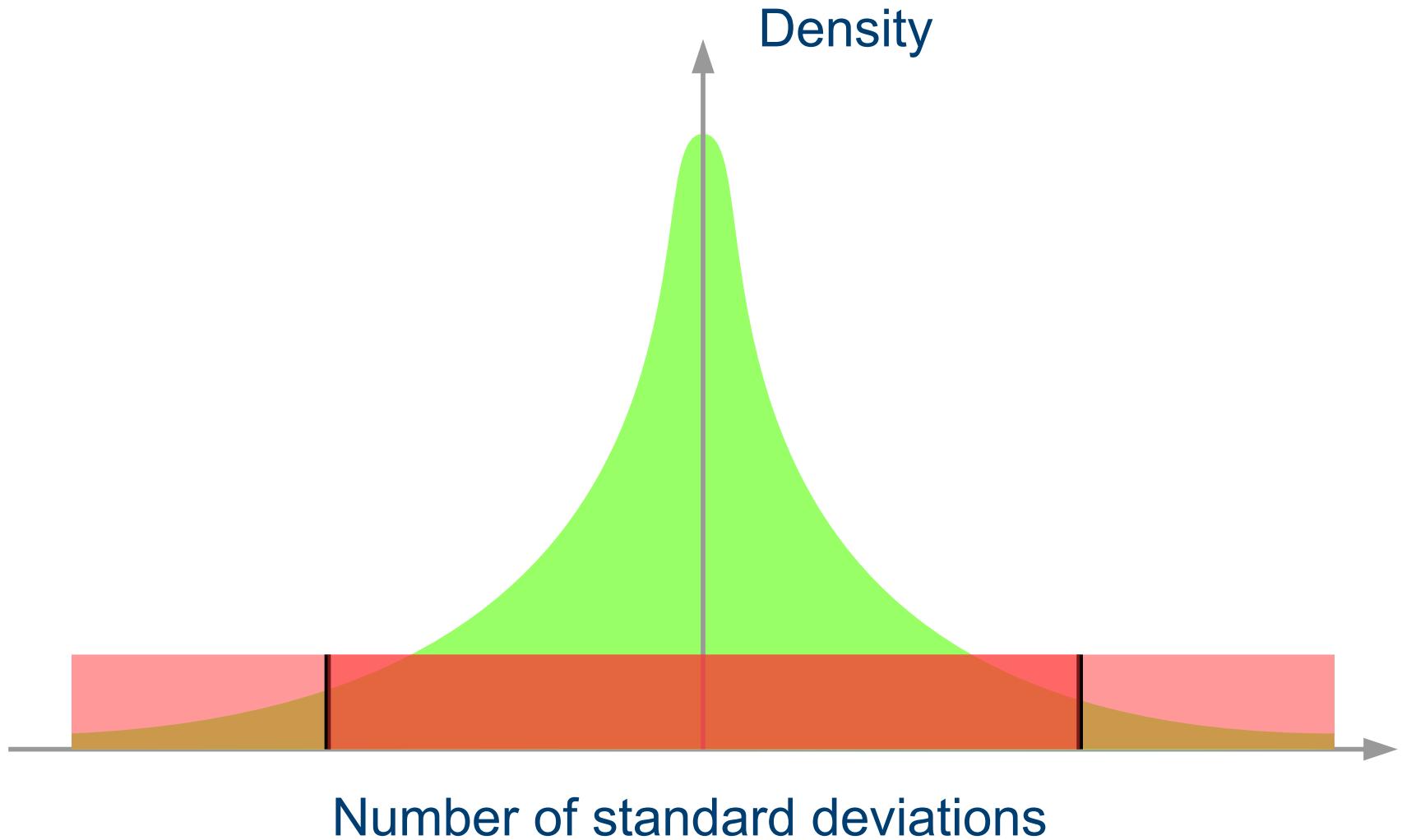


Erroneous points: density & yield stress

Alloy	Source density/gcc ⁻¹	NN density/gcc ⁻¹	#Sigma	Correct density /gcc ⁻¹
Stainless steel, duplex, Iium P, cast	7.60	7.9060	11.6	7.75-8.0
Tool steel, molybdenum alloy, AISI M43 (high speed)	8.44	8.0345	-11.5	7.7-8.03
Copper-nickel alloy, C70400, wrought, half hard (95/5 copper-nickel)	8.53	8.9254	10.7	8.94
Tool steel, AISI A3	8.00	7.7211	-20.2	7.86
Tool steel, AISI A4	7.91	7.80	8.8	8.03
	Granta Melting Point/K	NN Melting Point /K	#Sigma	Actual Value/K
				11.4
				7.70-8.03
Wrought iron	1973.0	Material		UTS /Mpa
Nickel-Fe-Cr alloy, INCOLOY 840, annealed	1419.0			Yield stress /Mpa
Titanium, alpha-beta alloy, Ti-4.5Al-3V-2Fe-2Mo, annealed	1593.2			Neural Network prediction
Carbon steel, AISI 1095	1650.2			# Sigma
		Low alloy steel, AISI 4150, annealed	726.64	520
		Low alloy steel, AISI 4150, normalized	1149.26	918
		Low alloy steel, AISI 4150, tempered at 650°C & oil quenched	952.52	729
				-15.1901

Found 156 erroneous points confirmed against primary sources

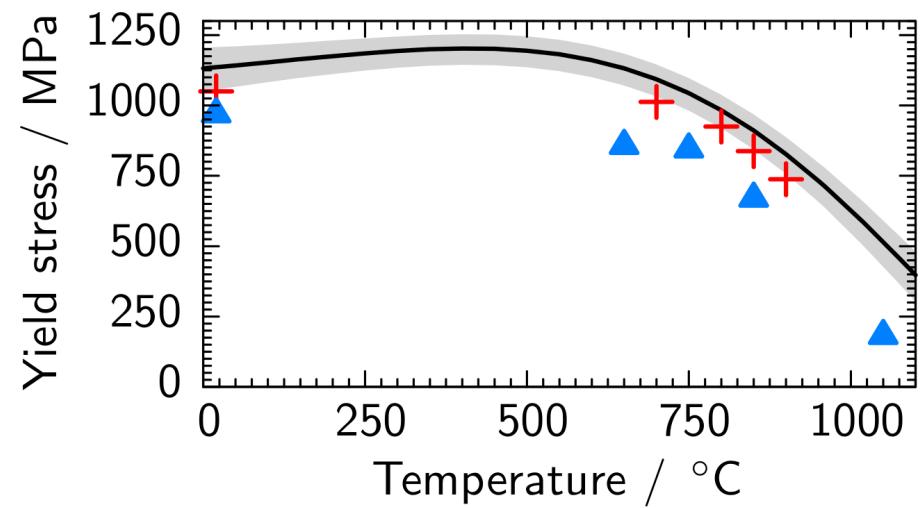
How many erroneous points remain?



Polymers

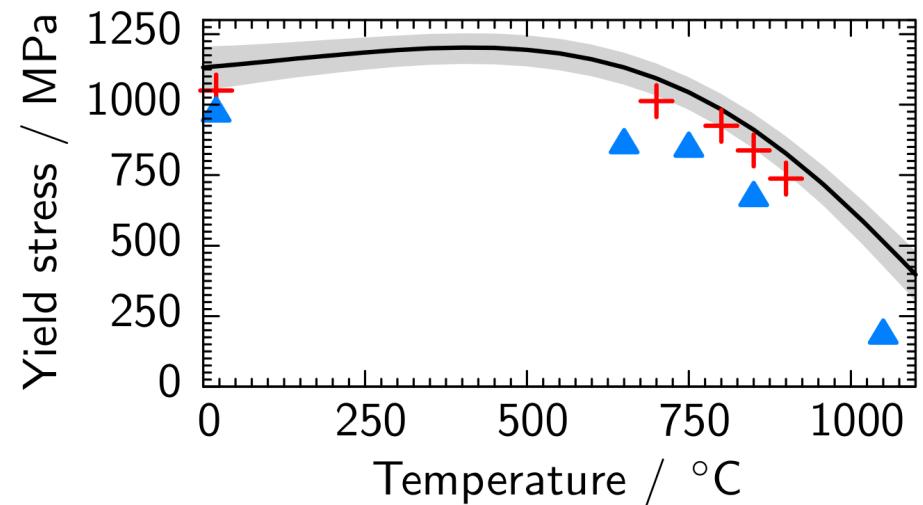
Name	Property	Value	Comment
4PROP®25C21120	Flexural Modulus (MPa)	2300	out by factor 10 ³
AZDEL™U400-B01N (Longitudinal)	Flexural Modulus (MPa)	8000	out by factor 10 ³
Hyundai EP PPF HT340	Flexural Strength (MPa)	46.9	out by factor 10
Borealis PP NJ201AI	Filler	Mineral: 20%	prediction: 19.9 ± 5.6%
Daplen™EE168AIB	Filler	Mineral: 10%	prediction 11.3 ± 3.0%
Hostacom M2 R03/2 105558	Filler	Mineral: 20%	prediction 14.8 ± 4.2%
Maxxam™NM-818.H001-1049	Filler	Glass Fibre: 20%	prediction 17.8 ± 4.4%
Beetle®PPC120M 9250	Filler	Mineral: 20%	prediction 9.9 ± 6.2%
EMOPLEN®CP GFR 20	Filler	Mineral: 20%	prediction 10.3 ± 2.7%
FORMULA P COMP 5220	Filler	Mineral: 20%	prediction 15.5 ± 3.2%
4PROP®9C13100	Filler	Mineral: 10%	prediction 13.5 ± 3.0%

Graphical data



Composition	Temperature	YS
A	200	1100
A	400	1150
A	600	1100
A	800	1000
A	1000	650
B	100	1000
C	100	1050

Graphical data



Composition	Temperature	YS
A	200,...,1000	1100,...,650
B	100	1000
C	100	1050

Exploiting experiment and simulations

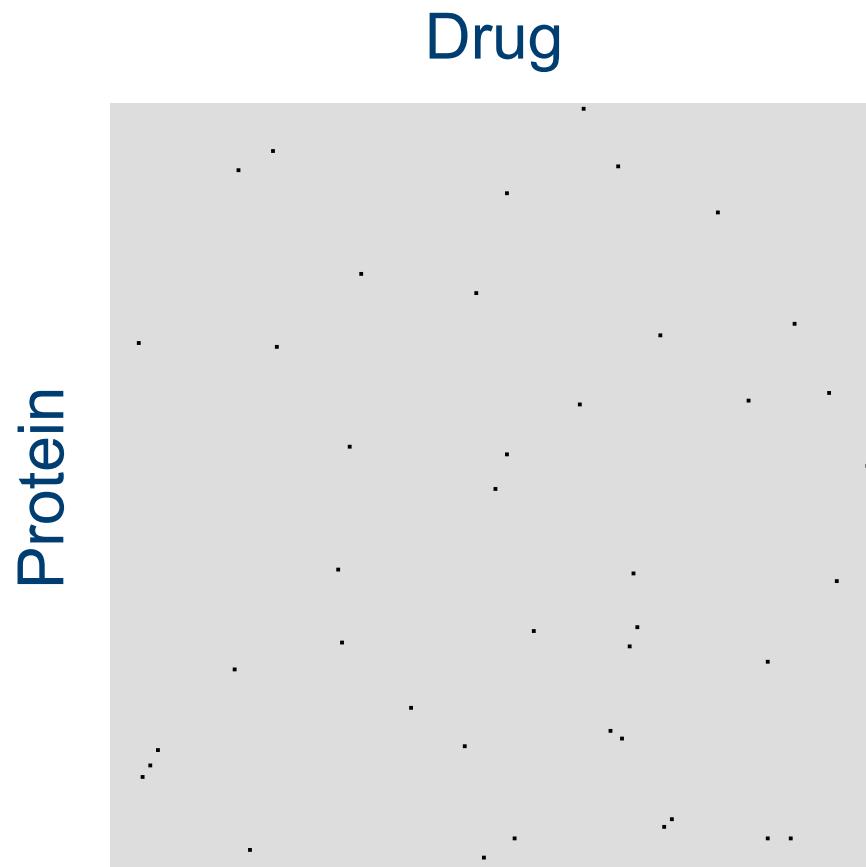


Lithium cathode materials



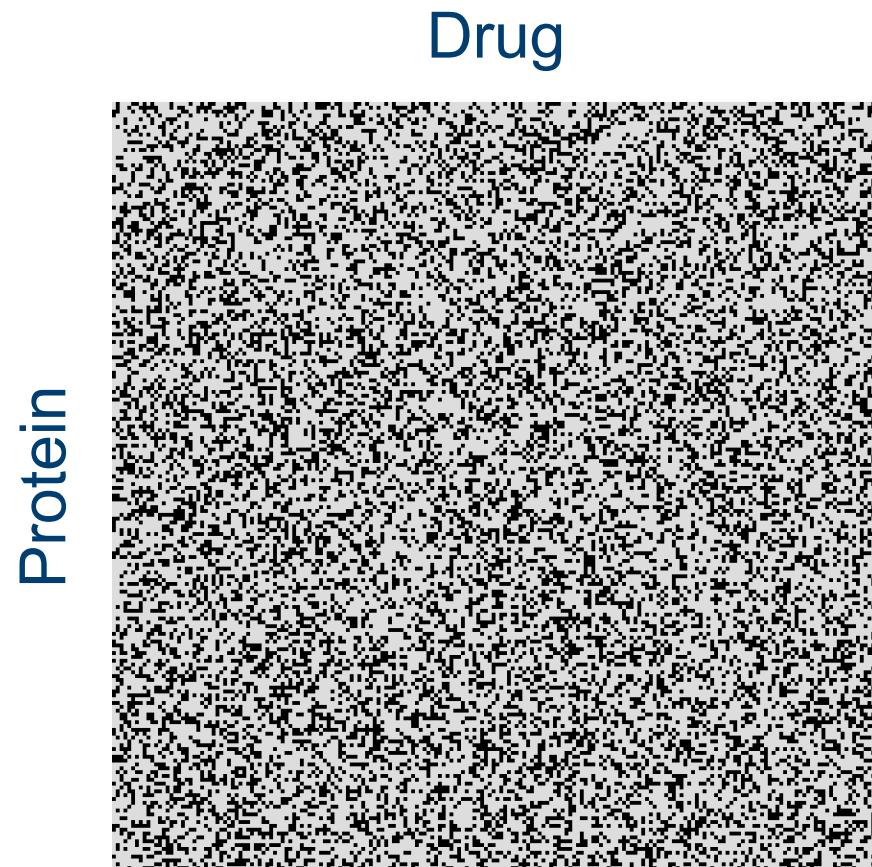
Protein activity data

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 handle fragmented data

Merge simulations and experiments into a holistic tool

Data validation and imputation