

Ontologies for machine learning driven materials design

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Machine learning can

Model datasets where relationships are **complex** and no analytical model exists

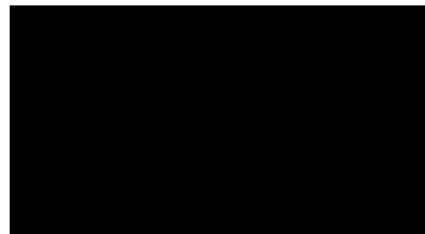
Exploit **property-property** relationships

Merge computer simulations and physical laws with experimental data

Reduce costly experiments to **accelerate** discovery

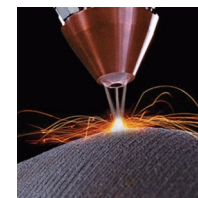
Black box machine learning for materials design

Composition



Properties

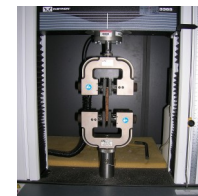
Defects



Fatigue



Strength



Train the machine learning

Composition

70	38	18	40	64	65	00
50	10	66	37	89	02	90
71	52	69	09	46	74	44
01	14	04	49	74	94	80
48	86	85	27	61	10	99
20	33	32	72	19	94	99
97	65	79	34	22	43	41
39	40	46	70	39	60	39



Properties

29	39	28	76	47	90	90
02	13	64	40	10	36	020
63	65	84	97	05	08	18
70	38	18	40	64	65	00
50	10	66	37	89	02	90
71	52	69	09	46	74	44
01	14	04	49	74	94	80
48	86	85	27	61	10	99
20	33	32	72	19	94	99
97	65	79	34	22	43	41
39	40	46	70	39	60	39
59	76	92	86	81	12	39
37	64	13	43	94	87	34
36	65	24	47	27	53	78
14	42	19	81	03	26	61
80	55	56	06	95	26	64
98	34	43	99	48	81	09

Defects

Fatigue

Strength

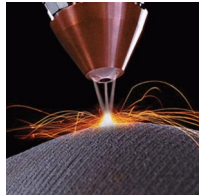
Machine learning predicts material properties

Composition



Properties

Defects



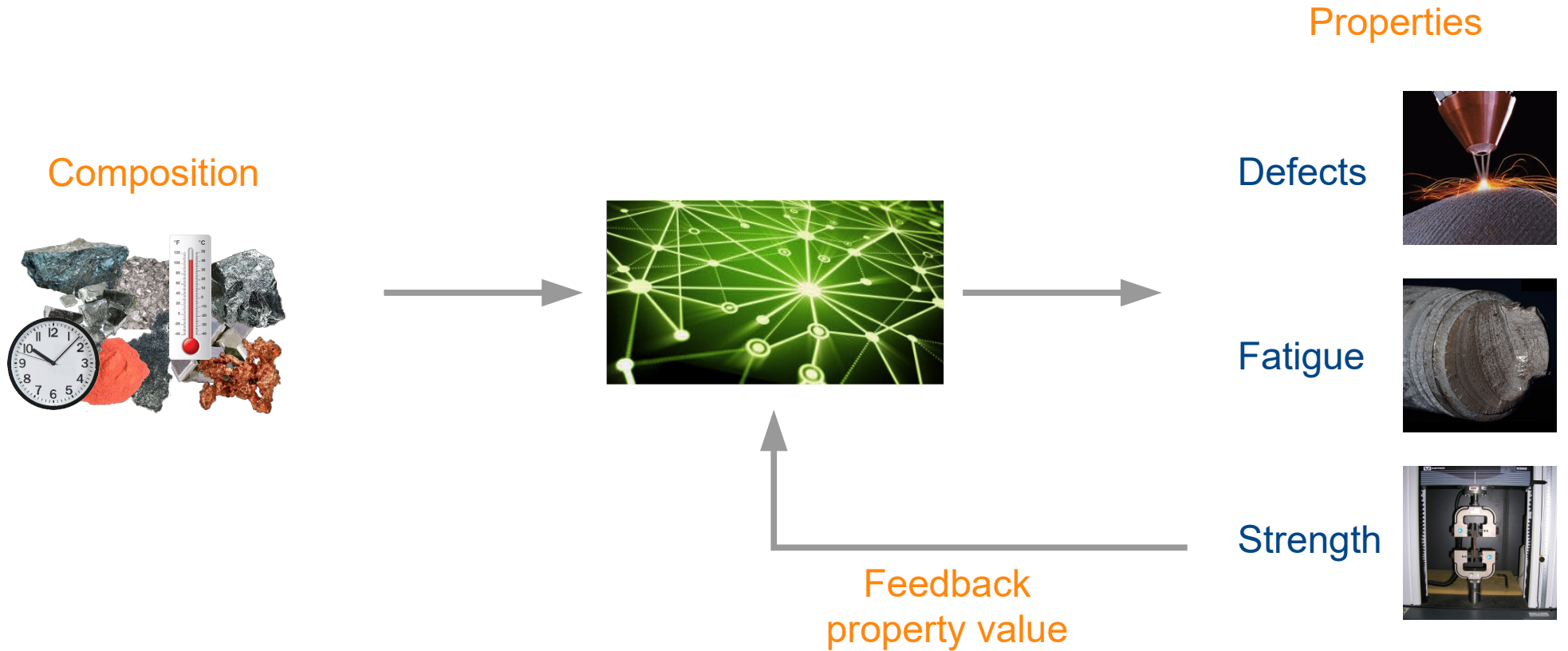
Fatigue

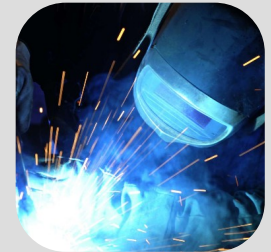
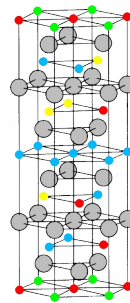
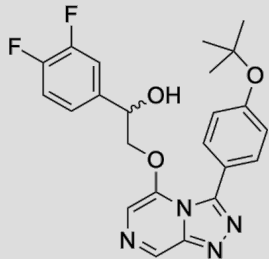
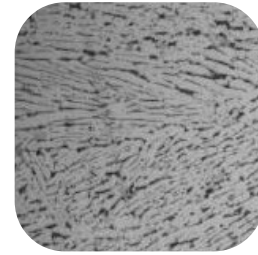
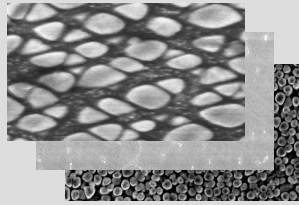
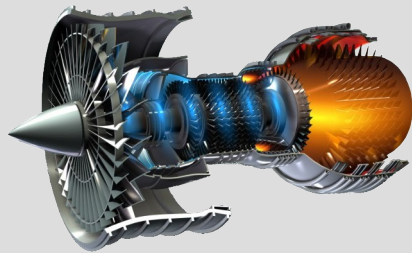


Strength



Exploit property-property relationships





Value of a material through its development cycle

Proposal



10%

Testing

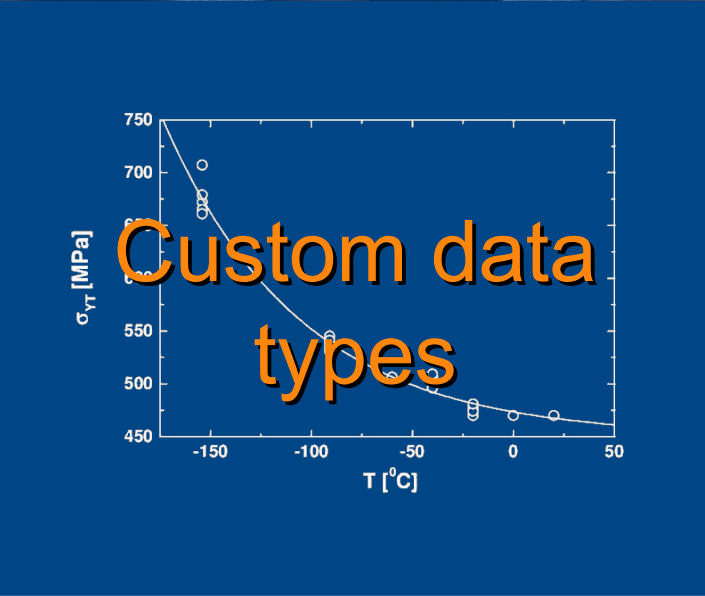
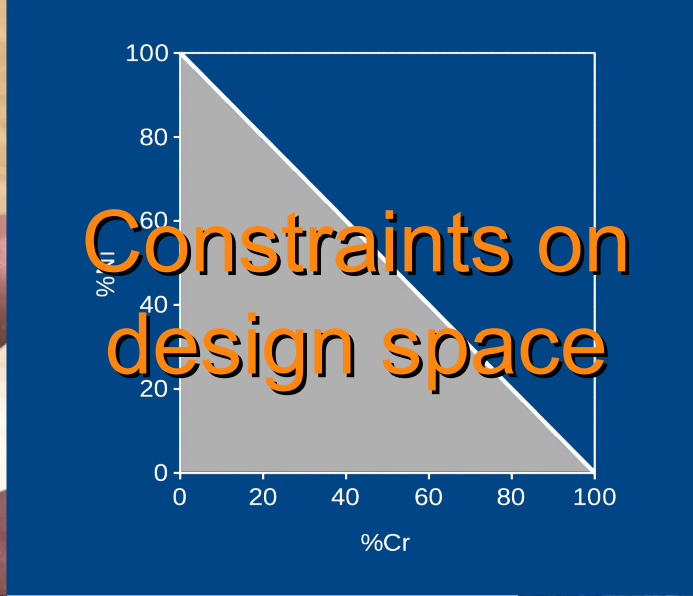
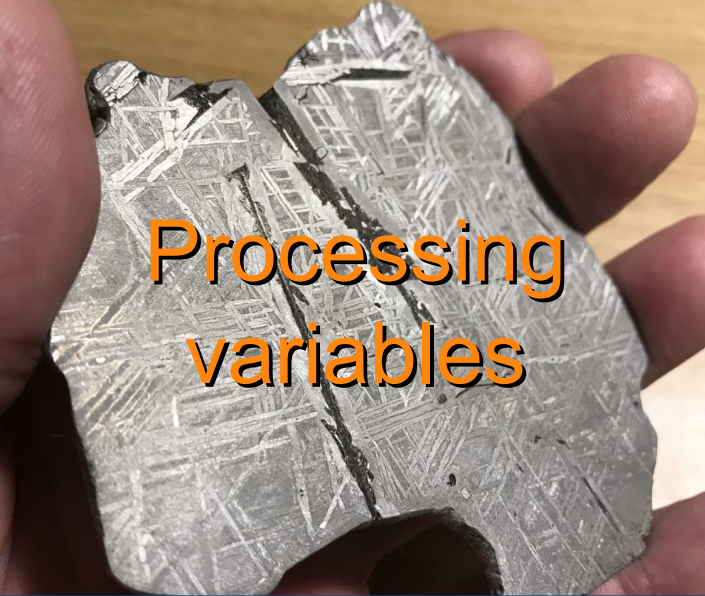


10%

Production



80%



Processing variables

Processing variables e.g. heat treatment time and temperature have significant impact on a material

Additive manufacturing processing variables can change material properties without requiring recertification of the material

Ontology must be sufficiently flexible to encompass these parameters



Widmanstätten pattern in Kamacite meteorite requires cooling over a million years

Esoteric variables

Significant fraction of the value in a material is setting up its mass manufacture to crucial to capture variables unique to a factory

Ontology must be able to adopt bespoke parameters



How many windows are open
in this steel mill?

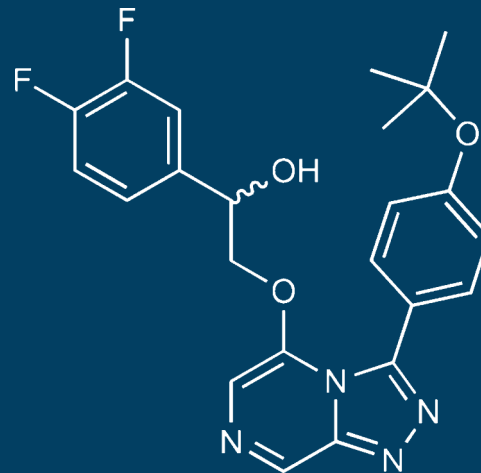
Basis set

Molecular basis set has e.g. 320 components

Over-complete plurality and not orthogonal but contains physically important information for chemists

Framework allows machine learning to exploit property-property relationships

Ingredient has several sub-ingredients, e.g. filler in polymer but some of their sub-ingredients are common e.g. water, which must then be summed over



# O	3	oxygen
# O - H	1	hydroxyl
# C - O - C	2	ether
# C = O	0	carbonyl

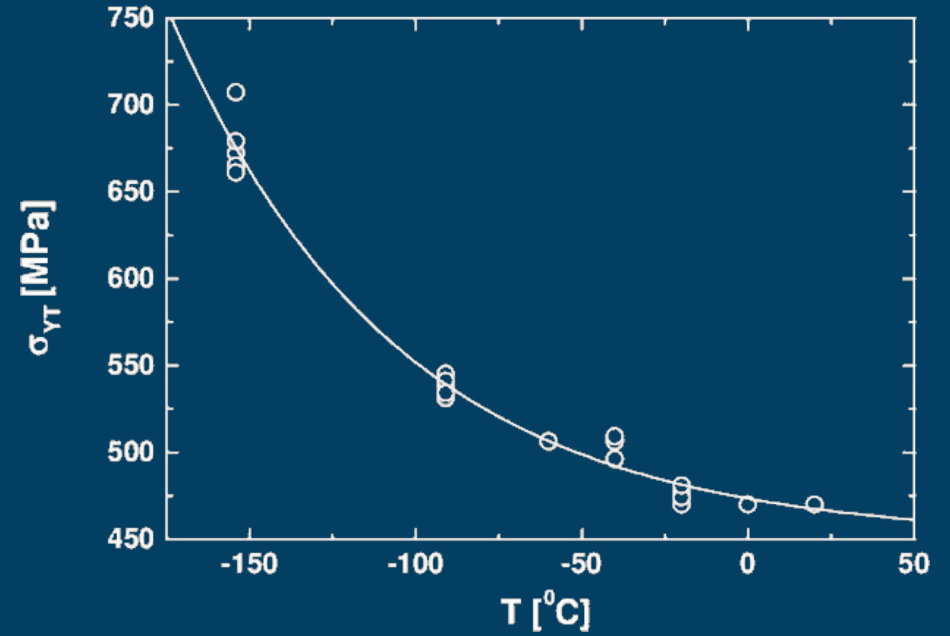
Winner of Open Source Malaria competition
Accepted for publication J. of Medicinal Chem.

Custom data types

Data is often a graph e.g. yield stress vs temperature

Inefficient to store each measurement as a separate material

Helpful to handle such vector data



22NiMoCr37

The “EURO” Fracture Toughness Curve.

Technical Research Centre of Finland

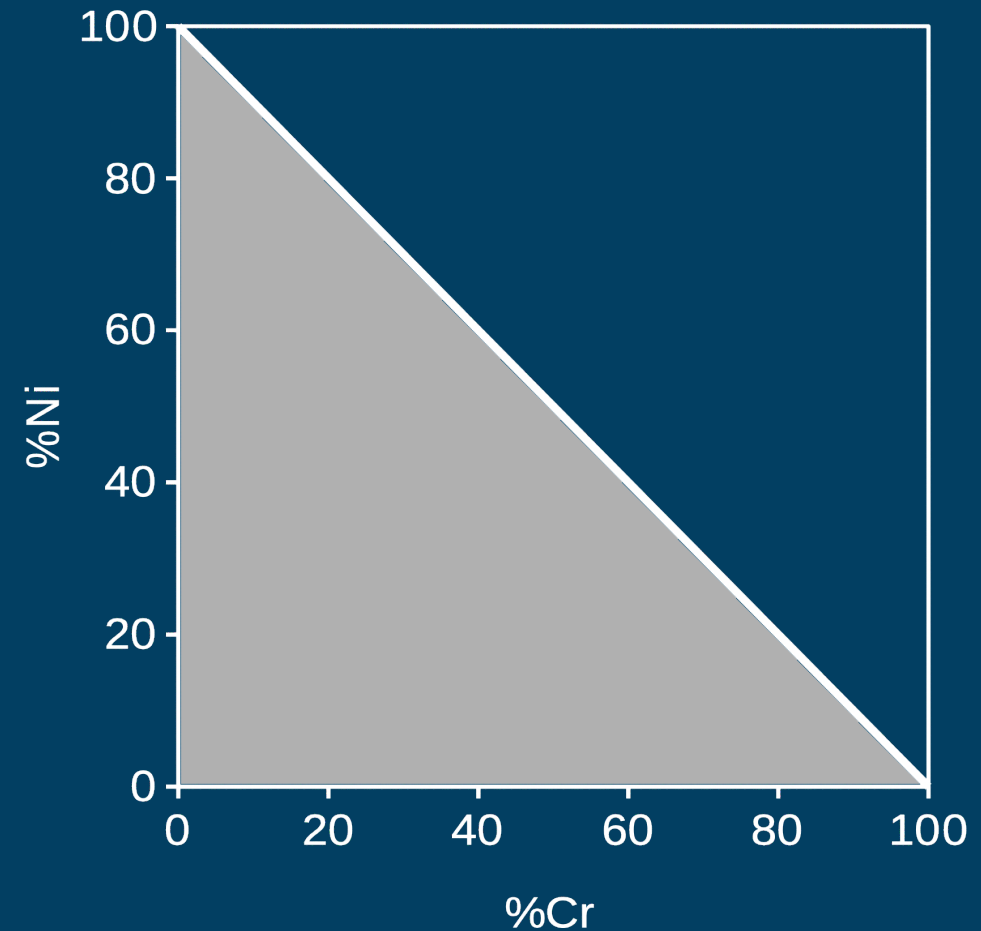
K. Wallin, (1998)

Constraints on design space

When designing materials the opportunity space can have non-trivial constraints

Composition usually has a balance element, remaining elements must sum to less than 100%, defining a simplex

In laser deposition volumetric energy density limited but actual parameter is laser power and part volume



Three component Fe-Ni-Cr alloy with Fe as balance element

Security and provenance

Companies confidential about data so ontology should enable access control

Versioning and referencing of data important to track provenance, quality, and give proper credit



Summary

Machine learning is a powerful tool to capture **complex** relationships

Approach with property-property relationships commercialized by **intellegens**

Highlight six considerations for development of a materials ontology for **industrial applications**

- Processing variables

- Esoteric variables

- Over-complete basis set

- Constraints on design space

- Custom data types

- Security and provenance