

Unveil the unseen: uncover hidden information with machine learning

Gareth Conduit

Model **Sparse** datasets

Exploit property-property relationships

Merge data, computer simulations, and physical laws

Exploit **uncertainties** to deliver most robust predictions

Model **Sparse** datasets

Exploit property-property relationships

Merge data, computer simulations, and physical laws

Exploit **uncertainties** to deliver most robust predictions

Extract information from **noise** itself

Machine learning architecture that can exploit uncertainty



Bogdan Zviazhynski

Unveil the unseen: exploit information hidden in noise, BZ & GJC, Applied Intelligence (2022)

Black box machine learning for materials design



Train the machine learning



Machine learning predicts material properties



Machine learning estimates uncertainty





Handling uncertainty

Unveil the unseen

Design robust formulations

Outlier detection

Design of experiments

Exploit information hidden in noise

Exemplar information extracted from noise

Renormalization group theory applied to phase transitions 1982 Nobel Prize in Physics

Markowitz model 1990 Nobel Memorial Prize





Machine learning exploits uncertainty



Exploit uncertainty to design concrete







Bogdan Zviazhynski

Jess Forsdyke

Professor Janet Lees

Concrete in construction



Cement & aggregate look like noise



Cement & aggregate look like noise



Mission



Design a concrete that is **robust** and **environmentally friendly**

Mission



Design a concrete that is **robust** and **environmentally friendly**

Experimentally validate the concrete

Carbonation is the probe of noise





Depth and uncertainty in carbonation



Machine learning exploits uncertainty



Original model accuracy



Uncertainty improves the model accuracy



Concrete specification



Concrete design



Concrete manufacture



Experimental validation of the proposed mixes

First mix

Second mix



Experiment Model

Target

Exploit uncertainty to predict cancer







Bogdan Zviazhynski

Adriana Fonseca

Dr Jamie Blundell

Screening for disease

Accurate but expensive



Entire population



Screening



Lateral flow 78%



Prostate-specific antigen 25%



Mammogram 87%



Glucose Screening for diabetes 70%







Cytosine nucleobase



Methylation of cytosine







Methylation of cytosine







Required process in mammals, can repress genes

Associated with **Cancer**

Rogue biology causes chaotic methylation resulting in cancer

Methylation of cytosine

40 cohort tracked patients (20 developed acute myeloid leukaemia cancer, 20 did not)



3,054,815,472 base pairs across all chromosomes

Split into chunks



Train a model for methylation density of the chunks



Model for chunk methylation density has uncertainty



Uncertainty is natural owing to permutations of methylation sites



Many permutations are possible



No uncertainty when either fully or not methylated



For each chunk train model for methylation density and uncertainty



Density follows a binomial distribution so its inherent uncertainty is

$$\sqrt{
ho(1-
ho)}$$

Extract the unexplained chaotic methylation due to rogue biology



Identify the critical chunk driving the emergence of cancer



90% accuracy on 40 patients available during study

Blind test on five more patients

Patient	Probability cancerous	Later outcome
19317_U017	0.946	Case
19316_U007	0.451	Control
19317_U015	0.966	Case
19317_U018	0.449	Control
19316_U012	0.934	Case

Further blind tests with data from US collaborators

Explore SCIENCE behind identified critical nucleobase chromosome 3, positions 31800700-31801700

Develop machine learning formalism that can extract information from noise itself

Design and experimentally verify two concrete mixes

Exploit Uncertainty to predict emergence of cancer

Generic approach applicable across the sciences