



The
University
Of
Sheffield.

On the use of Symbolic Regression for Population-Based Modelling of Structures

G. Tsialiamanis, N. Dervilis, K. Worden

IMAC XLII, Orlando, Florida

January 2023



- ▶ Motivation
- ▶ Population-based SHM
- ▶ Symbolic regression
- ▶ Crack-growth application example
- ▶ Conclusions



- ▶ Scarcity of data
- ▶ Motivation from the functionality of physics-based models
- ▶ Structures with common physics
- ▶ Available data from populations of structures
- ▶ Especially in prognosis, damage-evolution data are rare

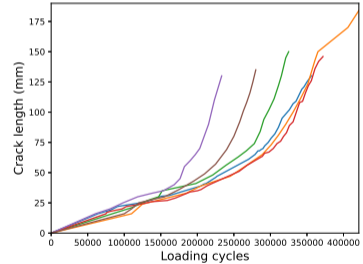
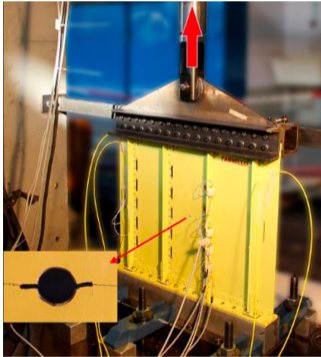


Population-based SHM:

- ▶ Motivated from healthcare practices
- ▶ Powerful transfer learning techniques
- ▶ Reduce the dependence on data
- ▶ Discovery of similar patterns in populations
- ▶ Often the only way to acquire damage-state data

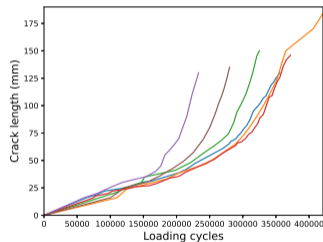


A simple crack-growth population example

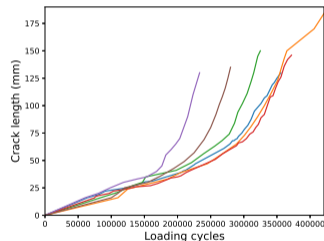


Aims:

- ▶ Transfer learning for regression
- ▶ Attempt **one-to-one** transfer
- ▶ Imitate human inference methodology
- ▶ Discovery of common underlying functional form
- ▶ Learn an underlying skeleton from a **training** structure
- ▶ Fit the trainable parameters to new data from **testing** structure

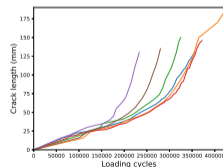
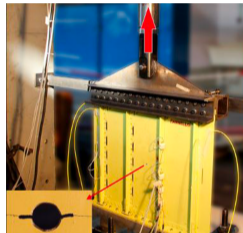


- ▶ Pool of candidate functions
- ▶ Genetic algorithm to create combinations of the basic functions
- ▶ Python implementation PySR [1]
- ▶ Impose prior beliefs via the initial function pool
- ▶ Human-inference inductive bias of smoothness
- ▶ Learn from a **training** structure, test on a **testing** structure



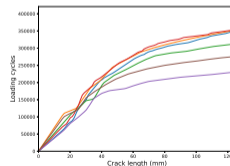
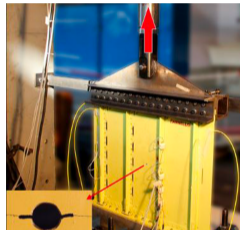
CRACK-GROWTH APPLICATION EXAMPLE

- ▶ Six nominally identical aluminium plates
- ▶ Cyclic loading
- ▶ Initialisation of a small crack
- ▶ Monitoring of the crack length as a function of the loading cycles [2, 3]



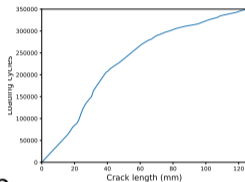
CRACK-GROWTH APPLICATION EXAMPLE

- ▶ Six nominally identical aluminium plates
- ▶ Cyclic loading
- ▶ Initialisation of a small crack
- ▶ Monitoring of the crack length as a function of the loading cycles [2, 3]
- ▶ Inversion of the input and output variables
- ▶ Consider critical crack length of 125mm

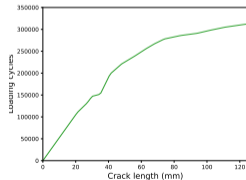


CRACK-GROWTH APPLICATION EXAMPLE

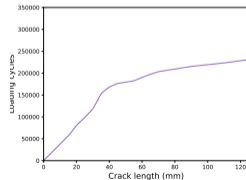
Training plate



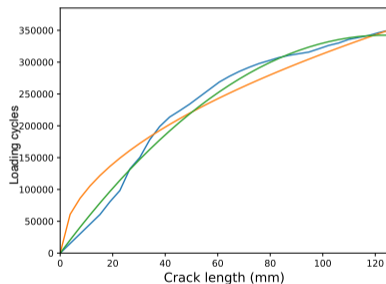
Training plate 2



Training plate 1



Fitted functions to the training data

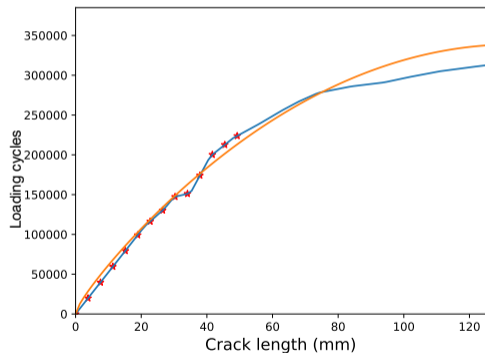


$$N_1 = 1.0221437\sqrt{x}$$
$$N_2 = 1.9781963594752x - x^2$$

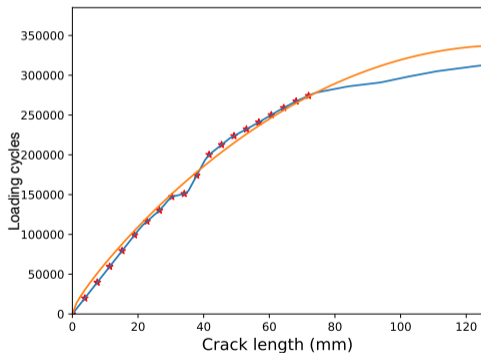


CRACK-GROWTH APPLICATION EXAMPLE

Results 1st testing structure



40% of the curve observed

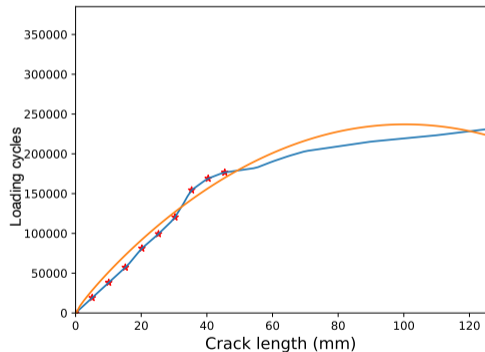


60% of the curve observed

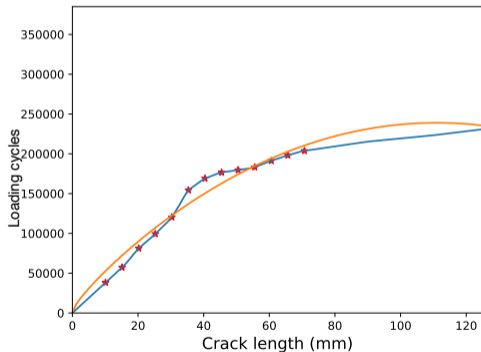


CRACK-GROWTH APPLICATION EXAMPLE

Results 2nd testing structure



40% of the curve observed



60% of the curve observed



- ▶ Promising results for one-to-one regression transfer learning
- ▶ Some artefacts which could be corrected imposing physical knowledge
- ▶ A many-to-one framework could improve the results
- ▶ A Bayesian fitting during testing could provide confidence intervals
- ▶ New deep-learning promising approaches for symbolic regression that scales could lead to improved results



- [1] M Cranmer. “Interpretable machine learning for science with PySR and SymbolicRegression. jl”. In: *arXiv preprint arXiv:2305.01582* (2023).
- [2] C. Sbarufatti, A. Manes, and M. Giglio. “Performance optimization of a diagnostic system based upon a simulated strain field for fatigue damage characterization”. In: *Mechanical Systems and Signal Processing* 40.2 (2013), pp. 667–690.
- [3] M. Corbetta et al. “On dynamic state-space models for fatigue-induced structural degradation”. In: *International Journal of Fatigue* 61 (2014), pp. 202–219.



Thank you!

