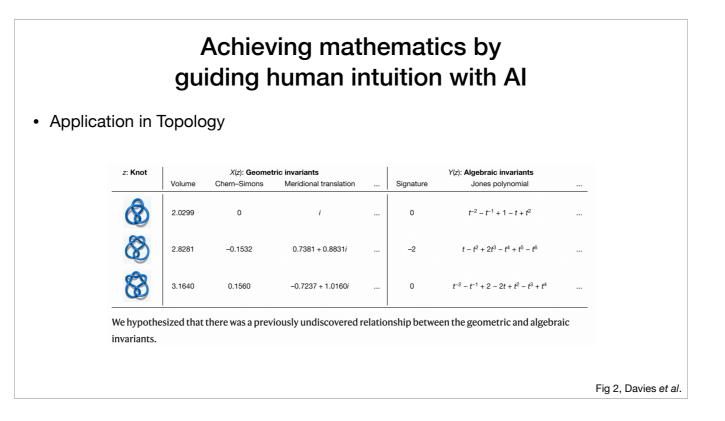
Advancing mathematics by guiding human intuition with AI • Objective of the framework: identify a function which associate X(z) to Y(z) • 2 stages • Supervised learning: train a function f that predicts Y(z) using only X(z) • Attribution techniques: using gradient saliency to calculate the derivative of f(X(z)) with respect to X(z) Generate data (X(z), Y(z))_{z-P2} Train supervised mode $\hat{f}(X(z)) \approx Y(z)$ Find patterns via attribution Interrogate f Hypothesize $\exists f: f(X(z)) = Y(z)$ Conjecture candidate f Alter sampling distribution Mathematician step: Prove theorem Com utational step Fig 1, Davies et al.

Supervised learning:

- · Training data is labelled and we seek to predict outputs given inputs
- · Classification where outputs are discrete
- · Regression where the outputs are real-valued
- Contribution of this stage: learning non-linear functions
- If \hat{f} is more accurate than would be expected by chance, there exists such a relationship between X(z) and Y(z)

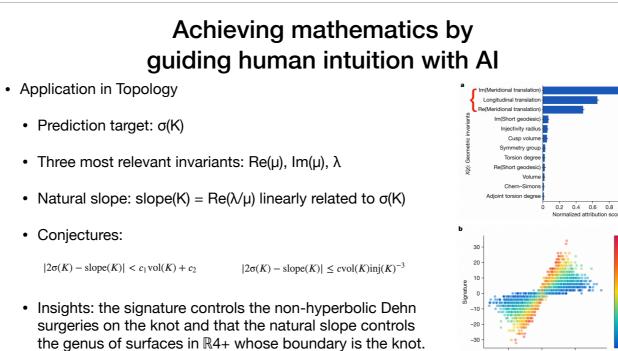
Attribution techniques:

- Quantify component of X(z) that \hat{f} is sensitive to
- Calculate how much \hat{f} changes in predictions of Y(z) given small changes in X(z)



Invariants: geometric or numerical quantities that are the same for two equivalent knots.

Notable example conjecture: the hyperbolic volume of a knot (geometric invariant) should be encoded within the asymptotic behaviour of its coloured Jones polynomials (algebraic invariants).



0.2 0.4 0.6 0.8 1.0 175 150 125 100 Fig 3, Davies et al.

meridional translation μ and the longitudinal translation λ

Dehn surgeries : diff ways of filling in a loop of the meridian

Inj: injectivity radius

Complement 3D space - the knot