

Seminar, WS 22/23 CONSTRUCTIONS IN COMBINATORICS VIA NEURAL NETWORKS Based on (WAGNER, 2021)

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Advancing mathematics by guiding human intuition with Al

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The practice of mathematics involves discovering patterns and using these to formulate and prove conjectures, resulting in theorems. Since the 1960s, mathematicians have used computers to assist in the discovery of patterns and formulation of conjectures¹, most famously in the Birch and Swinnerton-Dyer conjecture², a Millennium Prize Problem³. Here we provide examples of new fundamental results in pure mathematics that have been discovered with the assistance of machine learning-demonstrating a method by which machine learning can aid mathematicians in discovering new conjectures and theorems. We propose a process of using machine learning to discover potential patterns and relations between mathematical objects, understanding them with attribution techniques a using these observations to guide intuition and propose conjectures. We outline t machine-learning-guided framework and demonstrate its successful application current research questions in distinct areas of pure mathematics, in each case showing how it led to meaningful mathematical contributions on important op problems: a new connection between the algebraic and geometric structure of and a candidate algorithm predicted by the combinatorial invariance conject symmetric groups⁴. Our work may serve as a model for collaboration betwee fields of mathematics and artificial intelligence (AI) that can achieve surpris by leveraging the respective strengths of mathematicians and machine lea

One of the central drivers of mathematical progress is the discovery of patterns and formulation of useful conjectures: statements that are suspected to be true but have not been proven to hold in all cases. Mathematicians have always used data to help in this process-from the early hand-calculated prime tables used by Gauss and others that led to the prime number theorem5, to modern computer-generated data¹⁵ in cases such as the Birch and Swinnerton-Dyer conjecture².

that AI can also be used to assist in the discovery of the jectures at the forefront of mathematical research. T using supervised learning to find patterns²⁰⁻²⁴ by foc mathematicians to understand the learned function mathematical insight. We propose a framework f standard mathematician's toolkit with powerful and interpretation methods from machine learn its value and generality by showing how it led t

nature Article citing Wagner

The introduction of computers to generate data and test conjectures

In New Math Proofs, Artificial Intelligence

A new computer program fashioned after artificial intelligence system like AlphaGo has solved several open problems in combinatorics and



techniques. Instead, he used a game-playing machine "I was very happy to have the question answered. I was excited that

VIEW POSIPHINT MODE

Adam had done it with AI," said Hogben.

Hogben and Reinhart's problem was one of four that <u>Wagner solved</u> using artificial intelligence. And while AI has contributed to mathematics before, Wagner's use of it was unconventional: He turned the hunt for solutions to Hogben and Reinhart's question into a kind of contest, using an approach other researchers have applied with great



lewsletter



quantamagazine.com Article dedicated to Wagner's work

Generating structures using reinforcement learning

- Encode instances by fixed length word w over fin. alphabet Example: Graph on n vertices as $w \in \{0,1\}^{\binom{n}{2}}$
- Start with empty word
- In *i*-th step:



Generating structures using reinforcement learning

Deep cross-entropy method:

- \bullet Generate N>0 instances
- For each instance: Evaluate score function Ex.: *"How close is instance to conjectured bound?"*
- For top y percentage of instances:

Training: Fit predictor on pairs

 $((w_1, ..., w_{i-1}, 0, ..., 0), e_i) \rightarrow e_{w_i}$

 Keep top x < y percentage of instances for next iteration(s) Generating structures using reinforcement learning

Architecture of predictor

- Neural network with three hidden layers: dense layers with 128 / 64 / 4 nodes activation function: ReLU
- Output layer: sigmoid (binary case)
- Loss function: Cross entropy
- Optimizer: SGD

Example

M. Aouchiche and P. Hansen, A survey of automated conjectures in spectral graph theory, 2010:

Conjecture: G connected graph, $n \ge 3$, largest eigenvalue λ_1 , matching number μ . Then:

$$\lambda_1 + \mu \ge \sqrt{n-1} + 1$$

Apply cross-entropy method: Fix n, minimize score function $\lambda_1 + \mu$

Example

For n = 19, average score after many iterations:



Pattern avoiding 0-1-matrices

Definition: 0-1-matrix patterns

The 0-1-matrix $A \in \{0,1\}^{r \times s}$ contains 0-1-matrix $P \in \{0,1\}^{k \times l}$, if there exists a submatrix $D \in \{0,1\}^{k \times l}$ with $P \leq D$. Otherwise, A avoids P.

Example:





Pattern avoiding 0-1-matrices

- $P_{\pi} := [\delta_{j,\pi(i)}]_{i,j} \in \{0,1\}^{n \times n}$ permutation matrix of $\pi \in S_n$
- Def: 0-1-matrix M contains (resp. avoids) $\pi \in S_k$, if M avoids (resp. contains) P_{π} .
- **Example:** 312-avoiding 0-1-matrices:



• **Obs.:** P_{π} avoids P_{σ} iff π avoids σ as permutation pattern.

Permanent in pattern avoiding 0-1-matrices Motivation: If $A \in \{0,1\}^{n \times n}$ avoids π , then the permanent

$$per(A) := \sum_{\sigma \in S_n} \prod_{i=1}^n a_{i,\sigma(i)}$$

counts (π -avoiding) permutations S_n contained in A.

Question (Brualdi & Cao, 2020): Given $n \in \mathbb{N}$ and $\pi \in S_k$, what is the value of

$$f_{\pi}(n) := \max\left\{ \operatorname{per}(A) : A \in \{0, 1\}^{n \times n}, A \text{ avoids } \pi \right\}$$

Here: Wagner finds bounds on f_{312} .

Permanent in pattern avoiding 0-1-matrices

Using the cross-entropy method, find 312-avoiding matrices with high permanent:



Breaks conjectured value $f_{312}(5) = 12$ by Brualdi and Cao!

Permanent in pattern avoiding 0-1-matrices

Theorem:
$$2^{0.88n} \le f_{312}(n) \le 24^{n/4} \approx 2^{1.15n}$$

Proof of lower bound:



Observation:

- $\operatorname{per}(A * B) \ge \operatorname{per}(A) \cdot \operatorname{per}(B)$
- $f_{312}(n+m) \ge f_{312}(n) \cdot f_{312}(m)$

Idea: Use $per(A_{13}) = 2809 > 2^{0.88 \cdot 13}$ and Fekete's Lemma.

Proof of upper bound:

Theorem (Bregman & Minc, 1973): For $A \in \{0,1\}^{n \times n}$ with row sums $r_1, ..., r_n$: $per(A) \leq \prod_{i=1}^n (r_i!)^{1/r_i}$

Proposition (Brualdi & Cao, 2020): At most 4n - 4 one entries in 312-avoiding matrix $M \in \{0, 1\}^{n \times n}$.

Idea: rhs. expression maximized when all $r_i = 4$.

Organization of the seminar

For obtaining 6 LP:

- Read Wagner's paper, understand the method
- Pick a problem and think of an Al-attack:
 - Encoding of instances
 - Variation of Wagner's method
- Send me a short draft of your approach
- Set up the environment, implement your approach (and modifications of it)
- Give a short presentation in the end of the semester about the problem, the approach and the result.

Don't avoid questions!

