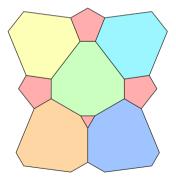


Neural Discovery in Mathematics

DOML 2025 @ RIMS

Christoph Spiegel

15th of May 2025





Results are joint work with...









Aldo Kiem Zuse Institute Berlin Technische U. Berlin

Konrad Mundinger Zuse Institute Berlin Technische U. Berlin

Sebastian Pokutta Zuse Institute Berlin Technische U. Berlin Max Zimmer Zuse Institute Berlin Technische U. Berlin



Neural Discovery in Mathematics

1. Proof by Picture in Extremal Combinatorics	3 slides
2. Constructions through Implicit Representation	3 slides
3. Applications to Hadwiger-Nelson	4 slides
4. Applications to Graph Theory	2 slide



What do extremal combinatorialists care about?

Extremal Graph Theory, on the most general level, investigates the extremal (maximal or minimal) value of various graph parameters over the family of graphs having a particular property. It is a lively subject with a rich history, where numerous natural questions have beautiful answers. It is a field very much driven by problems; many of the interesting ones are still wide open and stimulate an abundance of research.

Each such problem has two sides: one is the construction of an extremal structure, the other is the proof of its optimality. In this course we are putting extra emphasis on explicit constructions of extremal graphs, which do not customarily feature in standard treatments of the field. These constructions often require useful tools from algebra, geometry, or discrete Fourier analysis; the other main objective of these notes is to highlight them.

p.7 of LET'S BE EXPLICIT! lecture notes by Tibor Szabó, July 2024



Example 1: Graphs and graph sequences

Question. How many edges can a graph G of order n with $\omega(G) < r$ have?

Theorem (Mantel, 1907; Turán, 1941; Erdős-Stone, 1946)

At most $(1 - 1/r + o(1))\binom{n}{2}$, i.e., as many as the Turán graph T(n, r).

Question. How large can the order *n* a graph *G* with $max(\alpha(G), \omega(G)) < k$ be?

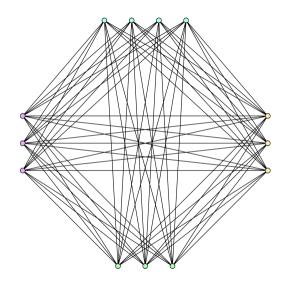
Theorem (Ramsey, 1930; many others)

We know that R(3) = 6, R(4) = 18, $43 \le R(5) \le 46$, and $2^{k/2} \le R(k) \le 3.78^k$.

A variant. How few cliques and independent sets of size r can a graph contain?

Theorem (Goodman, 1959)

Asymptotically at least 25% of all triangles need to be cliques or independent sets.





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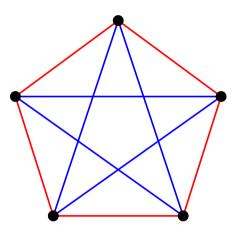
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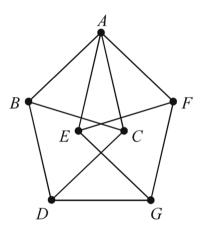
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Remark. Considering the infinite graph with vertex set \mathbb{E}^2 and edges $\{x, y\}$ for any ||x - y|| = 1, we are studying the **chromatic number of the plane** $\chi(\mathbb{E}^2)$.

Theorem (Aubrey D.N.J. de Grey, 2018)

There is a unit distance graph on 20425 vertices with chromatic number 5.

$$5 \leq \chi(\mathbb{E}^2) \leq \dots$$





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A Colorful Unsolved Problem -Numberphile

681K views • 5 years ago

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More links & stuff in full description below 111 Numberphile is supported by the Mathematical Science...

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GRAPH THEORY

Decades-Old Graph Problem Yields to Amateur Mathematician

By EVELYN LAMB | APRIL 17, 2018 | 🔲 26 | 📕

...number of vertices? The problem, now known as the Hadwiger-Nelson problem or the problem of finding the chromatic number of the plane, has piqued the interest of many mathematicians, including...





Aubrey de Grey and Alexander Soifer, Il Vicino, January 18, 2020



Ronald L. Graham presents Aubrey D.N.J. de Grey the Prize: \$1000, San Diego, September 22, 2018



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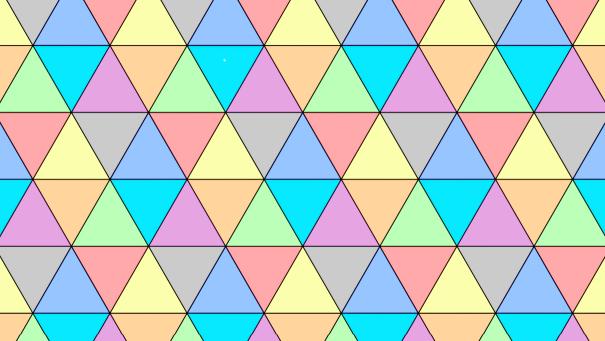
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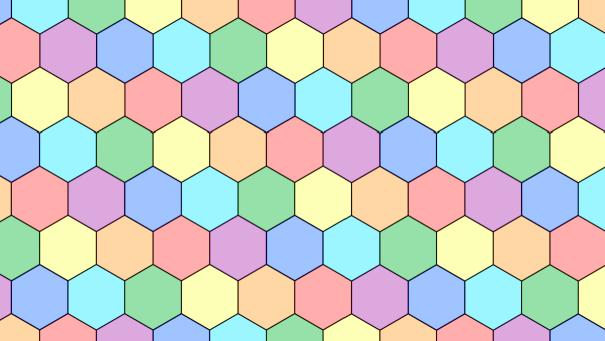
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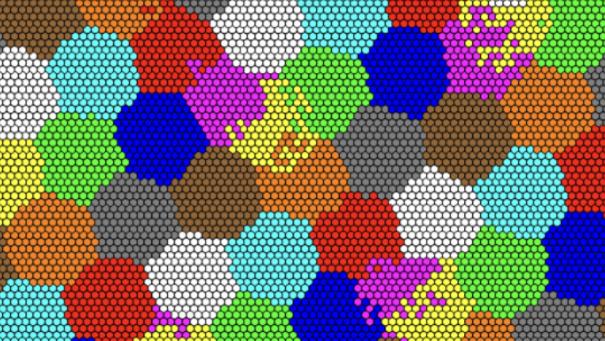
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Idea. Consider a probabilistic relaxation to functions $p: \mathbb{E}^2 \to \Delta_c$ minimizing the loss

$$L_R(p) := \int_{[-R,R]^2} \int_{\partial B_1(x)} p(x)^T p(y) \, \mathrm{d}y \, \mathrm{d}x.$$
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Challenge. Can we find a *parameterized and (easily) differentiable* family p_{θ} , i.e., an implicit representation, and optimize Equation (1) over θ through gradient descent?





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A few candidates: Taylor or Fourier Series, Splines, Wavelets, SVM, RBF, ... but we chose simple feed-forward Neural Networks with sinus activations:

• Efficient gradients through backprop

- Natural spectral bias towards low frequencies (Rahaman et al., 2019)
- Implicit representation with NNs is SOTA in physics-informed learning

- Sin activation performs well for implicit repr. (Sitzmann et al., 2020)
- Mature software (PyTorch) and hardware (GPU) ecosystem
- Universal approximation

Theorem (Universal Approximation Theorem)

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How do we find the correct parameters?

Algorithm. We used *batched gradient descent* to 'train' p_{θ} to minimize $L_R(\theta)$. At each step we sample points $x^{(i)} \in [-R, R]^2$ and $y^{(i)} \in \partial B_1(x^{(i)})$ and use the fact that

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to adjust the parameters θ with an appropriate step size α_k through

$$\theta_{k+1} = \theta_k - \alpha_k \,\hat{\nabla}_\theta L(\theta).$$

- MLP with sinus activation functions and two hidden linear layers à 256 neurons.
- We sampled around 2^{12} pairs for each step for a total of around 2^{26} samples.
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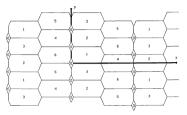
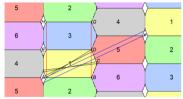


FIG. 3. A good 7-coloring of (R², 1).





Theorem (Pritikin 1995; refined by Parts 2020)

99.985% of the plane can be colored with 6 colors while avoiding unit distances. This implies that any unit distance graph with chromatic number 7 must have order ≥ 6993 .

But the principle works! Can we study some variants of the original problem?



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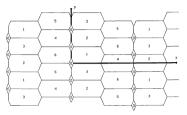
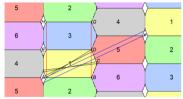
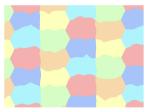


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 Applications to Hadwiger-Nelson 	
2. Constructions through Implicit Representation	3 slides
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Variant 1: Almost succeeding...

Question. What is the smallest percentage of the plane that needs to be removed so that we can color the rest with $1, 2, \ldots, 6$ colors without monochromatic conflicts?

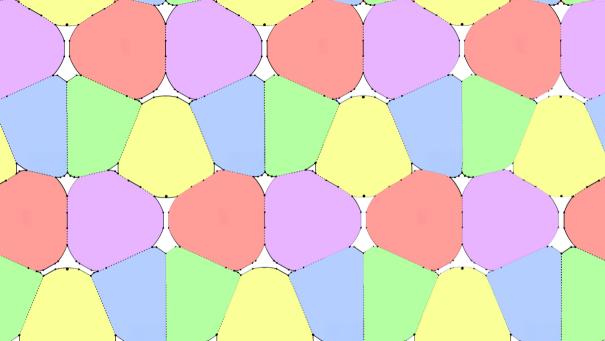
colors	1	2	3	4	5	6
best known	74.56%	54.13%	31.20%	8.25%	4.01%	0.02%
numerics	75.86%	54.14%	31.23%	8.27%	3.56%	0.02%

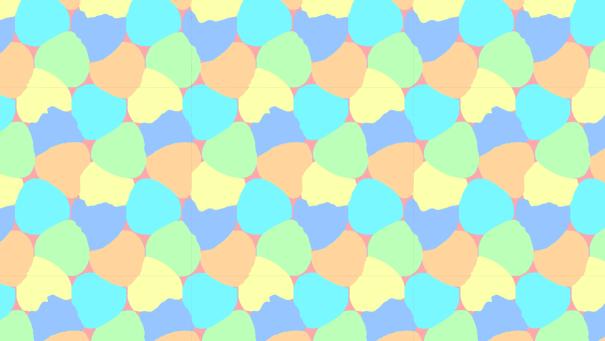
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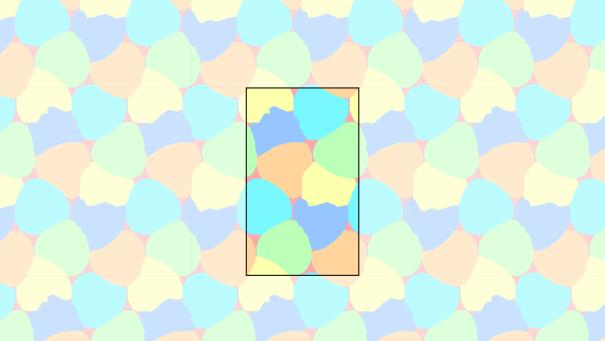
Theorem (Mundinger, Pokutta, S., Zimmer 2025+)

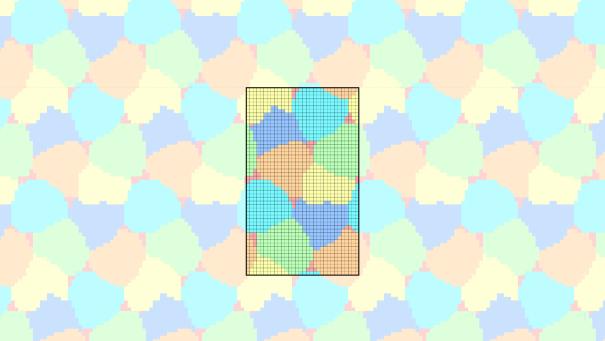
96.29% of the plane can be 5-colored with no monochromatic unit distance pairs.

Remark. We can also color $\sim 95\%$ of \mathbb{E}^3 using 14 colors (not yet formalized).











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Variant 2: Going off-diagonal...

A *c*-coloring realizes (d_1, \ldots, d_c) if color *i* does not contain distance d_i .

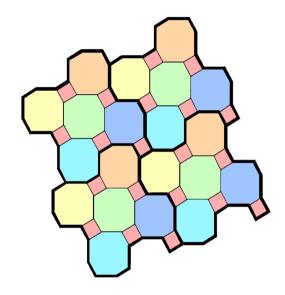
Problem (The continuum of six-colorings; Soifer in Nash and Rassias' *Open Problems in Mathematics*) *Determine the set of d for which* (1, 1, 1, 1, 1, d) *can be realized.*

Soifer (1991) found a coloring for $d = 1/\sqrt{5}$. Hoffman and Soifer (1993) also found one for $d = \sqrt{2} - 1$. Both of these are part of a family that covers any

$$0.414 \approx \sqrt{2} - 1 \le d \le 1/\sqrt{5} \approx 0.447.$$

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We extended the range of realizable types to $0.354 \le d \le 0.553$.





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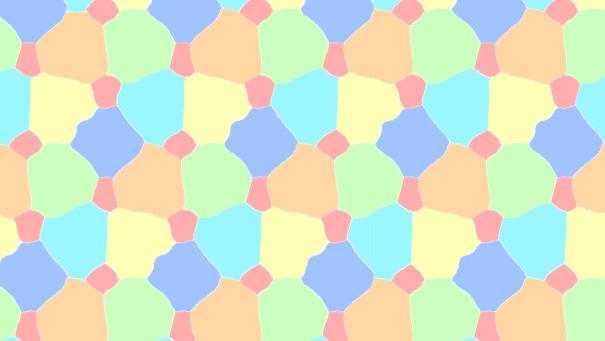
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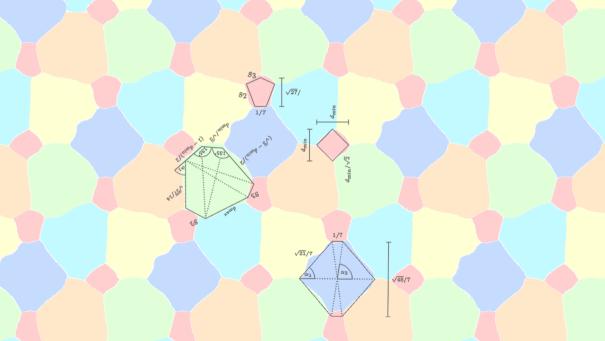
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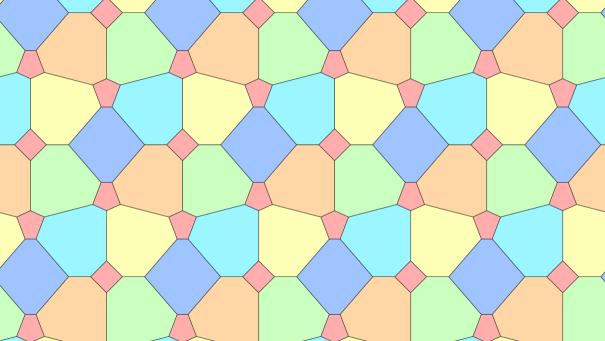
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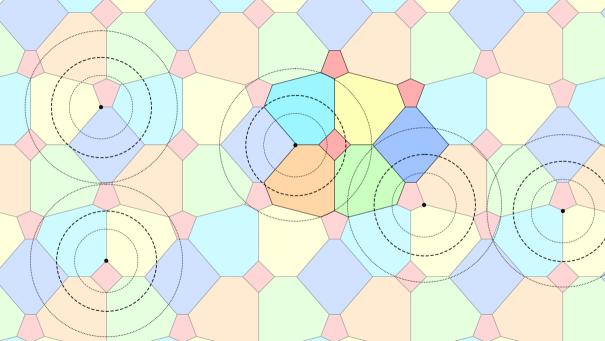
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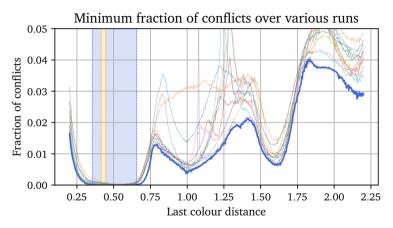
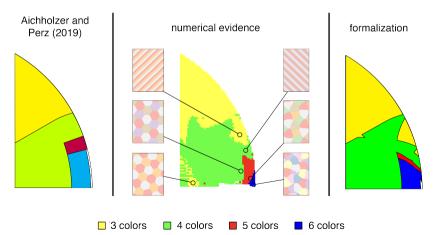


Figure: Numerical results showing the percentage of points with some conflict for a given forbidden distance in the sixth color minimized over several runs.



Variant 3: Three points are more than two...

Question. With how many colors can we color the plane while avoiding three points of the same color forming a triangles with edge lengths $0 \le a \le b \le 1$?





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Neural Discovery in Mathematics

4. Applications to Graph Theory	2 slide
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Graph sequences as continuous objects

Aren't graphs ... discrete? Yes, but ...

 We can formulate a probabilistic relaxation through a *random graph* model.
 Graphons (Lovasz and Szegedy, 2004) tell us that symmetric and measurable *W* : [0, 1]² → [0, 1] correspond bijectively to convergent graph sequences.

You can think of graphons as adjacency matrices viewed as black-and-white images.

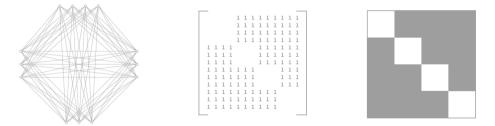


Figure: (left) T(13,4) (center) adjacency matrix (right) graphon of $(T(n,4))_{n\in\mathbb{N}}$



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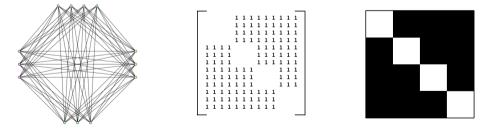


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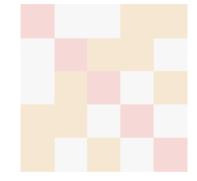


Figure: Result of maximizing the number of edges while penalizing cliques of size 5 with a Lagrangian term.



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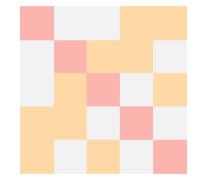


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Is the same approach applicable?



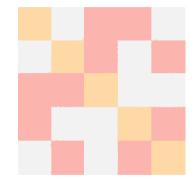
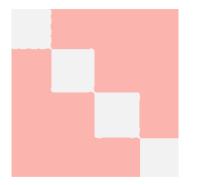


Figure: Result of maximizing the number of edges while penalizing cliques of size 5 with a Lagrangian term.



Is the same approach applicable?



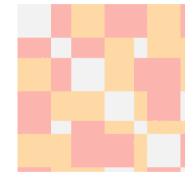


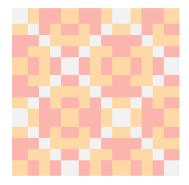
Figure: Result of maximizing the number of edges while penalizing cliques of size 5 with a Lagrangian term.



Is the same approach applicable?



Figure: Result of maximizing the number of edges while penalizing cliques of size 5 with a Lagrangian term.





Thank you for your attention!

A description of the methodology was accepted at ICML 2025 and is available at arxiv.org/abs/2501.18527.

A description of the two colorings was published by Geombinatorics Quarterly and is available at arxiv.org/abs/2404.05509.

Descriptions of the results for almost-colorings and triangle-free colorings are in preparation.