Towards generalized spectral determinacy of random graphs



Alexander Van Werde, Algebraic Graph Theory Seminar (2025)

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Can *all* information be recovered from spectrum?



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Conjecture: almost all graphs are determined by spectrum.

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2023 (Koval and Kwan)

At least exp(cn) graphs are determined by spectrum.

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Some history: generalized spectral determinacy

1980 (Johnson and Newman)

"It is our view, however, that to some extent these examples are algebraic accidents due to the interpretation of the formal symbols 0 and 1 as real numbers."

Definition. (Generalized cospectral)

Graphs G, H are said to be generalized cospectral if

$$\operatorname{spec}(A_{G}^{x,y}) = \operatorname{spec}(A_{H}^{x,y}) \quad \forall x, y \in \mathbb{R}.$$

where $A_G^{x,y}$ is the variant on the adjacency matrix with $1 \rightarrow x$ and $0 \rightarrow y$.

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Definition (Walk matrix)

Given an integer matrix $X \in \mathbb{Z}^{n \times n}$, consider the matrix

$$\pmb{W}\coloneqq [e,\pmb{X}e,,\pmb{X}^2e,...,\pmb{X}^{n-1}e]$$

where $e = (1, ..., 1)^{T}$.

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where $e = (1, ..., 1)^{T}$. Interpret $X_{i,i}$ as edge multiplicity. Then, $W_{i,i}$ counts walks of length j - 1 ending in *i*. 3 Example. $W_{5,3} = 3 \cdot 1 = 3$

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Notation

 $W(\mathbb{Z}^n) \coloneqq \{Wv : v \in \mathbb{Z}^n\} \qquad \text{and} \qquad \operatorname{coker}(W) \coloneqq \mathbb{Z}^n / W(\mathbb{Z}^n) \\ = \{p(X)e \colon p \in \mathbb{Z}[x]\},$

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Given an Abelian group G and a prime power p^m , let $G_{p^m} \coloneqq G/p^m G$.

Theorem. (Wang 2017; see also Qui, Wang, and Zhang 2023) Consider a simple graph G and set $X \coloneqq A_G$. Assume that $\operatorname{coker}(W)_{2^2} \cong (\mathbb{Z}/2\mathbb{Z})^{\lfloor n/2 \rfloor}$ and $\operatorname{coker}(W)_{p^2} \in \{0, \mathbb{Z}/p\mathbb{Z}\}$ for odd primes p.

Then, G is determined by generalized spectrum.

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Suppose *X* is random.

How can we study the distribution of coker(*W*)?

Disclaimer.

For technical reasons, all results in this talk assume that X has independent entries.

This implies that we can not (yet) deal with the adjacency matrices of *simple* random graphs: those have dependent entries due to the symmetry constraint $X = X^{T}$.

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Assumption 1<sup>st</sup> result
Fix a prime p and integer m \ge 0.
Assume that the entries of X are independent
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Assumption 1st result

X has independent Unif $\{0, 1, ..., p^m - 1\}$ -distributed entries.

Theorem 1.

We have

$$\lim_{n \to \infty} \mathbb{P}\left(\operatorname{coker}(\boldsymbol{W})_{p^{m}} \cong \bigoplus_{i=1}^{\ell} \frac{\mathbb{Z}}{p^{\lambda_{i}}\mathbb{Z}}\right) = \prod_{i=i_{0}}^{\infty} \left(1 - p^{-(i+1)}\right) \prod_{j=1}^{\ell} p^{-j\delta_{j}}$$
for every $0 = \lambda_{0} \le \lambda_{1} \le \dots \le \lambda_{\ell} \le m$.

Here,
$$i_0 \coloneqq #\{1 \le i \le \ell : \lambda_i = m\}$$
 and $\delta_j = \lambda_{\ell-j+1} - \lambda_{\ell-j}$.

Proof idea



Proof idea



Key observation. (Informally)

Aside from the obstruction above, there is independence.

Interpretable proof!

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Sadly, the technique is not robust.

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How can we study *unweighted* graphs?

Assumption simplified 2nd result

Suppose X has independent $\{0,1\}$ -valued entries. (Not necessarily identically distributed.)

Further, consider a sparse setting: $\mathbb{P}(X_{i,j} = 1) \leq \mathbb{P}(X_{i,j} = 0)$

But not *too* sparse: $\mathbb{P}(X_{i,j} = 1) \gg \ln(n)/n$



Assumption simplified 2nd result

X has independent {0,1}-valued entries with $\mathbb{P}(X_{i,j} = 1) \leq \mathbb{P}(X_{i,j} = 0)$ and $\mathbb{P}(X_{i,j} = 1) \gg \ln(n)/n$

Technical condition.

Additionally assume tightness:

С

$$\lim_{n\to\infty} \liminf_{n\to\infty} \mathbb{P}\big(\# \operatorname{coker}(W)_{p^m} \leq C \big) = 1.$$

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$$\lim_{C \to \infty} \liminf_{n \to \infty} \mathbb{P}\big(\# \operatorname{coker}(W)_{p^m} \leq C \big) = 1.$$

Theorem 2. (Simplified)

Fix a finite collection of primes \mathcal{P} .

Then, given the conditions above,

1. The same limiting law applies to $\operatorname{coker}(W)_{p^m}$ for every $p \in \mathcal{P}$.

2. We have asymptotic independence for different primes $p \in \mathcal{P}$.

Robust proof technique:

category-theoretic moment method.

Category-theoretic moment method

Definition. (Category-theoretic moment)

Consider a ring R, a deterministic R-module N, and a random R-module Y.

Then, the *N*-moment of *Y* is $\mathbb{E}[\#Sur_R(Y, N)]$.

Theorem. (Sawin and Wood, 2022)

Consider a random *R*-module *Y* and a sequence of random *R*-modules Y_n .

Then, under certain conditions, to prove that $Y_n \to Y$ in distribution it suffices to show that $\lim_{n \to \infty} \mathbb{E}[\#\operatorname{Sur}_R(Y_n, N)] = \mathbb{E}[\#\operatorname{Sur}_R(Y, N)]$ for every fixed finite *R*-module *N*.

Category-theoretic moment method

We show that $\mathbb{E}[\#\operatorname{Sur}_{\mathbb{Z}[x]}(\operatorname{coker}(W), N)] = (\#N)^{-1}$ for every finite $\mathbb{Z}[x]$ -module N.

Related problems were studied by e.g., Nguyen and Wood (2022) and Cheong and Yu (2023).

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Proof sketch.

Using that group morphism $F: \mathbb{Z}^n \to N$ descends to $\mathbb{Z}[x]$ -module morphism from coker(W) if and only if F(e) = 0 and FX = xF, $\mathbb{E}[\#\operatorname{Sur}_{\mathbb{Z}[x]}(\operatorname{coker}(W), N)] = \sum_{F \in \operatorname{Sur}_{\mathbb{Z}}(\mathbb{Z}^n, N): F(e) = 0} \mathbb{P}(FX = xF).$

There are approximately $(\#N)^{n-1}$ summands since $\#\{F \in Sur_{\mathbb{Z}}(\mathbb{Z}^n, N): F(e) = 0\} \approx \#\{F \in Hom_{\mathbb{Z}}(\mathbb{Z}^n, N): F(e) = 0\}.$

For typical *F*, one has $\mathbb{P}(FX = xF) \approx (\#N)^{-n}$.

Thank you!

Key reference related to this talk are as follows:

Generalized spectral determinacy:

W. Wang and C.-X. Xu. A sufficient condition for a family of graphs being determined by their generalized spectra. European Journal of Combinatorics, 2006.

W. Wang. A simple arithmetic criterion for graphs being determined by their generalized spectra. Journal of Combinatorial Theory, Series B, 2017.

L. Qiu, W. Wang, and H. Zhang. *Smith normal form and the generalized spectral characterization of graphs.* Discrete Mathematics, 2023

Category-theoretic moment method:

W. Sawin and M.M. Wood. *The moment problem for random objects in a category.* arXiv:2210.06279v1, 2022.

The current work:

A. Van Werde. *Cokernel statistics for walk matrices of directed and weighted random graphs.* Combinatorics, Probability and Computing, 2025

Feel free to contact me!

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