#### **RECONSTRUCTING RANDOM PICTURES**

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AMS Central Sectional

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#### **Reconstruction Problem**

Given a discrete structure, can we uniquely reconstruct it from the

list of its substructures of a fixed size?

Most famous example: graphs—Vertex and Edge Reconstruction

Conjectures (Kelly, Ulam 1957, Harary 1964)

#### Mossel-Ross '18

What about "shotgun assembly?" (motivated by shotgun sequencing of DNA)



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Today: Let  $P_n$  be a random picture, i.e. an  $n \times n$  grid with  $\{0, 1\}$  entries chosen uniformly at random. Let D be the deck of its  $k \times k$  subgrids.

#### Question

For what k = k(n) is  $P_n$  reconstructible from  $\mathcal{D}$  with high

probability?



A 10  $\times$  10 picture





A 10  $\times$  10 picture





A 10  $\times$  10 picture





A 10  $\times$  10 picture





A 10  $\times$  10 picture





A 10  $\times$  10 picture





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Let R(n, k) be the event that  $P_n$  is reconstructible from its k-deck.

Narayanan-Y. '23+

There exists  $k_c(n)$  such that as  $n \to \infty$ ,

$$\operatorname{Prob}[R(n,k)] \to \begin{cases} 0 & \text{if } k < k_c(n) \\ 1 & \text{if } k > k_c(n) \end{cases}$$

Moreover,  $k_c(n)$  takes one of two values:  $\lfloor \sqrt{2 \log_2 n} \rfloor, \lceil \sqrt{2 \log_2 n} \rceil$ .

# Main Theorem

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Proof of the 0-Statement: If  $k < k_c(n)$ , then  $n^2 2^{-k^2} \to \infty$  as  $n \to \infty$ . Counting argument; bound the number of reconstructible pictures by the number of k-decks.

## Main Theorem

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Moreover,  $k_c(n)$  takes one of two values:  $\lfloor \sqrt{2 \log_2 n} \rfloor, \lceil \sqrt{2 \log_2 n} \rceil$ .

Proof of the 1-Statement: If  $k > k_c(n)$ , then  $n^2k2^{-k^2+k} \to 0$ . Our goal is to give an algorithm for reconstructing  $P_n$  from its deck and prove that the probability of failure tends to 0.

Step 0: Randomly order the deck  $\ensuremath{\mathcal{D}}$  and begin with the first deck element.



Step 1: Extend downward to 3*k* rows by placing the first deck element that fits.



Step 2: Extend to the right one column at a time, first at each of the corners



Step 2: Extend to the right one column at a time, first at each of the corners then internally. Repeat to the right and left until *n* columns.



Step 3: Extend upward one row at a time, then downward until *n* rows.



## Analysis: Naive Extensions



Observe that for each naive extension,

 $\operatorname{Prob}[\mathsf{mistake}] \le n^2 2^{-k^2 + k}$ 

So by union bound,

Prob[there is a mistake in the first step]  $\leq 3kn^22^{-k^2+k}$ 

which tends to 0 by our assumption. However, we cannot afford to do naive extensions for the entire grid. This is why we introduce the corner and internal extensions.

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## Analysis: Corner Extensions

Suppose we have correctly reconstructed *S* and are extending to the right. Before placing a corner subgrid *T*, we check to see if it can be extended to a  $(2k - 1) \times (2k - 1)$  subgrid *S'* using deck elements.



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A *k*-grid is *bad* if it is incorrect with respect to *P<sub>n</sub>*. We mark bad *k*-grids, e.g. in the upper-right corner. An interface path is a path separating the good and bad entries.

## Analysis: Interface Paths



We compute probabilities associated with the interface paths. For example,

$$\operatorname{Prob}[\operatorname{first step}] \le n^2 2^{-k^2 + k}$$

but

Prob[second step | first step] 
$$\leq n^2 2^{-k^2+1} + 2(4k^2)(2^{-k+1})$$



The technique of computing a first moment along a path/contour originated with Peierls in a proof of phase coexistence for the Ising model on  $\mathbb{Z}^d$  and is often used in percolation.



Images from Friedli-Velenik, Statistical Mechanics of Lattice Systems and Grimmett, Percolation

- Demidovich–Panichkin–Zhukovskii use a variation of our techniques to give 2-point concentration for dimensions  $d \ge 2$  and colors  $r \ge 2$
- Sharp threshold?
- DPZ also connects their results to reconstruction of uniform *r*-colorings of G(n, 1/2) from *k*-decks (neighborhoods of radius *k*), but there is a gap from  $\sqrt{\log_2(n)}$  to  $\log_2 n$ .
- More variants: non-square, *p*-biased, noisy, correlated...

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# Thank you!