## Reconstructing Random Pictures

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AMS Central Sectional
October 7-8, 2023

## Background

## 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)

## Macsel-Pacc 18

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

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## Random Pictures

Today: Let $P_{n}$ be a random picture, i.e. an $n \times n$ grid with $\{0,1\}$ entries chosen uniformly at random. Let $\mathcal{D}$ 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?

Example


Deck of $9 \times 9$ subgrids

Example


Deck of $9 \times 9$ subgrids

Example


Deck of $9 \times 9$ subgrids

Example


Deck of $9 \times 9$ subgrids

Example


Deck of $9 \times 9$ subgrids


Deck of $4 \times 4$ subgrids


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## Main Theorem

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 $n \rightarrow \infty$,

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

Moreover, $k_{c}(n)$ takes one of two values: $\left\lfloor\sqrt{2 \log _{2} n}\right\rfloor,\left\lceil\sqrt{2 \log _{2} n}\right\rceil$.

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Moreover, $k_{c}(n)$ takes one of two values: $\left\lfloor\sqrt{2 \log _{2} n}\right\rfloor,\left\lceil\sqrt{2 \log _{2} n}\right\rceil$.
Proof of the 0-Statement: If $k<k_{c}(n)$, then $n^{2} 2^{-k^{2}} \rightarrow \infty$ as $n \rightarrow \infty$.
Counting argument; bound the number of reconstructible pictures by the number of $k$-decks.

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

## Reconstruction Algorithm

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


## Reconstruction Algorithm

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


## Reconstruction Algorithm

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


## Reconstruction Algorithm

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.


## Reconstruction Algorithm

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



Observe that for each naive extension,

$$
\operatorname{Prob}[\text { mistake }] \leq n^{2} 2^{-k^{2}+k}
$$

So by union bound,

which tends to 0 by our assumption. However, we cannot afford to do naive exiensions ror the entiregrid. This is why we iniroduce the

## Analysis: Naive Extensions



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$$

So by union bound,
Prob[there is a mistake in the first step] $\leq 3 k n^{2} 2^{-k^{2}+k}$
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.

## 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 $(2 k-1) \times(2 k-1)$ subgrid $S^{\prime}$ using deck elements.


## Analysis: Corner Extensions

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A $k$-grid is bad if it is incorrect with respect to $P_{n}$. 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,

$$
\operatorname{Prob}[f i r s t \text { step }] \leq n^{2} 2^{-k^{2}+k}
$$

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

## Digression

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

## Further Directions

- Demidovich-Panichkin-Zhukovskii use a variation of our techniques to give 2-point concentration for dimensions $d \geq 2$ and colors $r \geq 2$

Sharp threshold?
DPZ also connects their results to reconstruction of uniform $r$-colorings of $G(n .1 / 2)$ from $k$-decks (neighhorhoods 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!

