## Shotgun Assembly of Labelled Graphs

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${ }^{1}$ Shotgun assembly of Labelled Graphs (arxiv.org/abs/1504.07682)
${ }^{2}$ Shotgun Assembly of Random Regular Graphs, (arxiv.org/abs/1512.08473)
${ }^{3}$ Shotgun Assembly of Random Jigsaw Puzzles, in progress.
Simons Conference on Random Graph Processes

## Graph Shotgun Problem

- Can one reconstruct a graph from collection of subgraphs?
- Reconstruction Conjecture (Kelley, Harary 50s): Any two graphs on 3 or more vertices that have the same multi-set of vertex-deleted subgraphs are isomorphic.


Figure: From Topology and Combinatorics Blog by Max F. Pitz

## Graph Shotgun Problem

- Can one reconstruct a graph from collection of subgraphs?
- Reconstruction Conjecture (Kelley, Harary 50s): Any two graphs on 3 or more vertices that have the same multi-set of vertex-deleted subgraphs are isomorphic.
- Mossel-Ross-15: What if Graphs are Random or have random labels? (easier)
- And given only local neighborhoods of each vertex (harder)?


## DNA Shotgun Sequencing



Figure: From "Whole genome shotgun sequencing versus Hierarchical shotgun sequencing" by Commins, Toft, and Fares (2009).

## Q1: Deterministic

- Sequence of letters (A, C, G, T or other) of length $N$.
- All "reads" of length $r$ are given.

Example: $N=14, r=3$ :
AT GGGCACTGAGCC

Reads:

$$
\begin{aligned}
& \{A T G, T G G, G G G, G G C, G C A, C A C \\
& \quad A C T, C T G, T G A, G A G, A G C, G C C\}
\end{aligned}
$$

Combinatorial Question:
When does this multi-set uniquely determine the sequence?

## Q1: Deterministic

Ans (Ukkonen-Pevzner):
Identifiability is possible if and only if none of the following blocking patterns appear:

- Rotation:

$$
x \alpha y \beta x \Longleftrightarrow y \beta x \alpha y
$$

- Triple repeat:

- Interleaved repeat:

[ $x, y$ are $(r-1)$-tuples and $\alpha, \beta$ are non-equal strings]


## Q1: Deterministic

Proof is based on creating a de Bruijn graph:


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Identifiability is possible if and only if a unique Eulerian path (though not circuit).

## Setup Q2: Randomized

Random sequence, entries independent and uniform on $q$ letters.

- What is the probability of identifiability?
- Criteria on growth of $r=r_{N}$ as $N \rightarrow \infty$ such that the chance sequence is identifiable tends to zero or one?

Ukkonen-Pevzner useful - understand the probability of the appearance of the blocking patterns.

- If $r / \log (N)>2 / \log (q)$ eventually, then probability of identifiability tends to one.
- If $r / \log (N)<2 / \log (q)$ eventually, then probability of identifiability tends to zero.
- Dyer-Frieze-Suen-94,....
- Still active area of research: e.g.: reads with errors, e.g: Ganguly-M-Racz-16.


## What about other Graphs??

## Graph Shotgun Sequencing

Paninski et al. (2013) : How to reconstruct neural network from subnetworks?


Figure: wiki commons

## Random Puzzle Problem



Figure: wiki commons

Math Question: For an $n \times n$ puzzle with $q$ types of random jigs, how large should $q(n)$ be so that the puzzle can be assembled uniquely??

## A general setup

(1) $\mathcal{G}$ is a (fixed or random) graph,
(2) Possibly with random labeling of the vertices,
(3) For each vertex $v$, given a rooted neighborhood $\mathcal{N}_{r}(v)$ of "radius" $r$.


## Random jigsaw Puzzle

- Puzzle $=[n] \times[n]$ grid with uniform $q$-coloring of the edges of the grid.
- Piece $=$ vertex along with 4 adjacent colored half edges.
- Given: $n^{2}$ pieces.
- Goal: Recover the puzzle.
- Assume pieces at the edges also have 4 colors (harder).
- For presentation purposes: colored edges vs.
- Real Puzzle: colored half edges and a compatibility involution.



## The unique Assembly Question

- A feasible assembly is a permutation of the pieces such that adjacent two half-edges have the same color.
- A puzzle has unique vertex assembly (UVA) if (up to rotations) it has only one feasible assembly.
- A puzzle has unique edge assembly (UEA) if for every feasible assembly, every edge has the same color as in the planted solution (up to rotations).
- Question: How large should $q$ be to ensure unique edge/vertex assembly with high probability $(\rightarrow 1$ as $n \rightarrow \infty)$ ?


## Bounds on puzzle assembly

From M-Ross:

- $q \ll n \Longrightarrow P(U V A) \rightarrow 0$.


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> Theorem (Bordenave-Feige-M)
> For all $\varepsilon>0$, If $q \geq n^{1+\varepsilon}$ then $P(U V A) \rightarrow 1$.

- Open Problem 1: Zoom in on threshold?
- Open Problem 2: Threshold for UEA.


## Assembly algorithm

We use a simple assembly algorithm:

- A feasible $k$-neighborhood of piece $v$ is map $f$ from $[-k, k]^{2} \rightarrow$ pieces such that $f(0)=v$ and if $x \sim y \in[-k, k]^{2}$ then the corresponding half-edges in $f(x)$ and $f(y)$ have the same color.
- Algorithm: find all feasible $k$-neighborhoods for each vertex $v$.
- Declare piece $u$ to be a neighbor of $v$ if it is its neighbor of $v$ in each $k$-neighborhood.
- We take $k=O(1 / \varepsilon)$.
- How to analyze?


## Analysis 1

- Note: impossible to hope to recover k-neighborhood exactly, e.g - corners are often wrong.
- Fix $f:[-k, k]^{2} \rightarrow[n]^{2}$ with $f(0)=v$. What is the probability that $f$ is feasible?
- If $f(x)=v+x$ then probability 1 .
- If $f$ is random then probability $q^{-8 k^{2}(1+o(1))}$.


## Analysis 2

- Define a tile of $f$ to be a connected component of $f\left([-k, k]^{2}\right)$.
- Let $v \in T_{0}, T_{1}, \ldots, T_{r}$ be the tiles of $f$.


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P[f \text { feasible }]=q^{-\gamma}, \quad \gamma=\frac{1}{2}\left(\sum\left|\partial T_{i}\right|-8 k\right)
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- Isoperimetric lemma: If $f$ separates $v$ from its neighbors then:

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n^{2} n^{2 r} q^{-\gamma}=n^{2} n^{2 r} n^{-\gamma(1+\varepsilon)} \ll 1
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- E.g: many small tiles - each contributed at least 2 to $\gamma$.


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- Isoperimetric lemma + union bound $\Longrightarrow$ proof.


## Cheat and Punishment

Sadly boundary events are not independent.

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- Graph theoretic definition of $\gamma(f)$, the number of "unique constraints".
- Isoperimetric lemma to lower bound $\gamma(f)$.
- Interesting: lower bound uses both $\sum\left|\partial T_{i}\right|$ and $\sum\left|\partial f\left(T_{i}\right)\right|$


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- Easy direction: "name" vertex $v$ by $B_{k}(v)$.
- Other direction requires more work per-example.


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Blocking configuration for $r$-neighborhoods (line graph has is of length $r+1$ )


Since has same r-neighborhoods as


- if $r<\log N[\lambda-\log (\lambda)]^{-1}$, then the probability of identifiability tends to zero.


## Example 1a: Sparse Erdős-Rényi random graph

## Diameter

- For $\lambda \neq 1$, the diameter of the sparse Erdős-Rényi random graph is of order $\log (N)$ (different constants than that above).
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- Harder/Open: $r=C \log N(1+o(1))$ ?
- Critical case?


## Example 1b: Less sparse Erdős-Rényi random graph

Structure of the Erdős-Rényi graph depends on behavior of $N \times p_{N}$.
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- multiset of degrees of neighbors of each vertex become unique.
- Allows to give distinct names to vertices.
- Open: when is $r=2$ enough?
- Distributed computing perspective: unique i.d's from local information.


## Example 2: Random Regular Graphs

Theorem (M+Sun)
The threshold for assembly of random d regular graphs is

$$
r=\frac{\log n+\log \log n}{2 \log (d-1)}+\Theta(1)
$$

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- (Almost) all $0.5 \log _{d-1}(n)$ neighborhoods are happy trees.
- Each $0.5(1+\epsilon) \log _{d-1}(n)$ neighborhoods is unhappy due a unique cycle structure.


## The Upper Bound

## Theorem (Bollobas 82)

For all $\varepsilon>0$ if $r \geq(0.5+\varepsilon) \log _{d-1} n$ then for all $u \neq v$ it holds that $\left(d_{1}(v), \ldots, d_{r}(v)\right) \neq\left(d_{1}(u), \ldots, d_{r}(u)\right)$ where $d_{i}(v)$ are the number of nodes at distance $i$ from $v$.

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Main ideas:

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- Fix No. 2: Define a metric on cycle structures and study corresponding measure metric space.


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Find the following:


Figure: Two neighborhoods that are hard to distinguish

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