### Random Lifts of Graphs

#### Nati Linial

#### 27th Brazilian Math Colloquium, July '09

Nati Linial Random Lifts of Graphs

- A brief introduction to the probabilistic method.
- A quick review of expander graphs and their spectrum.
- ► Lifts, random lifts and their properties.
- Spectra of random lifts.

# What is the probabilistic method? Introduction by example.

#### Theorem

In every party of 6 people there are either 3 who know each other or 3 who are strangers to each other.

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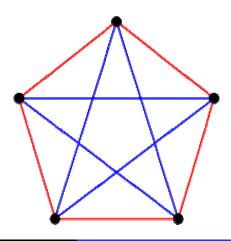
In every party of 6 people there are either 3 who know each other or 3 who are strangers to each other.

In other words: If you color the edges of  $K_6$  (the complete graph on 6 vertices) blue and red, you necessarily find a monochromatic (either red or blue triangle.

The result is tight. There is a red-blue coloring of the edges of  $K_5$  with no monochromatic triangle.

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More generally, **Theorem (Ramsey; Erdős-Szekeres)** Let  $N = \binom{r+s-2}{r-1}$ . If you color the edges of  $K_N$  red and blue, then you necessarily get a red  $K_r$  or a blue  $K_s$ .

### **Diagonal Ramsey Numbers**

Theorem

In every red-blue coloring of the edges of  $K_N$  there is a monochromatic complete subgraph on at least

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Theorem

Every N-vertex graph contains either a clique or an anti-clique on at least

$$\frac{1}{2}\log_2 N$$
 vertices.

### Diagonal Ramsey Numbers (contd.)

#### Theorem (Erdős '49)

There are red-blue colorings of  $K_N$  where no monochromatic subgraph has more than

 $2\log_2 N$  vertices.

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- We do not know how to do this. In fact it is a major challenge to find such explicit colorings.
- Instead, we use the probabilistic method.

There are  $2^{\binom{N}{2}}$  ways to color the edges of  $K_N$  by red and blue. We think of them as a probability space  $\Omega$  with the uniform distribution.

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In other words, we give the following recipe for sampling from  $\Omega$ : For each edge of  $K_N$ , flip a coin (independently from the rest). If it comes out heads color the edge red if you get tails, color it blue.

Let us consider an integer r (to be determined later) and a random variable B defined on  $\Omega$ . for a given coloring C of  $K_N$ , we define B(C) to be the number of sets of r vertices in C all of whose edges are blue. The expectation of X is:

$$\binom{N}{r} \frac{1}{2^{\binom{r}{2}}}.$$

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$$\binom{N}{r} \frac{1}{2\binom{r}{2}}$$

We likewise define a random variable R that counts red subgraphs.

Note that if  $B(C) = \mathcal{R}(C) = 0$ , there is no monochromatic set of *r* vertices in C, which is just what we need.

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$$r=2\log_2 N.$$

### What has happened here?

Why does the probabilistic method work so well?

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In every mathematical field intuition is created from examples that we know.

But it is hard to analyze large specific examples and the probabilistic method allows us to bypass this difficulty. It serves us as an observational tool, much like the astronomer's telescope or a biologist's microscope. What we saw is a close relative of the most basic model of random graphs, Erdős-Rényi's G(n, p)model. In this model we start with *n* vertices. For each pair of vertices x, y we decide, independently and with probability *p*, to put an edge between *x* and *y*. What we saw is a close relative of the most basic model of random graphs, Erdős-Rényi's G(n, p)model. In this model we start with *n* vertices. For each pair of vertices x, y we decide, independently and with probability *p*, to put an edge between *x* and *y*.

There are other important and interesting models of graphs. For example, we have known for 30 years now how to sample random *d*-regular graphs.

### What's there, what's still needed?

Random graphs come up a lot in (mathematical) statistical mechanics. A major example is Percolation Theory.

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A typical basic question in this area: What is the probability that an infinite connected component remains.

You could also seek models to describe natural or artificial phenomena such as the Internet graph or biological control networks. In particular, one shortcoming of the G(n, p) model is the lack of control we have over the graph's structure. In particular, one shortcoming of the G(n, p) model is the lack of control we have over the graph's structure.

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Random lifts of graphs, our subject today, are such a model.

There are three main perspectives of expansion:

- Combinatorial isoperimetric inequalities
- Linear Algebraic spectral gap
- Probabilistic Rapid convergence of the random walk (which we do not discuss today)

For (much) more on this: see our survey article with Hoory and Wigderson.

A graph G = (V, E) is said to be  $\epsilon$ -edge-expanding if for every partition of the vertex set V into X and  $X^c = V \setminus X$ , where X contains at most a half of the vertices, the number of cross edges

$$e(X, X^c) \geq \epsilon |X|.$$

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In words: in every cut in G, the number of cut edges is at least proportionate to the size of the smaller side.

## The combinatorial definition (contd.)

#### The edge expansion ratio of a graph G = (V, E), is

$$h(G) = \min_{S \subseteq V, |S| \le |V|/2} \frac{|E(S,\overline{S})|}{|S|}.$$

The **Adjacency Matrix** of an *n*-vertex graph *G*, denoted A = A(G), is an  $n \times n$  matrix whose (u, v)entry is the number of edges in *G* between vertex *u* and vertex *v*. Being real and symmetric, the matrix *A* has *n* real eigenvalues which we denote by  $\lambda_1 \ge \lambda_2 \ge \cdots \ge \lambda_n$ .

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 A substantial spectral gap implies logarithmic diameter.

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# **Theorem** Let G be a d-regular graph with spectrum $\lambda_1 \ge \cdots \ge \lambda_n$ . Then

$$\frac{d-\lambda_2}{2} \leq h(G) \leq \sqrt{(d+\lambda_2)(d-\lambda_2)}.$$

The bounds are tight.

## What's a "large" spectral gap?

If expansion is "good" and if a large spectral gap yields large expansion, then it's natural to ask:

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Theorem (Alon, Boppana)

$$\lambda_2 \geq 2\sqrt{d-1} - o(1)$$

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What, then, is the ideal expander? A good candidate is the infinite *d*-regular tree. Using (a little) spectral theory it is possible to define a spectrum for infinite graphs. It turns out that the spectrum of the *d*-regular infinite tree spans the interval

$$(-2\sqrt{d-1},2\sqrt{d-1})$$

How tight is this bound?

# Problem

Are there d-regular graphs with second eigenvalue

$$\lambda_2 \leq 2\sqrt{d-1}$$
 ?

When such graphs exist, they are called Ramanujan Graphs.

What is the typical behavior?

**Problem** How likely is a (large) random d-regular graph to be Ramanujan?

# What is currently known about Ramanujan Graphs?

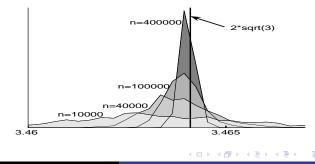
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Friedman: If you are willing to settle for  $\lambda_2 \leq 2\sqrt{d-1} + \epsilon$ , they exist. Moreover, almost every *d*-regular graph satisfies this condition.

### The distribution of the second eigenvalue



# Some open problems on Ramanujan Graphs

• Are there arbitrarily large *d*-regular Ramanujan Graphs (i.e.  $\lambda_2 \leq 2\sqrt{d-1}$ ) for every  $d \geq 3$ ? The first unknown case is d = 7.

# Some open problems on Ramanujan Graphs

- Are there arbitrarily large *d*-regular Ramanujan Graphs (i.e. λ<sub>2</sub> ≤ 2√d − 1) for every *d* ≥ 3? The first unknown case is *d* = 7.
- Can we find combinatorial/probabilistic methods to construct graphs with large spectral gap (or even Ramanujan)? As we'll see random lifts of graphs (Bilu-L.) yield graphs with

 $\lambda_2 \leq O(\sqrt{d}\log^{3/2} d).$ 

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Definition A map  $\varphi : V(H) \rightarrow V(G)$  where G, H are graphs is a covering map if for every  $x \in V(H)$ , the neighbor set  $\Gamma_H(x)$  is mapped 1 : 1 onto  $\Gamma_G(\varphi(x))$ .

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This is a special case of a fundamental concept from topology. Recall that a graph is a one-dimensional simplicial complex, so covering maps can be defined and studied for graphs. When there is a covering map from H to G, we say that H is a lift of G.

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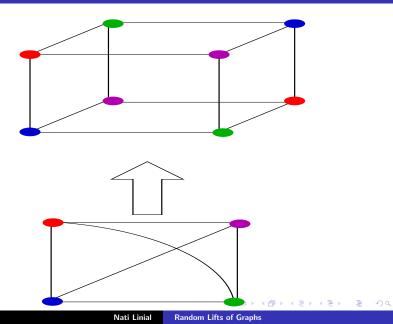
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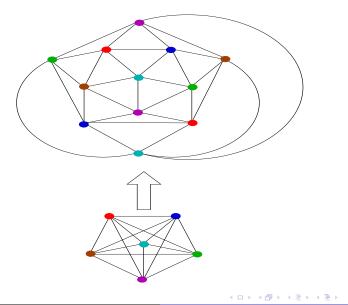
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Convention: We will always assume that the base graph is connected. This creates no loss in generality.

### An example - The 3-cube is a 2-lift of $K_4$



## The icosahedron is a 2-lift of $K_6$



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We see in the previous examples that the covering map  $\varphi$  is 2 : 1.

• The 3-cube is a 2-lift of  $K_4$ .

▶ The graph of the icosahedron is a 2-lift of  $K_6$ . In general, if G is a connected graph, then every covering map  $\varphi : V(H) \rightarrow V(G)$  is n : 1 for some integer n (easy).

#### • We call *n* the fold number of $\varphi$ .

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- The set of those graphs that are *n*-lifts of G is denoted by L<sub>n</sub>(G).

### A direct, constructive perspective

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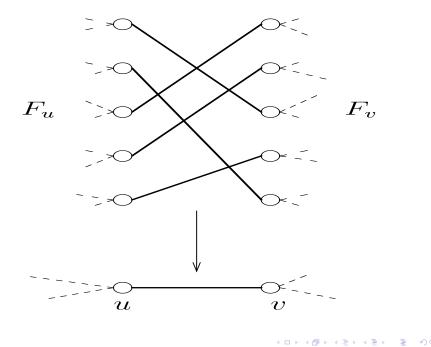
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  - To construct graphs with certain desirable properties. In our case, to achieve large spectral gaps.
  - To model various phenomena.
  - To study their typical properties.

# A few more general properties of lifts

Vertex degrees are maintained. If x has d neighbors, then so do all the vertices in the fiber of x. In particular, a lift of a d-regular graph is d-regular.

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- Vertex degrees are maintained. If x has d neighbors, then so do all the vertices in the fiber of x. In particular, a lift of a d-regular graph is d-regular.
- The cycle  $C_n$  is a lift of  $C_m$  iff m|n.
- The *d*-regular tree covers every *d*-regular graph. This is the *universal cover* of a *d*-regular graph. Every connected base graph has a universal cover which is an infinite tree.

Here is an easy observation:

The lifted graph inherits every eigenvalue of the base graph.

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## Old vs. New Eigenvalues (contd.)

These are called the old eigenvalues of H. If G is given, the old eigenvalues appear in every lift, and we can only hope to control the values of the new eigenvalues.

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This suggests the following approach to the construction of d-regular Ramanujan Graphs by repeated lifts:

- Start from a small *d*-regular Ramanujan Graph (e.g. K<sub>d+1</sub>).
- ▶ In every step apply a lift to the previous graph while keeping all new eigenvalues in the interval  $[-2\sqrt{d-1}, 2\sqrt{d-1}]$

## How to think about 2-lifts

In an *n*-lift of a graph *G* we associate with every edge e = xy of *G* a permutation  $\pi_e \in S_n$  which tells us how to connect the *n* vertices in the fiber  $F_x$  with the *n* vertices of  $F_y$ .

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But in  $S_2$  the only permutations are the identity **id** and the switch  $\sigma = (12)$ . So a 2-lift of G is specified by deciding, for every edge e whether  $\pi_e$ equals **id** or  $\sigma$ . In an *n*-lift of a graph *G* we associate with every edge e = xy of *G* a permutation  $\pi_e \in S_n$  which tells us how to connect the *n* vertices in the fiber  $F_x$  with the *n* vertices of  $F_y$ .

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Alternatively, we sign the edges of G where +1 stands for **id** and -1 for  $\sigma$ .

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### Proposition

The new eigenvalues of a 2-lift of G are the eigenvalues of the corresponding signing matrix.

Recall: The spectral radius of a matrix is the largest absolute value of an eigenvalue.

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### Conjecture

Every d-regular Ramanujan Graph has a signing with spectral radius  $\leq 2\sqrt{d-1}$ .

#### But it seems that something much stronger is true

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This conjecture, if true, is tight.

## Theorem (Yonatan Bilu + L.)

By repeated application of 2-lifts it is possible to explicitly construct d-regular graphs ( $d \ge 3$ ) whose second eigenvalue

$$\lambda_2 \leq O(\sqrt{d} \log^{3/2} d)$$

The most unexpected part of the proof is a converse of the so-called Expander Mixing Lemma.

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What's involved is the graph's discrepancy, i.e. the maximum of

$$\frac{e(A,B) - \frac{d}{n}|A||B|}{\sqrt{|A||B|}}$$

A matching M in a graph G is a collection of disjoint edges. If the edges in M meet every vertex in G, we say that M is a perfect matching=PM. The defect of G is the number of vertices missed by the largest matching in G. (So the existence of a PM is the same as zero defect).

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Question: Given a base graph G and a large even integer n, how likely is an n-lift of G to contain a perfect matching?

Let G be a base graph and let H be a random 2n lift of G. For every base graph exactly one of the following four situations occurs:

• Every 2n-lift H of G contains a PM.

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- Every H must have defect ≥ αn for some constant α > 0.

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- The probability that H has a PM is 1 o(1) (but not 1).
- Almost surely H has defect  $\Theta(\log n)$ .

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► The typical degree of connectivity of a lift of *G*.

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- ► The typical degree of connectivity of a lift of *G*.
- The chromatic numbers of typical lifts.
- The typical distribution of new eigenvalues.
- ... and more ....

## ... and much more that we do not know...

**Open Problem** 

▶ Is there a zero-one law for Hamiltonian cycles?

## **Open Problem**

- Is there a zero-one law for Hamiltonian cycles?
- What is the typical chromatic number of an n-lift of K<sub>5</sub>? Is it 3 or 4, perhaps each with positive probability?