

# The Zig-Zag Product and Reingold's Theorem

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# Overview of the Talk

- 1 Introduction
  - Reingold's Theorem:  $\text{ustcon} \in L$
- 2 Expander Graphs
  - Diameter of an Expander Graph
- 3 Initial Ideas
  - Graph Exponentiation
  - Tensor Product
- 4 The Zig-Zag Product
- 5 Reingold's Algorithm

# References



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Recall the definition of `ustcon` and  $L$ :

## Definition (Undirected $st$ -connectivity)

`ustcon` is the following decision problem:

Input: an undirected graph  $G = (V, E)$ , and two vertices  $s, t \in V$

Output: 1 if  $s \rightsquigarrow t$  in  $G$ , and 0 otherwise.

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## Definition (LogSpace)

$$L = DSPACE(\log n)$$

In other words,  $L$  contains all languages  $L \subseteq \{0, 1\}^*$  for which a Turing machine  $M$  decides  $L$  using no more than  $\mathcal{O}(\log n)$  space on input of length  $n$ .

# Reingold's Theorem

The goal of this talk is to describe the proof of Reingold's theorem:

Theorem (Reingold, 2005)

$$\text{ustcon} \in L$$

Note that the graph is not assumed to be simple (multiedges and loops are both allowed). We will restrict ourselves largely to regular graphs, and then show how the problem in a general graph can be reduced to the regular case.

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Loosely speaking, under the right assumptions on the graphs, all of the above notions are equivalent.

# Definition of Expander Graphs

We adopt the spectral point of view, restricted to regular graphs:

## Definition $((n, d, \lambda)$ -graph)

We say that  $G$  is an  $(n, d, \lambda)$ -graph if  $G$  is a  $d$ -regular graph on  $n$  vertices such that

$$\lambda \geq \max_{i \neq 1} \frac{|\lambda_i|}{d}$$

where  $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n$  is the spectrum of the adjacency matrix  $A$  of  $G$ .

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Note here that  $\lambda_1 = d$ , and that the maximum on the right is 1 if and only if  $G$  is either disconnected or bipartite.

## Definition (Expander Graphs)

A family of  $d$ -regular graphs  $\{G_j\}_j$  is called an expander family with spectral gap  $1 - \lambda$  (or a  $\lambda$ -expander family), if every  $G_j$  is an  $(n, d, \lambda)$ -graph for some  $n = n_j$ .



# The Punchline: Expanders have Logarithmic Diameter

The usefulness of expanders in solving `ustcon` comes from the following lemma about their diameters:

## Lemma (Diameter of an Expander)

Let  $G = (V, E)$  be a connected  $(n, d, \lambda)$ -graph. Then,

$$\text{diam}(G) = \max_{u,v} d(u, v) = \mathcal{O}_\lambda(\log n) = \mathcal{O}\left(\frac{\log n}{1 - \lambda}\right)$$

where the implicit constant is effective, and efficiently computable.

# Solving `ustcon` for $\lambda$ -expanders

We will now describe a log-space algorithm for  $d$ -regular graphs for which every connected component is a  $\lambda$ -expander with  $\lambda < 1$ .

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- Each path  $s = v_0, \dots, v_k$  of length  $k \leq \Delta$  can be encoded as a string in  $\{1, \dots, d\}^{\leq \Delta}$  where the  $j$ th letter encodes which of the  $d$  vertices adjacent to  $v_{j-1}$  is the next step,  $v_j$ .

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- Keep a counter storing which of these paths we are on, initialized to all 1s. This takes  $\mathcal{O}(\Delta \log d) = \mathcal{O}_\lambda(\log d \log n) = \mathcal{O}_{d,\lambda}(\log n)$  space.

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- Trawl through this path to see if  $t$  appears. If it does, halt and accept. This takes  $\mathcal{O}(1)$  space.
- Increment the counter by 1, treating it as a  $d$ -ary integer. If the counter overflows to all 1s, then halt and reject.

# Proof of Logarithmic Diameter

## Proof.

Let  $M = A/d$  be the random walk matrix, and  $\vec{u} = \vec{v}_1 = (1/n, \dots, 1/n)^T$  be the uniform distribution on the vertices. Let  $\vec{p}$  be any probability distribution on the vertices, and let  $t$  be a time parameter.



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In particular, this means that  $M$  is a  $\lambda$ -contraction on  $\vec{u}^\perp$ . Note that  $(\vec{p} - \vec{u}) \cdot \vec{u} = 0$ , and hence if  $t \geq 1000 \left( \frac{\log n}{\log(1/\lambda)} \right) \sim \frac{\log n}{1-\lambda}$ ,

$$\|M^t \vec{p} - \vec{u}\|_2 = \|M^t(\vec{p} - \vec{u})\|_2 \leq \lambda^t \|\vec{p} - \vec{u}\|_2 \leq 2\lambda^t \leq \frac{2}{n^{1000}}$$

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But this can only happen if every vertex has positive probability at time  $t$ . Thus, every vertex is reachable in  $t$  steps;  $\text{diam}(G) \leq t$  as desired.

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Note  $s$  is connected to  $t$  in  $G^k$  if and only if it is connected in  $G$ , and it is immediate that if  $G$  is an  $(n, d, \lambda)$ -graph then  $G^k$  is an  $(n, d^k, \lambda^k)$ -graph. This vastly improves expansion – however, the degree blows up very quickly. Our algorithm (for constant degree expanders) does not apply!

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It is not hard to see that if  $G$  is an  $(n, d, \alpha)$ -graph and  $H$  is an  $(m, d', \beta)$ -graph, then  $G \otimes H$  is an  $(mn, dd', \lambda)$ -graph with  $\lambda = \max\{\alpha, \beta\}$  – this follows from the fact that the spectrum of a tensor product of matrices is the pointwise product of their spectra.

# The Replacement Product

This brings us to the novel construction known as the Zig-Zag product. For this, we assume that all graphs are on the vertex set  $\{1, \dots, n\}$  for some  $n$ .

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- All edges are either zigs or zags.

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- In other words, there is an edge between  $(u, i)$  and  $(v, j)$  in  $G \circledast H$ , whenever  $i$  is adjacent to  $k$  in  $H$  and  $j$  is adjacent to  $\ell$  in  $H$ , where the  $k$ th edge at  $u$  is the same as the  $\ell$ th edge at  $v$ .

# The Zig-Zag Product

We have the following theorem relating the expansion properties of  $G \circledast H$  with those of  $G$  and  $H$ :

**Theorem (Reingold-Vadhan-Wigderson [RVW02])**

*Let  $G$  be an  $(n, m, \alpha)$ -graph and  $H$  an  $(m, d, \beta)$ -graph. Then  $G \circledast H$  is an  $(nm, d^2, \varphi)$ -graph where  $\varphi = \varphi(\alpha, \beta)$  satisfies the following:*

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- $\alpha, \beta < 1$  implies  $\varphi < 1$ .

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For now, note that if  $\beta \leq 1/2$ , then  $(1 - \phi) \geq \frac{3}{8}(1 - \alpha)$ .

# Sketch of the Algorithm

We can now describe the algorithm. Suppose  $G$  is a  $d^{16}$ -regular graph ( $d$  will be fixed later) none of whose connected components are bipartite. Let  $H$  be a fixed  $(d^{16}, d, \beta)$ -graph with  $\beta \leq 1/2$ .

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- $s, t$  is connected in  $G_0$  only if  $s_j$  and  $t_j$  are connected in  $G_j$  where  $s_j$  and  $t_j$  are recursively chosen as any vertex in the cloud representing  $s_{j-1}$  and  $t_{j-1}$ . Check  $s_j, t_j$  connectivity instead.

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*Given two vertices  $u$  and  $v$  in  $G_j$ , it can be checked in  $\mathcal{O}_d(\log n + j)$  space whether  $u$  and  $v$  are adjacent in  $G_j$  without explicitly storing the graph  $G_j$ .*

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- Putting  $j \approx_d \log n$  in the above lemma tells us that the algorithm for connectivity in  $G_j$  can be executed with log-space overhead without ever constructing  $G_j$ .
- Finally, a graph  $(d^{16}, d, 1/2)$  can be found using the probabilistic method, for some fixed small  $d$ .

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- First, we reduce to the 3-regular case: for a general graph  $G$ , create  $G'$  by replacing a vertex  $v$  with a cycle of length of  $\deg(v)$ , and edges of  $G$  by matchings (that is, if the  $i$ th edge of  $u$  and the  $j$ th edge of  $v$  coincide then draw an edge between  $(u, i)$  and  $(v, j)$ ).

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We now return to the intuition behind Reingold-Vadhan-Wigderson.

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There are two basic intuitions, both of which can be converted into (somewhat technical) proofs:



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  - The general case is a superposition of the above two extremes.

Thank You!