

Improved Bounds for Online Preemptive Matching

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with Leah Epstein, Asaf Levin, and Oren Weimann

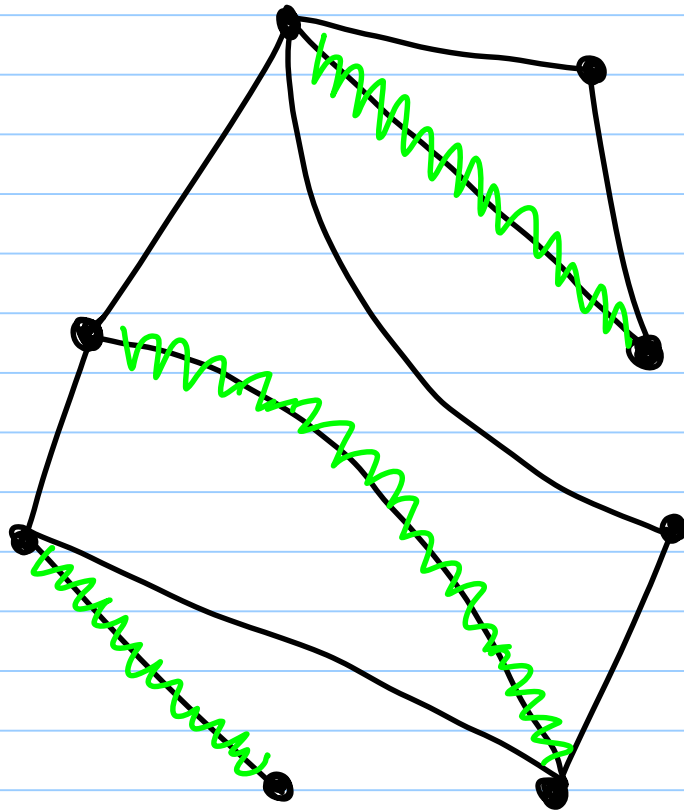
OUTLINE

- Introduction
- Lower bound for max-cardinality matching
- Additional results

INTRODUCTION

MAX-CARDINALITY MATCHING : OFFLINE VERSION

$G = (V, E)$

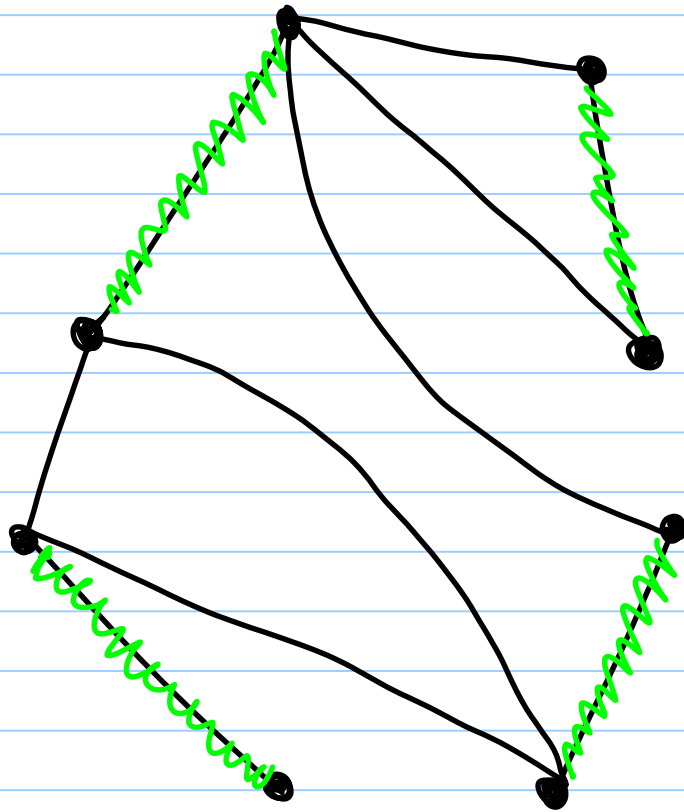


matching: collection of disjoint edges

Objective: Find a matching $M \subseteq E$ of maximum cardinality

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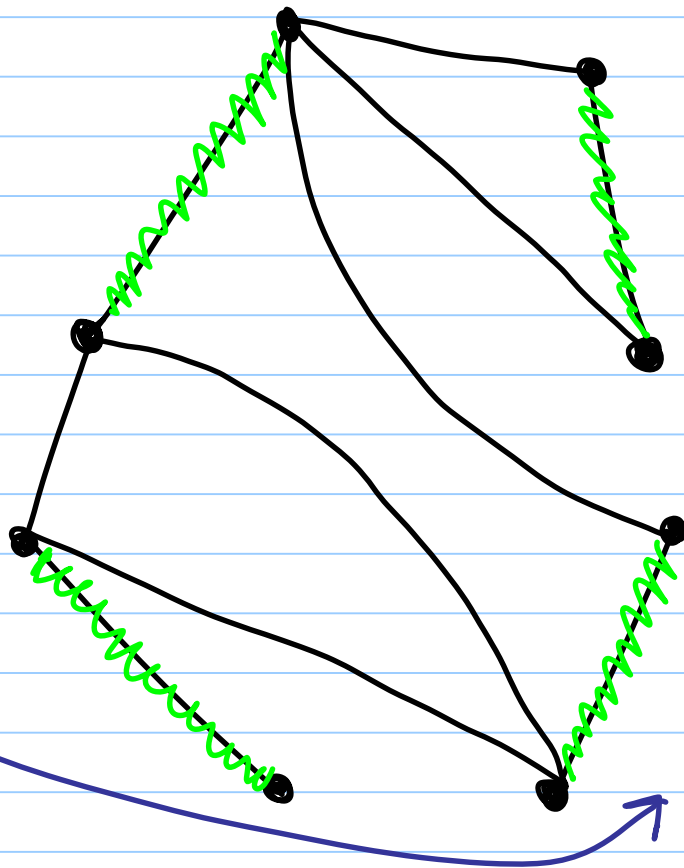


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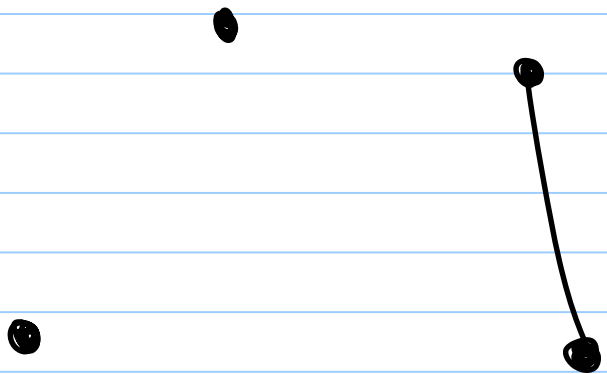
EFFICIENT ALGORITHMS
+ MASSIVE LITERATURE

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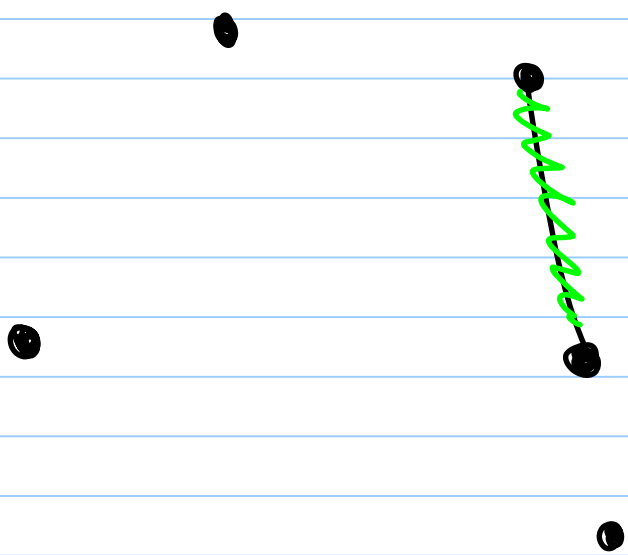
- Edges are revealed one by one
- Algorithm's decision : add the current edge or not ???
- Needs to keep a feasible matching at any time

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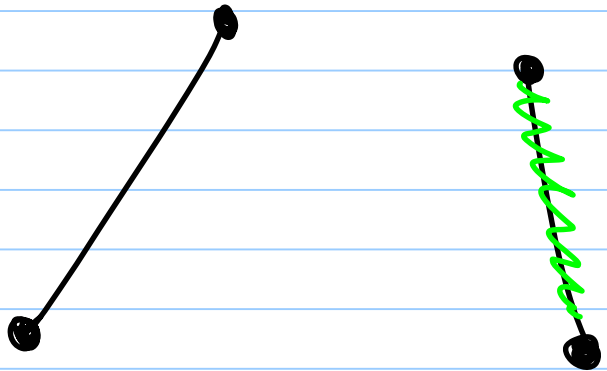
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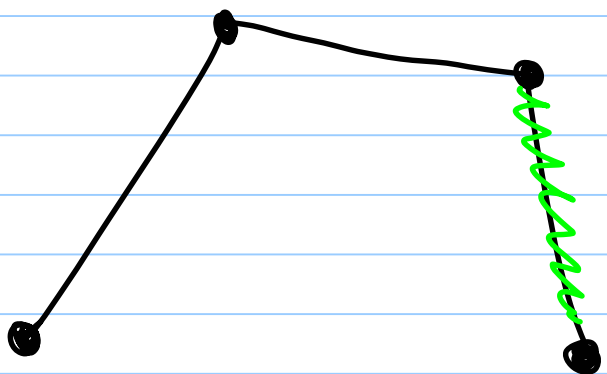
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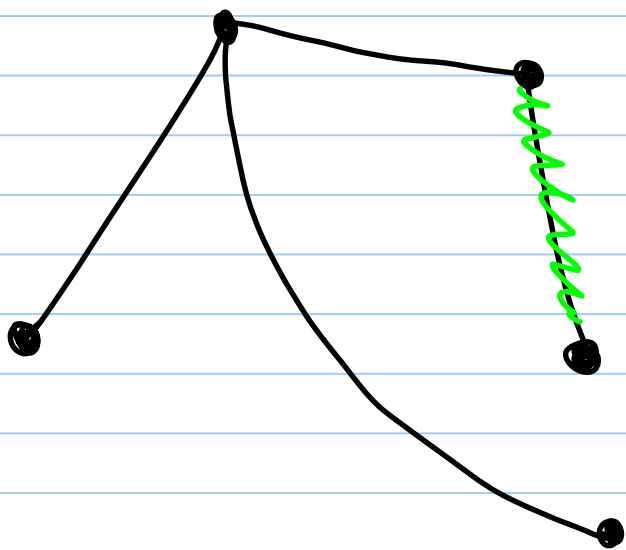
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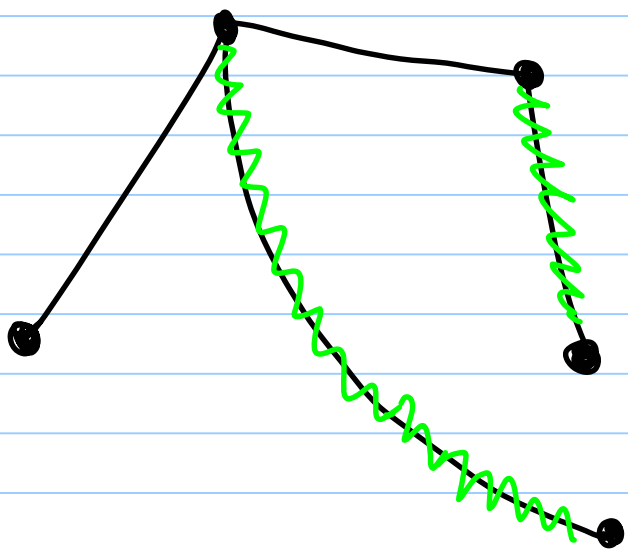
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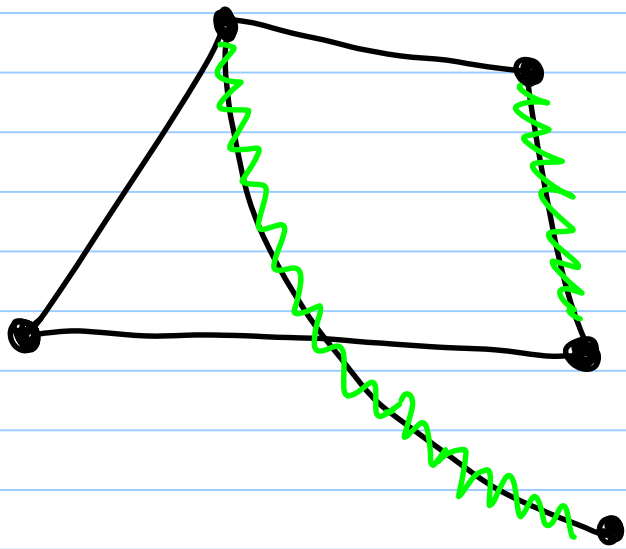
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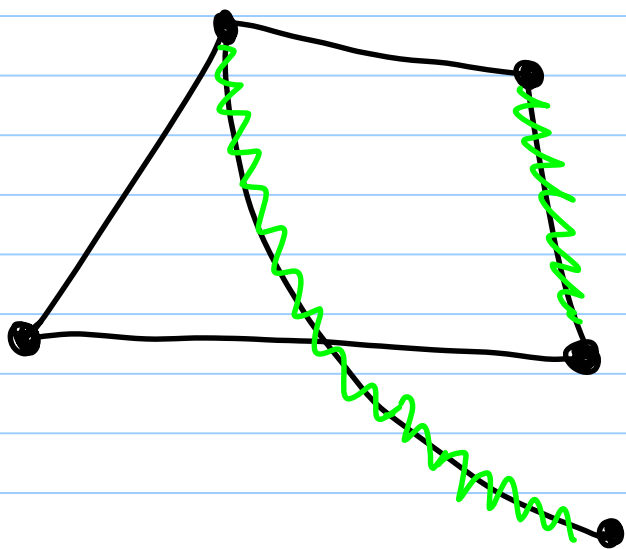
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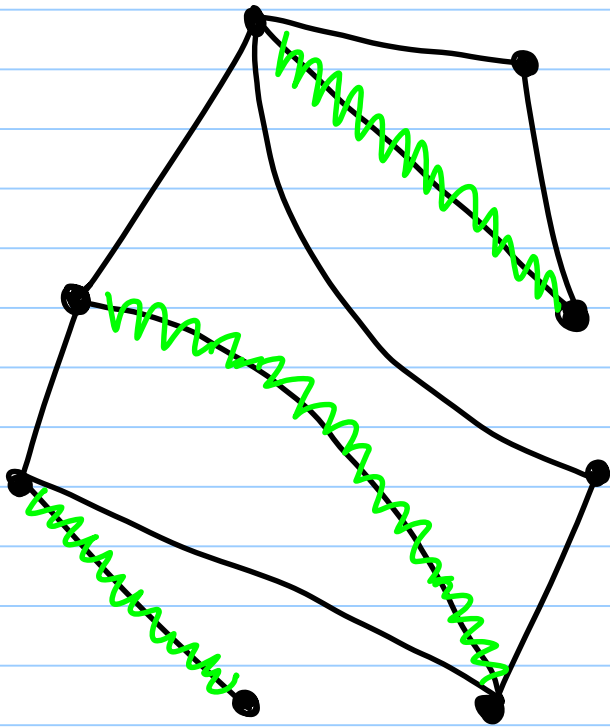


- Preemptive setting: can discard previously-picked edges

- Non-preemptive setting: picked edges are permanent

→ THIS TALK

COMPETITIVE RATIO



Worst possible performance
over all input sequences:

$$CR(A) = \sup_s \frac{\text{max-matching}(s)}{A\text{-matching}(s)}$$

PREEMPTIVE SETTING: PREVIOUS WORK

- **Deterministic** algorithms
 - greedy is **2-competitive**
 - this is **best-possible**
- **Randomized** algorithms
 - greedy has not been beaten yet
 - lower bound of **$e/(e-1) \approx 1.581$**
[Karp, Vazirani, and Vazirani '90]

PREEMPTIVE SETTING: PREVIOUS WORK

- **Deterministic** algorithms
 - greedy is **2-competitive**
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NEW RESULT: lower bound
of $1 + \ln 2 \approx 1.693$

- **Randomized** algorithms
 - greedy has not been beaten yet
 - lower bound of $e/(e-1) \approx 1.581$
[Karp, Vazirani, and Vazirani '90]

SKETCH OF THE
LOWER BOUND PROOF

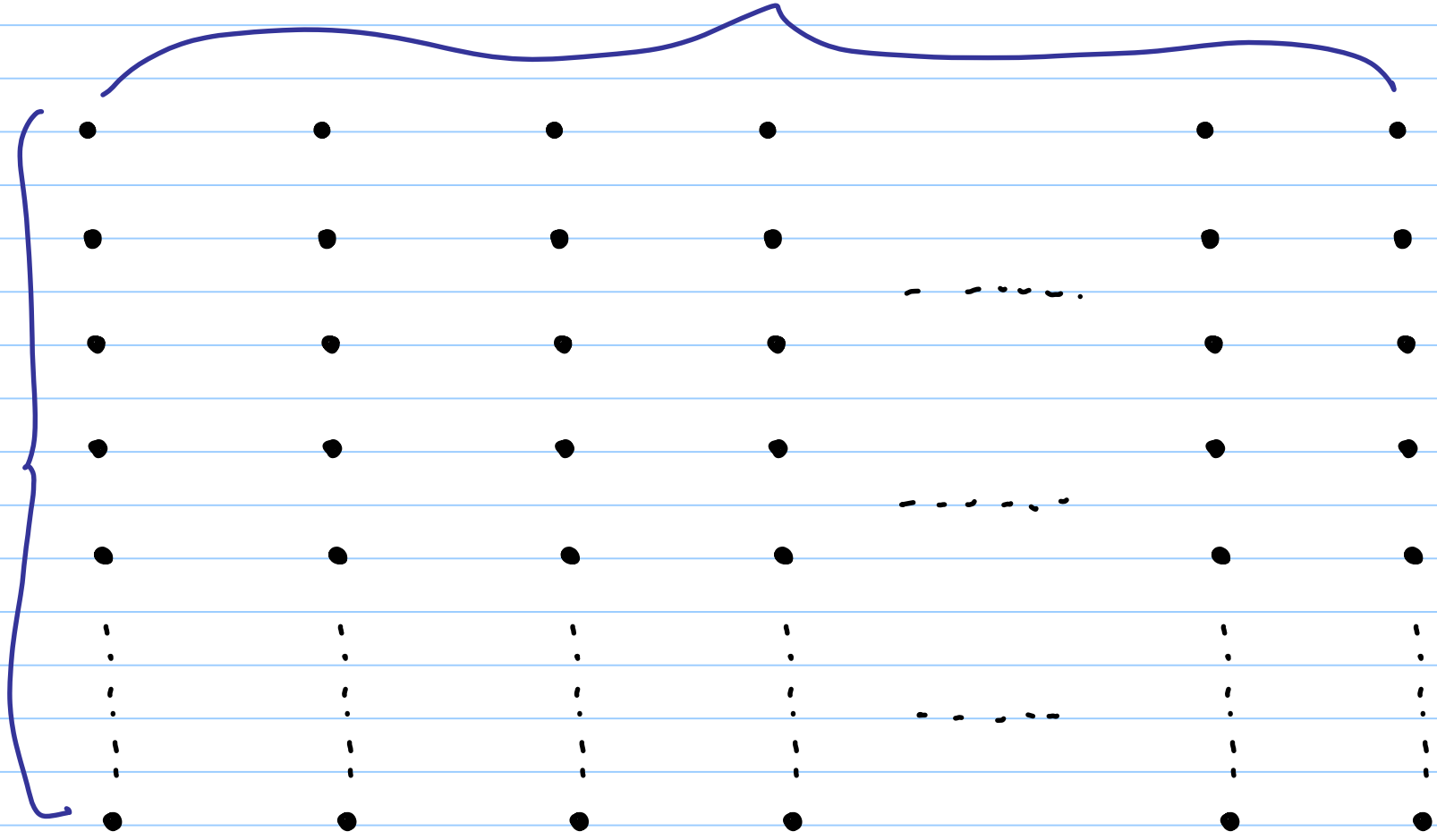
LB FOR RANDOMIZED ALGORITHMS?

- Online version of ^[1977] Yao's Lemma:
To obtain a lower bound for randomized algorithms, will evaluate the performance of any deterministic algorithm on some probability distribution of the inputs.
- Will assume that:
 - matching ALG is deterministic
 - sequence of revealed edges is random

GENERAL STRUCTURE

L

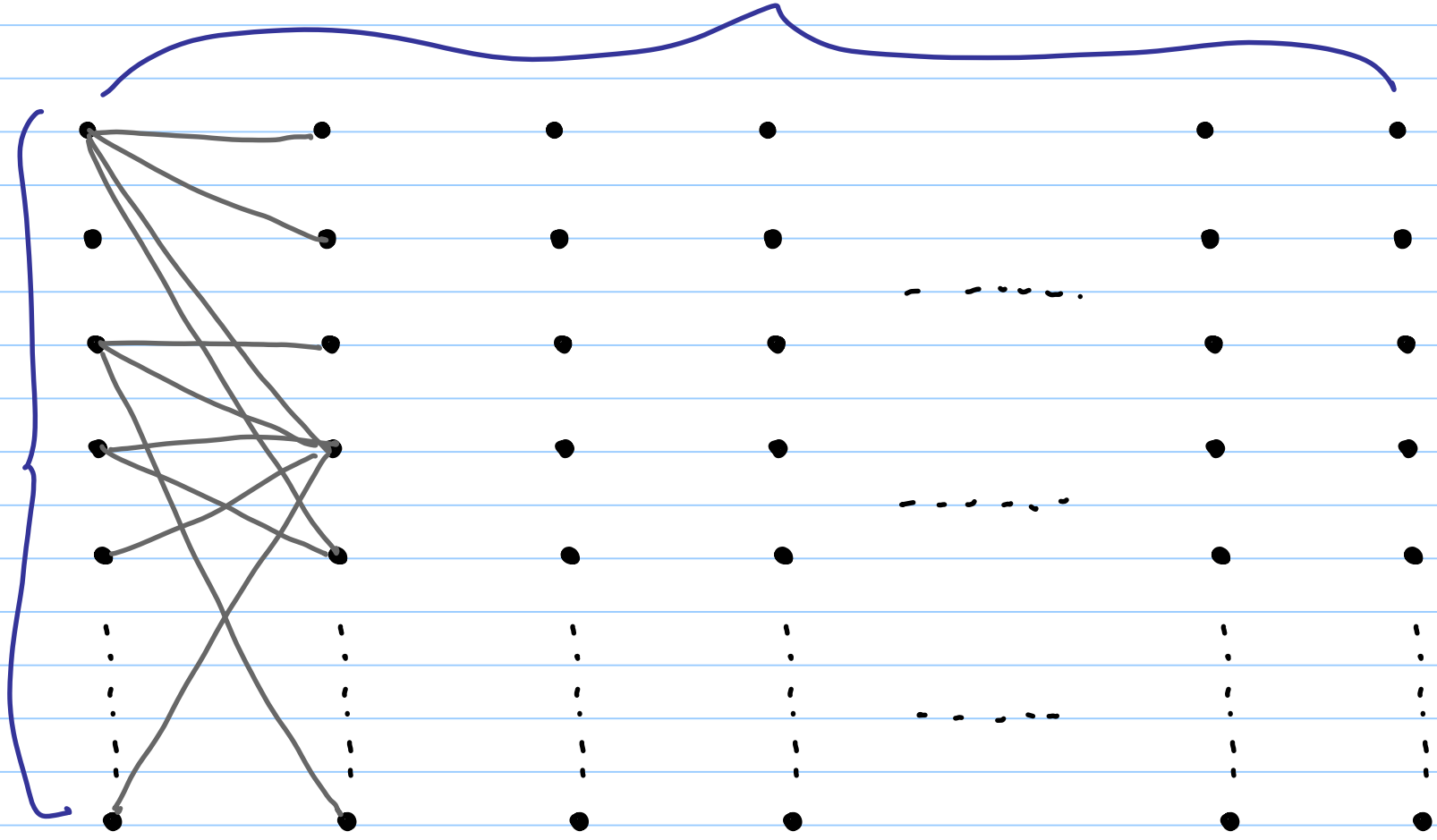
$2n$



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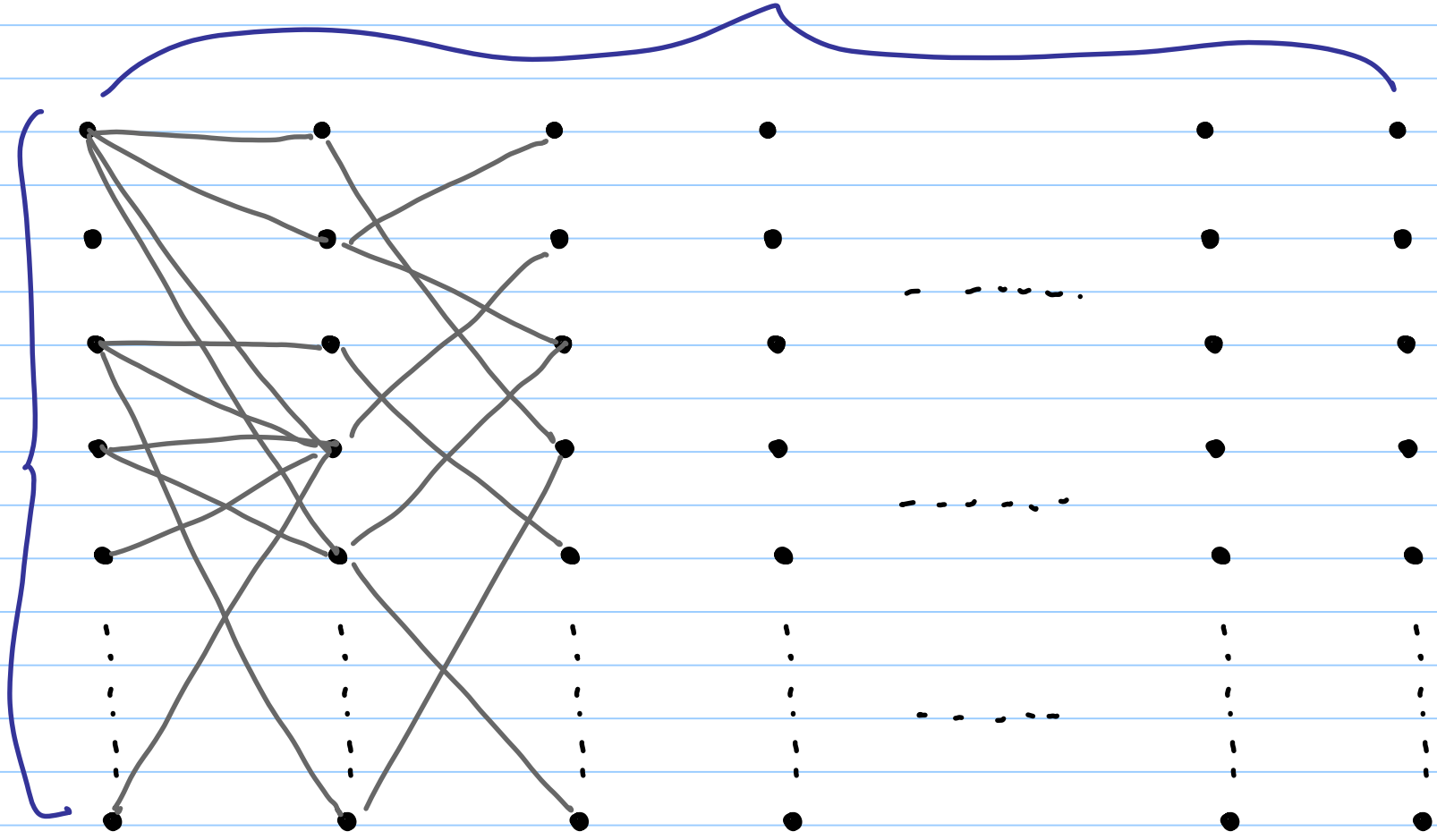
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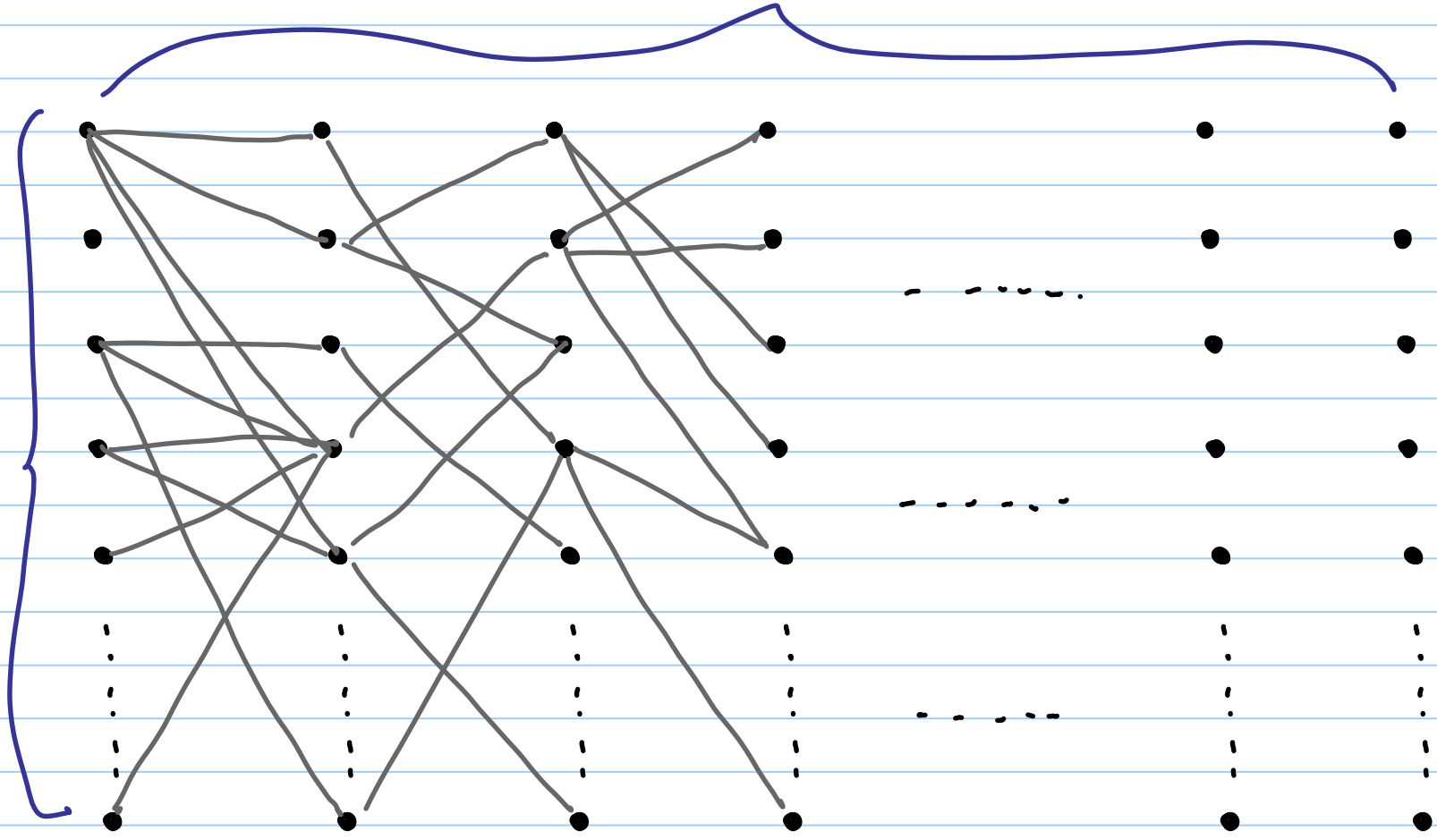
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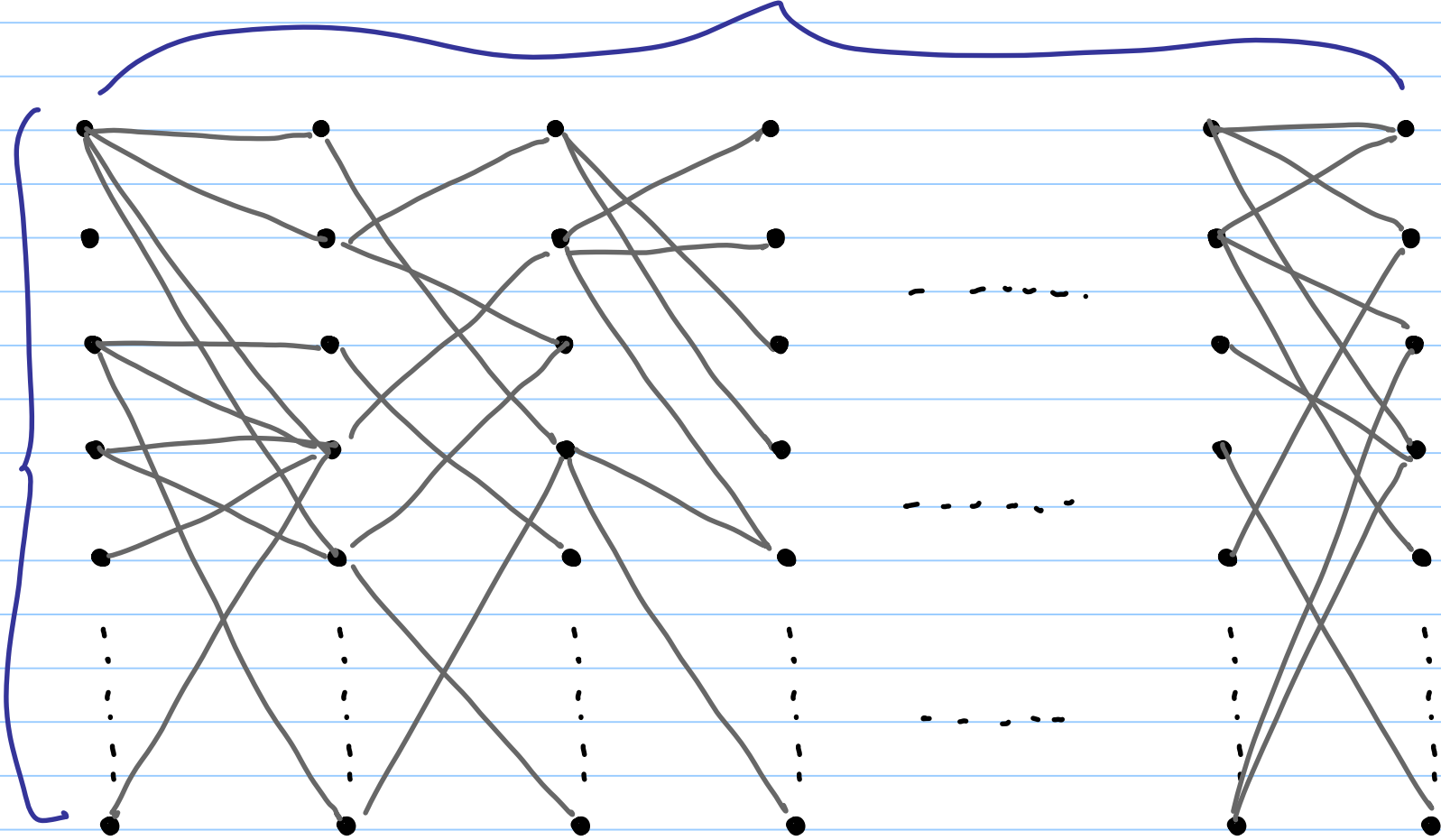
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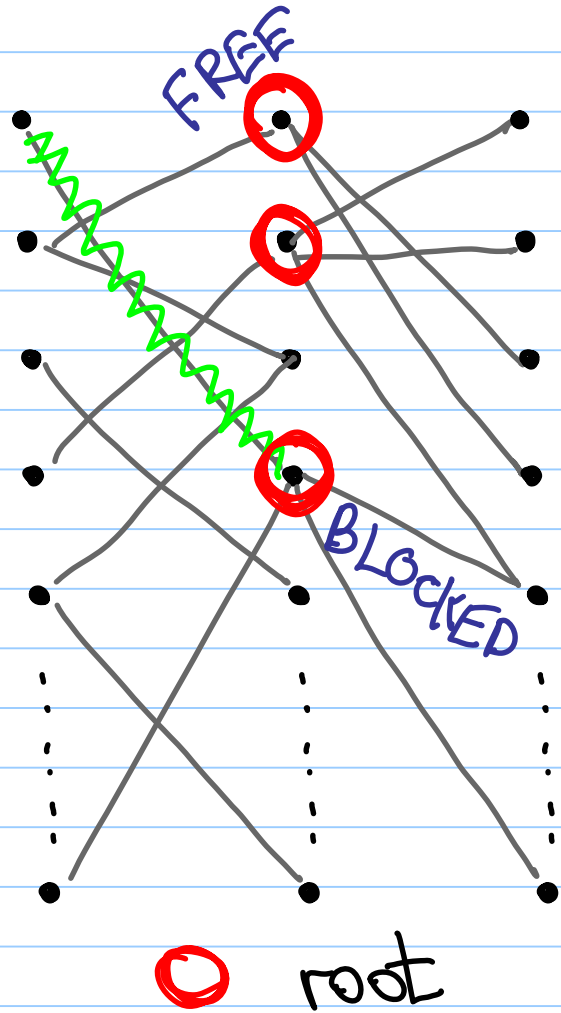
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HOW DOES ALG OPERATE?



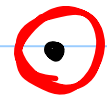
- Edges are revealed only next to **roots** (always n)
- **Blocked roots** - not matched
- **Free roots** - matched (always possible)

$$E[ALG] = \sum_{l=1}^{L-1} E[F_l]$$

number of free roots in layer l

REVEALING EDGES

l



⋮

⋮

⋮



$l+1$



⋮

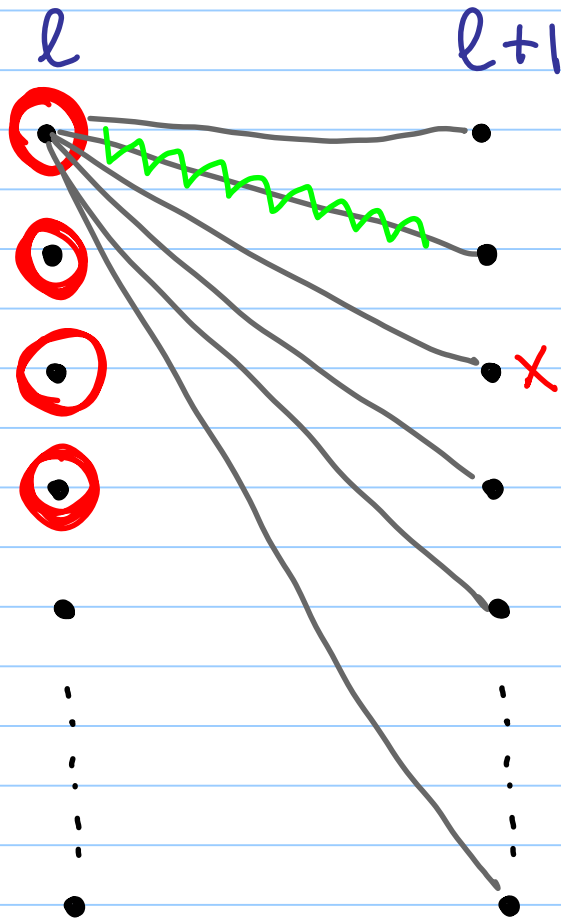
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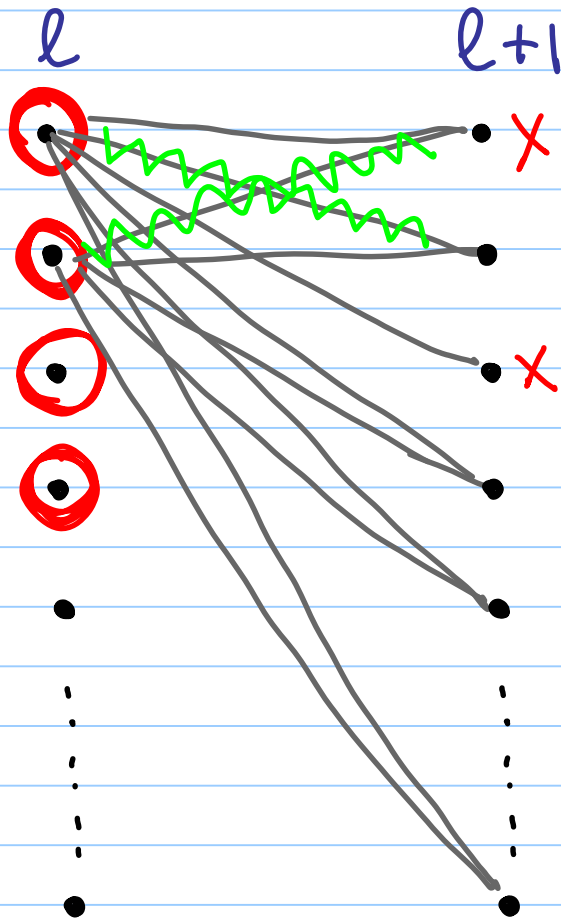
- Pick **random permutation** of the roots
- Connect current root to all undeleted vertices
- Delete a **random vertex**
- Roots of layer $l+1$:
 n undeleted vertices

REVEALING EDGES



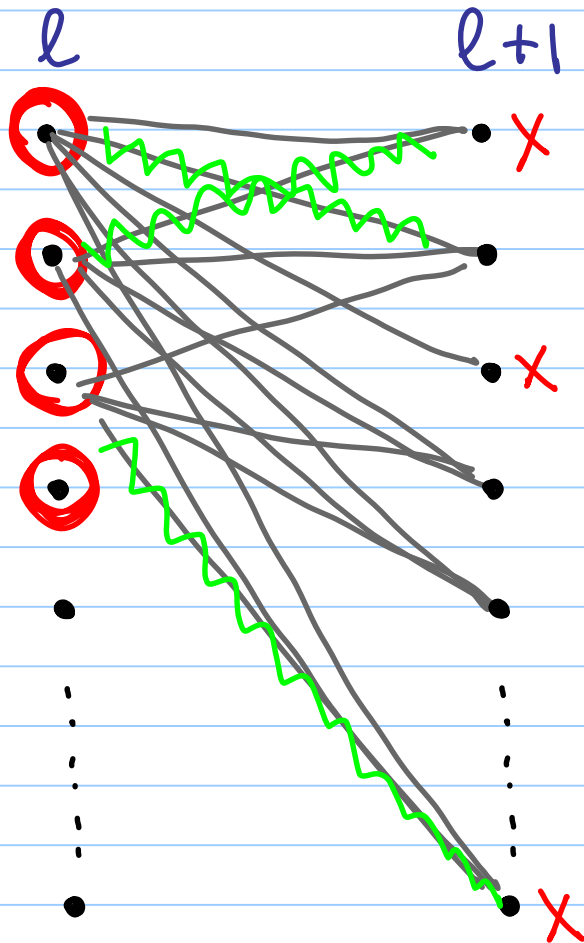
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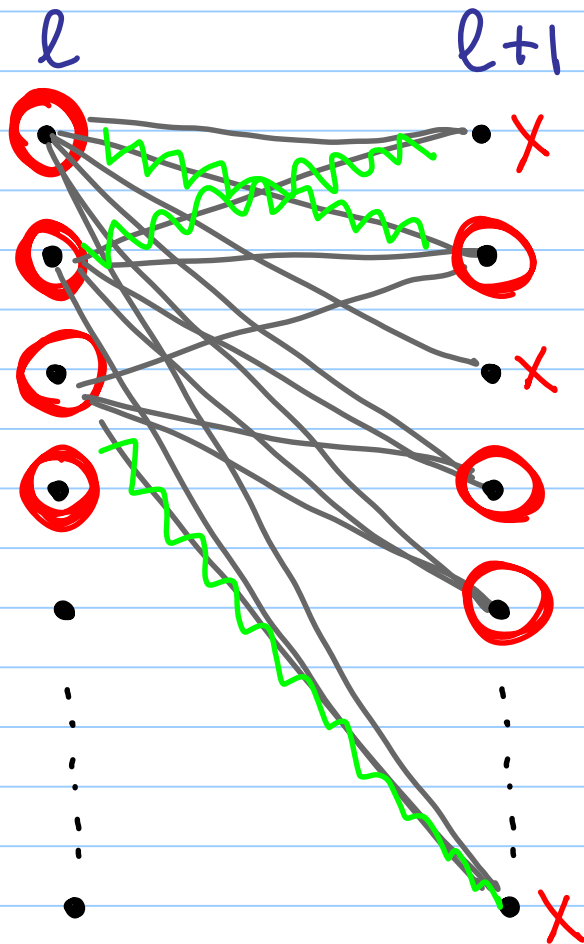
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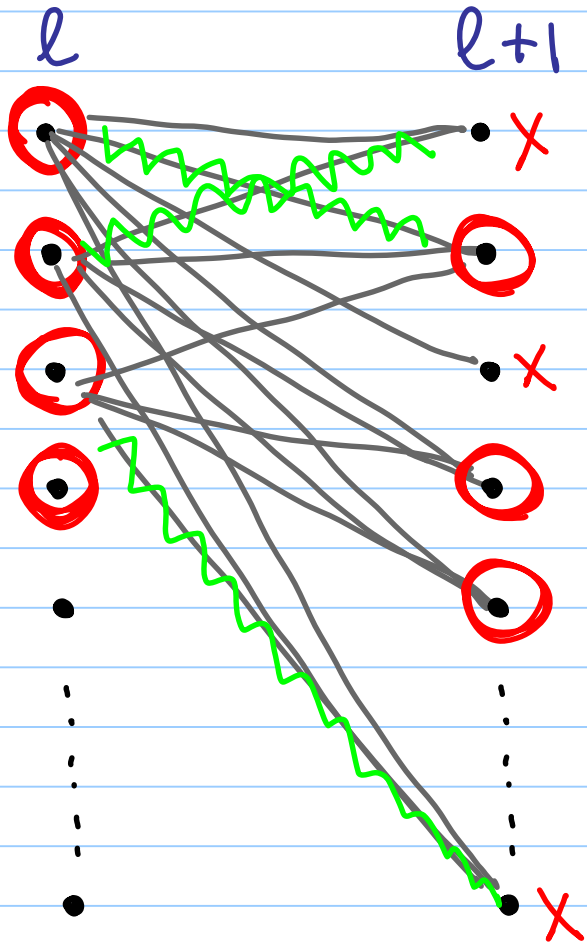
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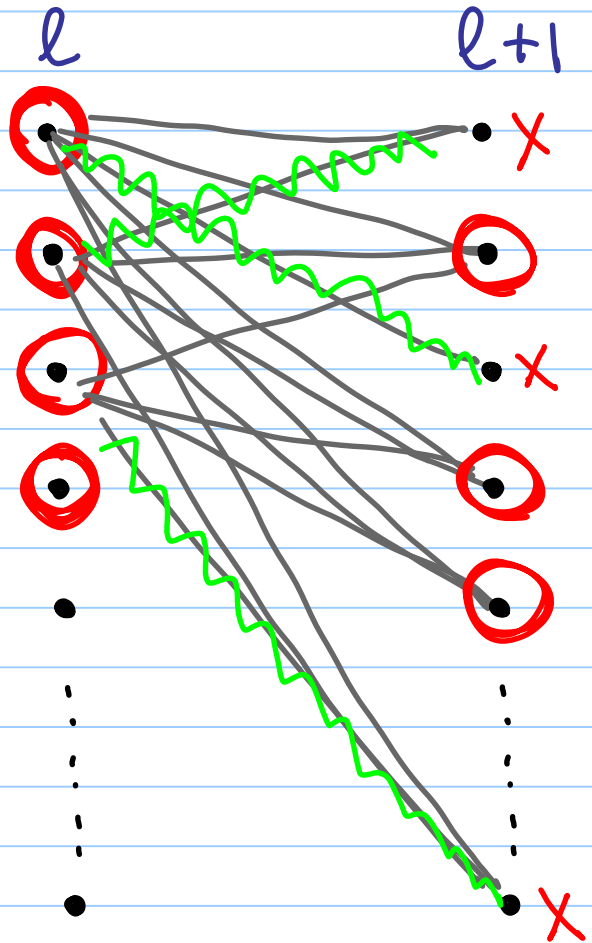


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OBSERVATIONS

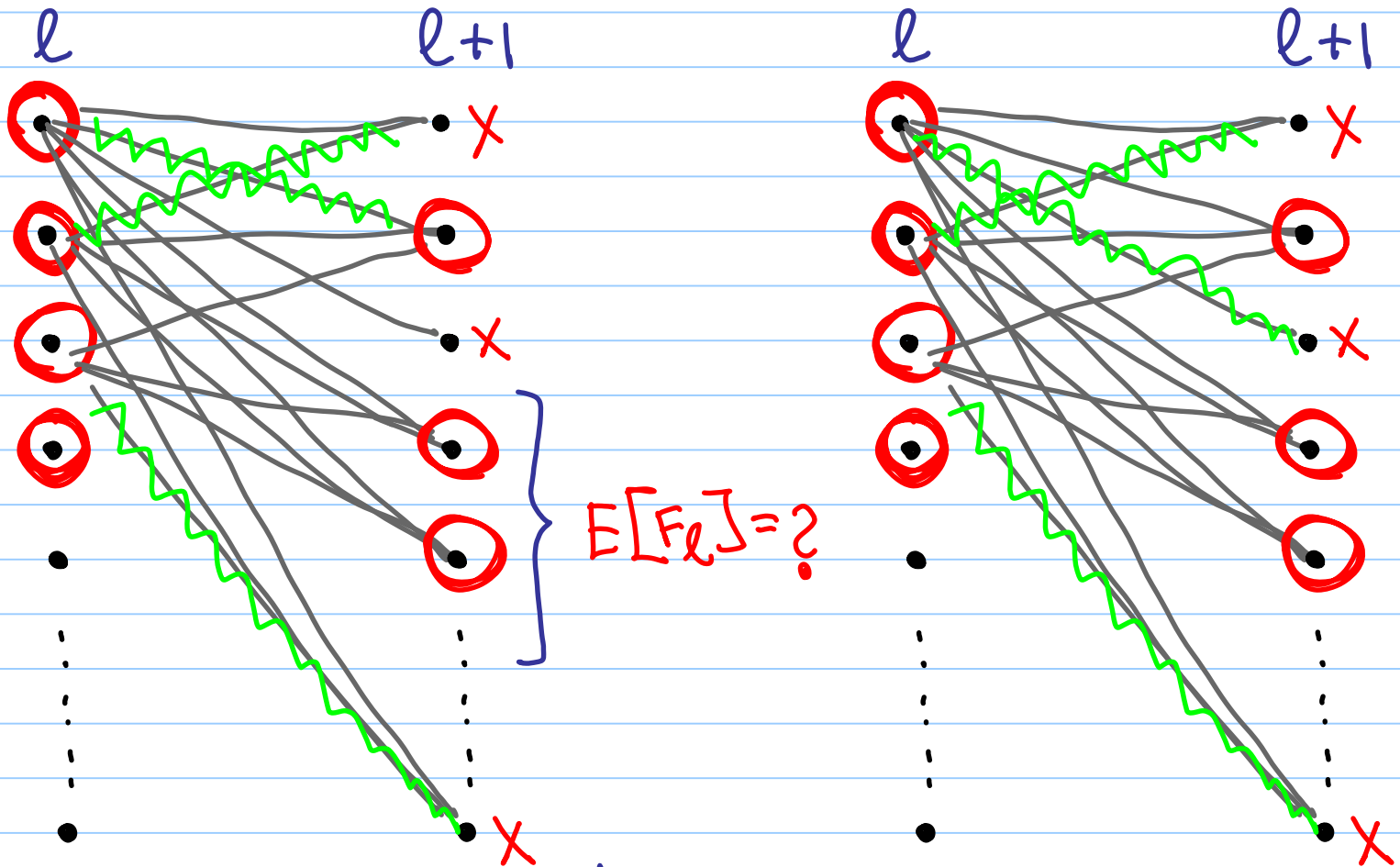


ALG's choices may lead to blocked roots in layer $l+1$



OPT can avoid this
 $OPT = (L-1)N$

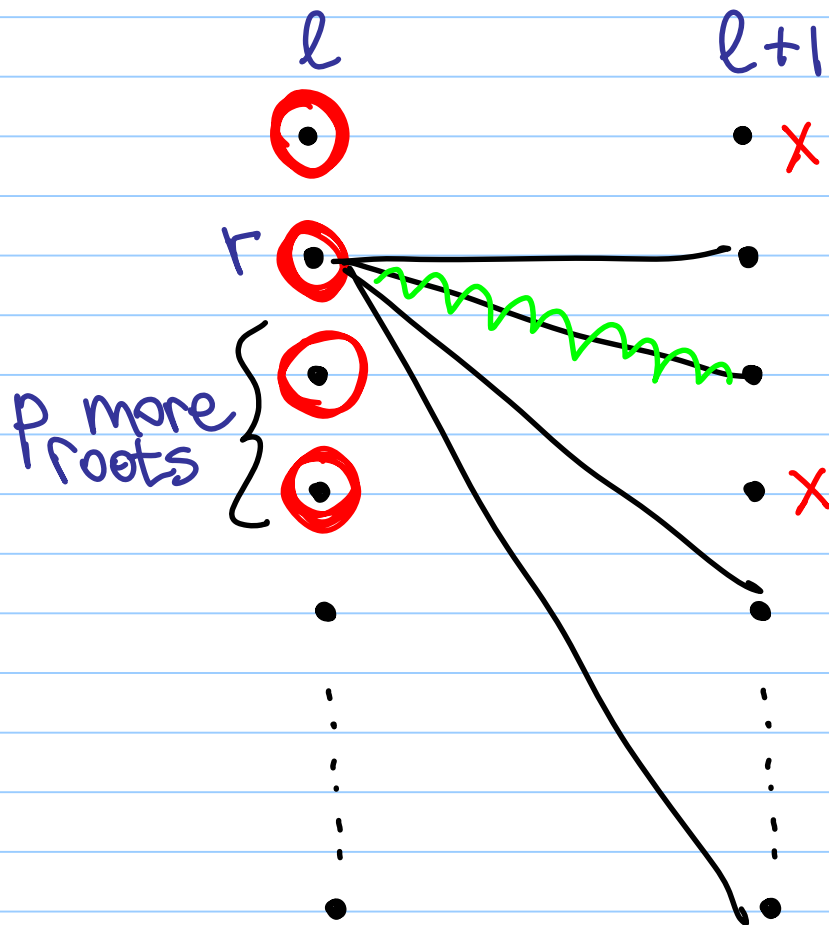
OBSERVATIONS



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PROBABILITY OF CREATING A BLOCKED ROOT

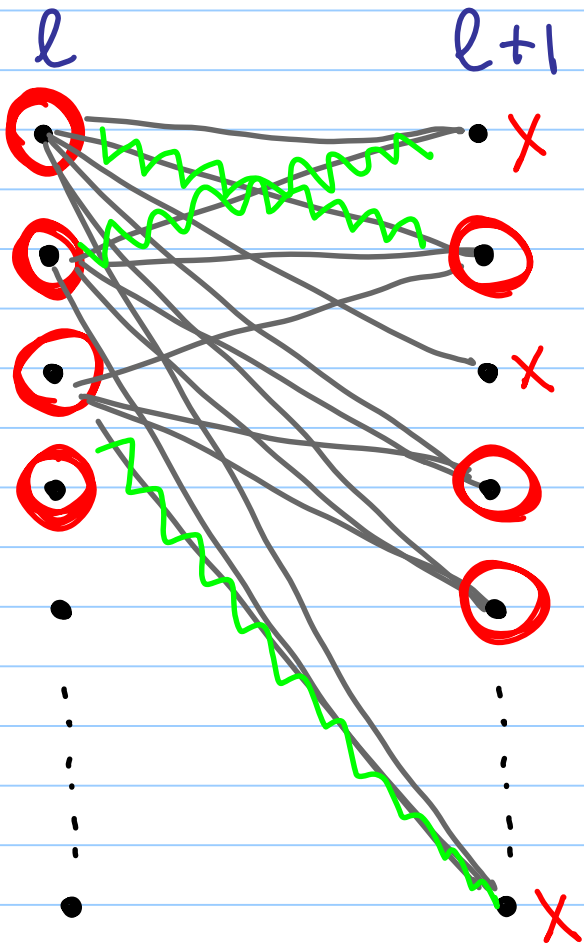


$$\begin{aligned}
 & \Pr [r \text{ creates blocked root} \mid p \text{ more roots}] \\
 &= \left(1 - \frac{1}{n+p}\right) \left(1 - \frac{1}{n+p-1}\right) \dots \left(1 - \frac{1}{n+1}\right) \\
 &= \frac{n}{n+p}
 \end{aligned}$$

$$\begin{aligned}
 & \Pr [r \text{ creates blocked root}] \\
 &= \ln 2 + \mathcal{O}\left(\frac{1}{n}\right) \\
 &= \sum_{p=0}^{n-1} \frac{1}{n} \cdot \frac{n}{n+p}
 \end{aligned}$$

random permutation

COMPUTING $E[F_l]$



$$E[F_{l+1}] = n - E[B_{l+1}]$$

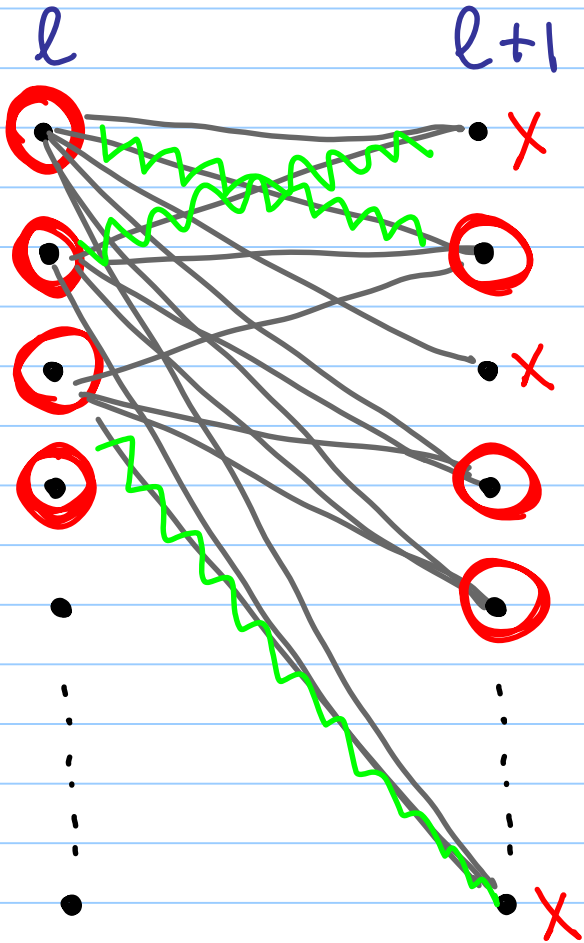
$$= n - E[E[B_{l+1} | F_l]]$$

$$= n - (\ln 2 + \theta(\frac{1}{n})) \cdot E[F_l]$$



$$E[F_l] = \frac{n}{1 + \ln 2} + \theta(n \cdot \ln^2)$$

RESULTING LOWER BOUND



$$E[ALG] = \sum_{l=1}^{L-1} E[F_l]$$

$$= \frac{1}{1 + \ln 2} (L-1)n + \Theta(n)$$

$$OPT = (L-1)n$$



$$\frac{OPT}{E[ALG]} \xrightarrow{L, n \rightarrow \infty} 1 + \ln 2$$

ADDITIONAL RESULTS

UPPER BOUND FOR WEIGHTED GRAPHS

- An upper bound of $3 + 2\sqrt{2} \approx 5.828$ can be attained by a deterministic algorithm [Feigenbaum et al. '05] [McGregor '05]
- This bound is best-possible without randomization [Varadaraja '11]

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NEW RESULT: upper bound of ≈ 5.356
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THANK YOU!