

# Advanced Techniques for Combinatorial Algorithms: Randomized Algorithms

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- Horner's formula.  $2^{m-1} T[i + m] \bmod p$  computed iteratively



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$P[\#FP \geq 1] \leq O(nm/l)$  if  $p$  is randomly (w.r.t. uniform distribution) chosen among all  
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## Decreasing error probability

Choosing  $k$  random primes (independently, without repetitions).

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Let  $K_n$  be the complete graph with  $n$  vertices. If  $\binom{n}{k} 2^{1-\binom{k}{2}} < 1$  then we can color the edges of  $K_n$  so that we have no monochromatic  $K_k$  subgraph.

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- 7  $P[\bigcap \bar{A}_i] = 1 - P[\bigcup A_i] > 0$

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- There exists a cut with at least  $m/2$  edges.

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## Question

The algorithm?

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## Question

How many random cuts are needed?

# The sample and modify method

## Independent set

$G$ : undirected graph. Average degree  $d = 2m/n$

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### Algorithm 1: Random Independent Set

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- 1  $S \leftarrow$  a random sample of  $V$ , each vertex is picked with  $p = 1/d$ ;
  - 2  $I \leftarrow S$ ;
  - 3 **while** *there exists an edge  $e$  in  $G|I$*  **do**
  - 4     Remove an endpoint of  $e$  from  $I$
-

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Bounding  $\|$

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Bounding  $|I|$

- $E(G) = \frac{nd}{2}$



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- At most  $E(G|S)$  are removed in the second step, hence  
 $E[|I|] = E[|S|] - E[E(G|S)] = \frac{n}{2d}$