

# Minicourse on parameterized algorithms and complexity

## Part 3: Randomized techniques

Dániel Marx

Jagiellonian University in Kraków  
April 21-23, 2015

## Why randomized?

- A guaranteed error probability of  $10^{-100}$  is as good as a deterministic algorithm.  
(Probability of hardware failure is larger!)
- Randomized algorithms can be more efficient and/or conceptually simpler.
- Can be the first step towards a deterministic algorithm.

# Polynomial-time vs. FPT randomization

## Polynomial-time randomized algorithms

- Randomized selection to pick a **typical, unproblematic, average** element/subset.
- Success probability is constant or at most polynomially small.

## Randomized FPT algorithms

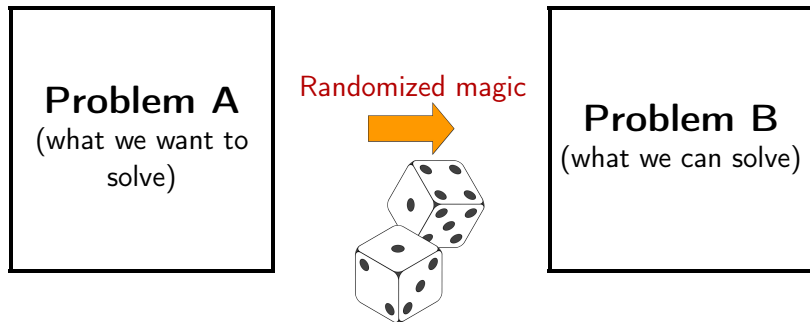
- Randomized selection to satisfy a **bounded number** of (unknown) constraints.
- Success probability might be exponentially small.

# Randomization

There are two main ways randomization appears:

- Algebraic techniques
  - Schwartz-Zippel Lemma
  - Linear matroids
- This lecture: combinatorial techniques.

## Randomization as reduction



# Color Coding

## $k$ -PATH

**Input:** A graph  $G$ , integer  $k$ .

**Find:** A simple path of length  $k$ .

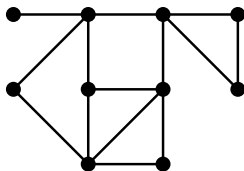
**Note:** The problem is clearly NP-hard, as it contains the HAMILTONIAN PATH problem.

Theorem [Alon, Yuster, Zwick 1994]

$k$ -PATH can be solved in time  $2^{O(k)} \cdot n^{O(1)}$ .

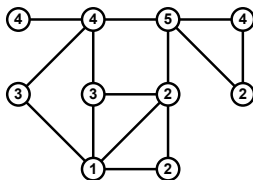
## Color Coding

- Assign colors from  $[k]$  to vertices  $V(G)$  uniformly and independently at random.



## Color Coding

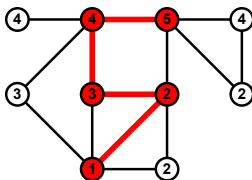
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## Color Coding

- Assign colors from  $[k]$  to vertices  $V(G)$  uniformly and independently at random.



- Check if there is a path colored  $1 - 2 - \dots - k$ ; output “YES” or “NO”.
  - If there is no  $k$ -path: no path colored  $1 - 2 - \dots - k$  exists  $\Rightarrow$  “NO”.
  - If there is a  $k$ -path: the probability that such a path is colored  $1 - 2 - \dots - k$  is  $k^{-k}$  thus the algorithm outputs “YES” with at least that probability.

## Error probability

### Useful fact

If the probability of success is at least  $p$ , then the probability that the algorithm **does not** say “YES” after  $1/p$  repetitions is at most

$$(1 - p)^{1/p} < (e^{-p})^{1/p} = 1/e \approx 0.38$$

## Error probability

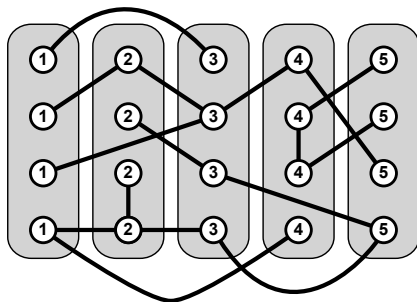
### Useful fact

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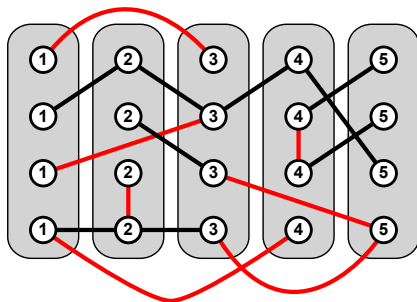
- Thus if  $p > k^{-k}$ , then error probability is at most  $1/e$  after  $k^k$  repetitions.
- Repeating the whole algorithm a constant number of times can make the error probability an arbitrary small constant.
- For example, by trying  $100 \cdot k^k$  random colorings, the probability of a wrong answer is at most  $1/e^{100}$ .

## Finding a path colored $1 - 2 - \dots - k$



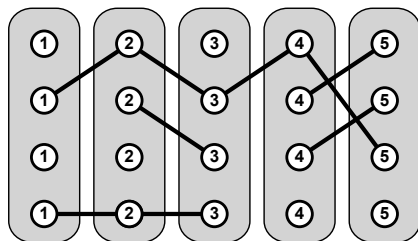
- Edges connecting nonadjacent color classes are removed.
- The remaining edges are directed towards the larger class.
- All we need to check is if there is a directed path from class  $1$  to class  $k$ .

## Finding a path colored $1 - 2 - \dots - k$



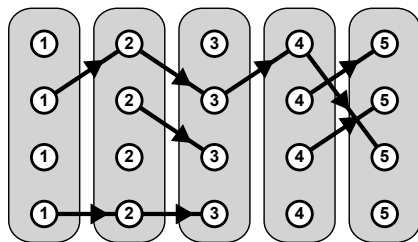
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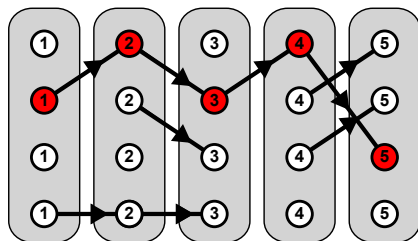
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# Color Coding

$k$ -PATH

Color Coding  
success probability:

$$k^{-k}$$

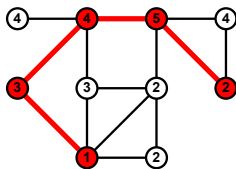


Finding a  
 $1 - 2 - \dots - k$   
colored path

polynomial-time  
solvable

## Improved Color Coding

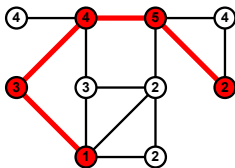
- Assign colors from  $[k]$  to vertices  $V(G)$  uniformly and independently at random.



- Check if there is a **colorful** path where each color appears exactly once on the vertices; output "YES" or "NO".

## Improved Color Coding

- Assign colors from  $[k]$  to vertices  $V(G)$  uniformly and independently at random.



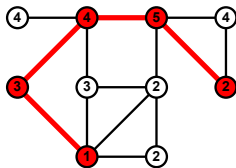
- Check if there is a **colorful** path where each color appears exactly once on the vertices; output “YES” or “NO”.
  - If there is no  $k$ -path: no **colorful** path exists  $\Rightarrow$  “NO”.
  - If there is a  $k$ -path: the probability that it is **colorful** is

$$\frac{k!}{k^k} > \frac{\left(\frac{k}{e}\right)^k}{k^k} = e^{-k},$$

thus the algorithm outputs “YES” with at least that probability.

## Improved Color Coding

- Assign colors from  $[k]$  to vertices  $V(G)$  uniformly and independently at random.



- Repeating the algorithm  $100e^k$  times decreases the error probability to  $e^{-100}$ .

How to find a colorful path?

- Try all permutations ( $k! \cdot n^{O(1)}$  time)
- Dynamic programming ( $2^k \cdot n^{O(1)}$  time)

## Finding a colorful path

### Subproblems:

We introduce  $2^k \cdot |V(G)|$  Boolean variables:

$x(v, C) = \text{TRUE}$  for some  $v \in V(G)$  and  $C \subseteq [k]$

$\Updownarrow$

There is a path  $P$  ending at  $v$  such that each color in  $C$  appears on  $P$  exactly once and no other color appears.

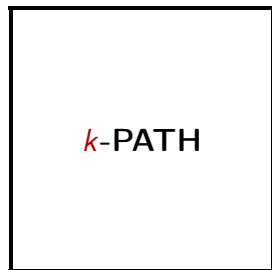
### Answer:

There is a colorful path  $\iff x(v, [k]) = \text{TRUE}$  for some vertex  $v$ .

### Initialization & Recurrence:

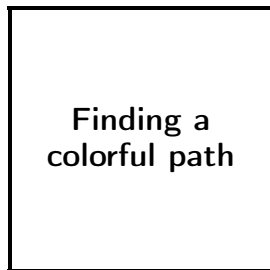
Exercise.

# Improved Color Coding



Color Coding  
success probability:

$$e^{-k}$$



Solvable in time  
 $2^k \cdot n^{O(1)}$

# Derandomization

## Definition

A family  $\mathcal{H}$  of functions  $[n] \rightarrow [k]$  is a  **$k$ -perfect** family of hash functions if for every  $S \subseteq [n]$  with  $|S| = k$ , there is an  $h \in \mathcal{H}$  such that  $h(x) \neq h(y)$  for any  $x, y \in S$ ,  $x \neq y$ .

## Theorem [Alon, Yuster, Zwick 1994]

There is a  $k$ -perfect family of functions  $[n] \rightarrow [k]$  having size  $2^{O(k)} \log n$  (and can be constructed in time polynomial in the size of the family).

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Instead of trying  $O(e^k)$  **random colorings**, we go through a  **$k$ -perfect family**  $\mathcal{H}$  of functions  $V(G) \rightarrow [k]$ .

If there is a solution  $S$

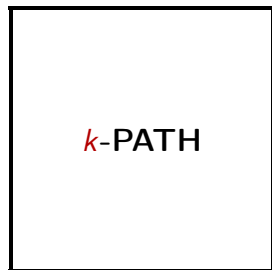
$\Rightarrow$  The vertices of  $S$  are colorful for at least one  $h \in \mathcal{H}$

$\Rightarrow$  Algorithm outputs “YES”.

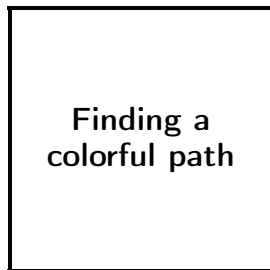
$\Rightarrow$   $k$ -PATH can be solved in **deterministic** time  $2^{O(k)} \cdot n^{O(1)}$ .



# Derandomized Color Coding



$k$ -perfect family  
 $2^{O(k)} \log n$  functions



Solvable in time  
 $2^k \cdot n^{O(1)}$

## Bounded-degree graphs

Meta theorems exist for bounded-degree graphs, but randomization is usually simpler.

### DENSE $k$ -VERTEX SUBGRAPH

**Input:** A graph  $G$ , integers  $k, m$ .

**Find:** A set of  $k$  vertices inducing  $\geq m$  edges.

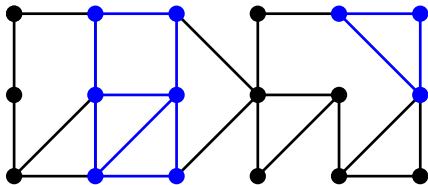
**Note:** on general graphs, the problem is  $W[1]$ -hard parameterized by  $k$ , as it contains  $k$ -CLIQUE.

### Theorem

DENSE  $k$ -VERTEX SUBGRAPH can be solved in randomized time  $2^{k(d+1)} \cdot n^{O(1)}$  on graphs with maximum degree  $d$ .

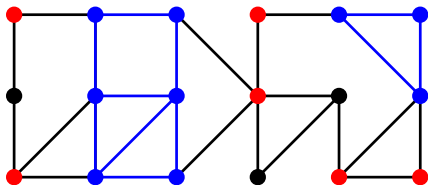
## DENSE $k$ -VERTEX SUBGRAPH

- Remove each vertex with probability  $1/2$  independently.



## DENSE $k$ -VERTEX SUBGRAPH

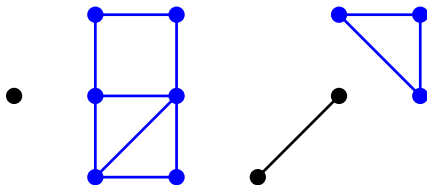
- Remove each vertex with probability  $1/2$  independently.



- With probability  $2^{-k}$  no vertex of the solution is removed.
- With probability  $2^{-kd}$  every neighbor of the solution is removed.
- $\Rightarrow$  We have to find a solution that is the union of connected components!

## DENSE $k$ -VERTEX SUBGRAPH

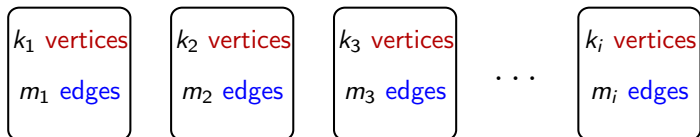
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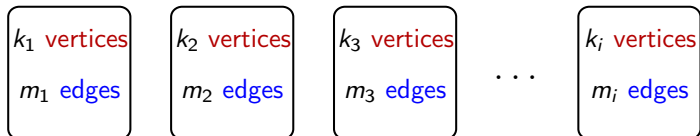
Select connected components with

- at most  $k$  vertices and
- at least  $m$  edges.

What problem is this?

## DENSE $k$ -VERTEX SUBGRAPH

- Remove each vertex with probability  $1/2$  independently.



Select connected components with

- at most  $k$  vertices and
- at least  $m$  edges.

What problem is this?

**KNAPSACK!**

## DENSE $k$ -VERTEX SUBGRAPH

Select connected components with

- at most  $k$  vertices and
- at least  $m$  edges.

This is exactly KNAPSACK!

(I.e., pick objects of total weight at most  $S$  and value at least  $V$ .)

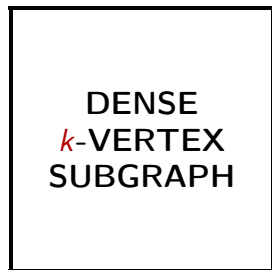
We can interpret

- number of vertices = weight of the items
- number of edges = value of the items

If the weights are integers, then DP solves the problem in time polynomial in the number of objects and the maximum weight.



# DENSE $k$ -VERTEX SUBGRAPH



Random deletions  
success probability:

$$2^{-k(d+1)}$$



Polynomial time

# BALANCED SEPARATION

Useful problem for recursion:

## BALANCED SEPARATION

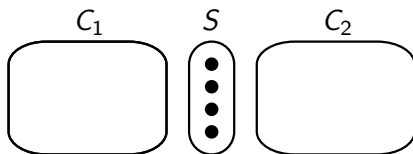
**Input:** A graph  $G$ , integers  $k, q$ .

**Find:** A set  $S$  of at most  $k$  vertices such that  $G \setminus S$  has **at least two** components of size at least  $q$  each.

## Theorem

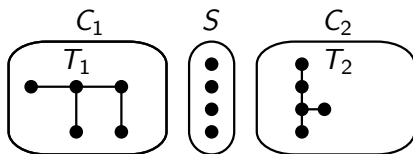
BALANCED SEPARATION can be solved in randomized time  $2^{O(q+k)} \cdot n^{O(1)}$ .

## BALANCED SEPARATION



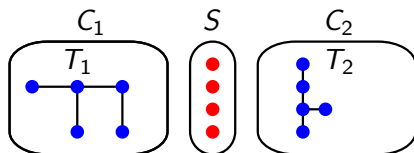
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## BALANCED SEPARATION



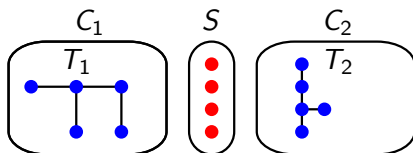
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## BALANCED SEPARATION



- Remove each vertex with probability  $1/2$  independently.
- With probability  $2^{-k}$  every vertex of the solution is removed.
- With probability  $2^{-q}$  no vertex of  $T_1$  is removed.
- With probability  $2^{-q}$  no vertex of  $T_2$  is removed.

## BALANCED SEPARATION



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- With probability  $2^{-q}$  no vertex of  $T_1$  is removed.
- With probability  $2^{-q}$  no vertex of  $T_2$  is removed.
- $\Rightarrow$  The reduced graph  $G'$  has two components of size  $\geq q$  that can be separated in the original graph  $G$  by  $k$  vertices.
- For any pair of large components of  $G'$ , we find a minimum  $s - t$  cut in  $G$ .

# BALANCED SEPARATION

BALANCED  
SEPARATION

Random deletions  
success probability:

$$2^{-(k+2q)}$$



MINIMUM  $s - t$   
CUT

Polynomial time

# Conclusions

- Randomization gives elegant solution to many problems.
- Derandomization is sometimes possible (but less elegant).
- Small (but  $f(k)$ ) success probability is good for us.
- Reducing the problem we want to solve to a problem that is easier to solve.