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Color Coding

CS 602 - Applied Algorithms

K. Mittal, M. Pareek and M. Taraviya

Department of Computer Science
Indian Institute of Technology, Bombay

March 22, 2019



Outline

- 1 Problem Definitions
- 2 Random Orientations
- 3 Random Colorings
- 4 Derandomization
- 5 Related results



k -PATH and k -CYCLE

Given a graph $G = (V, E)$ we are interested in the following problems (in both the directed and undirected case):

- Does the graph contain a path of length k ?
- Does the graph contain a cycle of length k ?



k -PATH and k -CYCLE

Given a graph $G = (V, E)$ we are interested in the following problems (in both the directed and undirected case):

- Does the graph contain a path of length k ?
- Does the graph contain a cycle of length k ?

We'll present ideas from a paper by Alon, Yuster and Zwick [AYZ95].



k -PATH and k -CYCLE

It is easy to see that both of these problems are NP-complete.

- (k -CYCLE) A cycle of length n is the same as a Hamiltonian cycle.
- (k -PATH) A path of length $n - 1$ is the same as a Hamiltonian path.



Theorem

A path of length k , if present, can be found in expected time $O((k+1)! \cdot E)$ in a directed graph and $O((k+2)! \cdot V)$ in an undirected graph.^a

^a $|E|, |V|$ are denoted by E, V for convenience.



k-PATH using Random Orientations

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Proof.

Consider the **directed** case first.

- Choose random permutation π on V .



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If there is a path of length k , it is found with probability $\geq \frac{1}{(k+1)!}$, giving the required expected runtime. □



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This can be made more efficient. □



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This gives expected runtime $O((k+1)! \cdot E)$.

This can be made more efficient.

First do a DFS from an arbitrary vertex. If no path of length k is found, then use the previous algorithm. In such a case, it must be true that

$$|E| \leq k|V|.$$



Deterministic Algorithms

By combining techniques of Monien [Mon85] and Bodlaender [Bod93], one can also get deterministic algorithms achieving runtime of $O(k! \cdot V)$ and $O(k! \cdot E)$ for undirected and directed graphs respectively.



Theorem

A cycle of length, if present, k can be found in expected time $O((k-1)! \log k \cdot V^\omega)$ in a directed or an undirected graph, where ω is the matrix multiplication exponent.



k-CYCLE using Random Orientations

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Random colorings

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Random colorings

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- A path is said to be **colorful** if each vertex on it is colored by a distinct color.
- Each simple path of length $k - 1$ has a chance of $\frac{k!}{k^k} > e^{-k}$ to become colorful.



k-PATH using Random Colorings

Lemma

Given a graph G and a coloring $c : V \rightarrow [k]$, a colorful path of length $k - 1$, if it exists, can be found in $2^{O(k)} \cdot E$ time.



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 - A vertex v and a length i ($1 \leq i \leq k$).
 - A set $S \subseteq \binom{[k]}{i}$ of all possible color sets on some colorful path of length i from s to v .



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- Perform updates by iterating over i , and over E within each iteration.



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- Perform updates by iterating over i , and over E within each iteration.

The total number of states is $(\sum_{i=0}^k i \cdot \binom{k}{i}) \cdot V = 2^{O(k)} \cdot V$ and the runtime is $2^{O(k)} \cdot E$. □

Lemma

Given a graph G and a coloring $c : V \rightarrow [k]$, all pairs of vertices connected by colorful paths of length $k - 1$ can be found in $2^{O(k)} \cdot VE$ or $2^{O(k)} \cdot V^\omega$ time.



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The $2^{O(k)} \cdot VE$ algorithm is obtained by simply running the previous algorithm $|V|$ times.



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- Take OR of the matrices $A_1 B A_2$ over all partitions $\{C_1, C_2\}$.

The number of matrix multiplications is given by

$$N(k) = (2 \cdot N(k/2) + 2) \times \binom{k}{k/2} = 2^{O(k)}.$$



Theorem

In a graph G , a path of length $k - 1$, if it exists, can be found in $2^{O(k)} \cdot V$ expected time in the undirected case and in $2^{O(k)} \cdot E$ expected time in the directed case.



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Theorem

In a graph G , a cycle of length k , if it exists, can be found in expected time $2^{O(k)} \cdot VE$ or $2^{O(k)} \cdot V^\omega$.



k -Perfect Family of Hash Functions

A family of hash functions from $[n] \rightarrow [k]$ is called k -perfect if for all $S \subseteq [n]$, $|S| = k$, there is a function in the family that is one to one on S .



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Using ideas from Schmidt and Siegal [SS90] and Moni Naor, the existence of such families of size $2^{O(k)} \log(n)$ can be shown.



Derandomizing Random Colorings

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Derandomization

We can use the above family to get colorings in which for each k subset V' of V , there is a coloring that assigns each vertex in V' a distinct color. This incurs an extra $\log(|V|)$ multiplicative factor in the runtime.



- k -PATH problem is in P for $k \leq \log(|V|)$ (this is the LOG PATH problem).
- The derandomization of color-coding method can be easily parallelized, yielding efficient NC algorithms.
- Cycles of length k for $k \leq 7$ can be found in time $O(V^\omega)$. [AYZ97]



Counting the number of paths of length k

k -Perfect Family of Balanced Hash Functions

A family of hash functions from $[n] \rightarrow [k]$ is called k -perfect balanced if for some $T > 0$, we have that for all $S \subseteq [n]$, $|S| = k$, the number of functions in the family such that $f(S) = [k]$ is exactly T .



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Do we have a small k -perfect balanced family of functions?



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No (Proof later).



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But we can **approximate**!

Allow the number to be between $(1 - \epsilon)T$ and $(1 + \epsilon)T$.



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Theorem

There is an explicit construction of an ϵ -balanced family of functions from $[n]$ to $[k]$ consisting of $e^{(1+o(1))k} \log n$ functions. Such a family can be constructed in time $e^{(1+o(1))k} n \log n$.



Size of k -perfect balanced family of functions

Theorem

*If F is a k -perfect balanced family of functions $\{f : [n] \rightarrow [k]\}$,
 $|F| \geq c(k)n^{k/2}$.*



Size of k -perfect balanced family of functions

Theorem

If F is a k -perfect balanced family of functions $\{f : [n] \rightarrow [k]\}$,
 $|F| \geq c(k)n^{k/2}$.

Proof.

- For each $R \subseteq [n]$, $|R| = k/2$ consider vectors u_R and w_R of length $\binom{k}{k/2} \cdot |F|$, where
 - $u_R(f, S) = 1$ if $f(R) = S$ (0 otherwise)
 - $w_R(f, S) = 1$ if $f(R) = [k] \setminus S$ (0 otherwise)



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- Let M_u be matrix with u_R as rows.
- Let M_w be matrix with w_Q as columns.



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- Let M_u be matrix with u_R as rows.
- Let M_w be matrix with w_Q as columns.
- $M_u M_w$ has full rank $\implies \binom{k}{k/2} \cdot |F| \geq \binom{n}{k/2} \implies |F| \geq c(k)n^{k/2}$.



Thank You!



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