Fast Exact Algorithm for L(2,1)-Labeling of Graphs

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joint work with:

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DP algorithm

combinatorial result

faster exact algorithm

conclusion

Outline

1 Definitions and Known Results

- 2 A (Simple) Dynamic Programming Based Algorithm
- **(3) A Combinatorial Result**
- (4) A Faster Exact Exponential-Time Algorithm

5 Conclusion

DP algorithm

combinatorial result

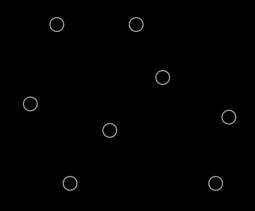
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conclusion

Frequency assignment problem

broadcast network

- assign frequencies to transmitters
- avoid undesired interference



DP algorithm

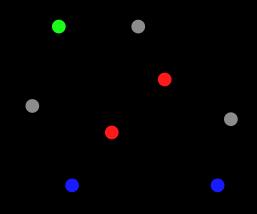
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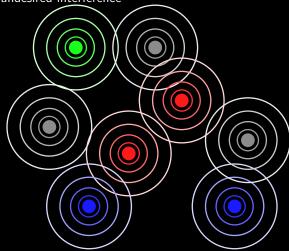
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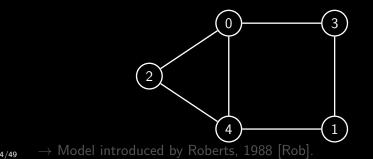
Definition of L(2, 1)-labeling

L(2,1)-LABELING

Input : A graph G = (V, E). **Question** : Compute a function ℓ of minimum span k $\ell : V \to \{0, \dots, k\}$ s.t.

• *u* and *v* adjacent
$$\Rightarrow |\ell(u) - \ell(v)| \ge 2$$

• *u* and *v* at distance two
$$\Rightarrow |\ell(u) - \ell(v)| \ge 1$$



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Known complexity results

Theorem [GY92] Determining the minimum span $\lambda(G)$ of a graph G is NP-hard.

Theorem[FKK01]Deciding whether $\lambda(G) \leq k$ remains NP-complete for every fixed $k \geq 4$.(trivial for $k \leq 3$)

Theorem

[CK96, FGK05]

When the span k is part of the input, L(2,1)-labeling problem is polynomial time solvable on trees. However, the problem is NP-complete for series-parallel graphs.

 \rightarrow The problem "separates" graphs of treewidth 1 and 2 by P / NP-completeness dichotomy.

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Known complexity results

The distance constrained labeling problem is more difficult than ordinary coloring :

[FGK05] Theorem Deciding whether $\lambda(G) \leq k$ is NP-complete for series-parallel graphs (k is part of the input).

[BKTvL04, JKM09] Theorem Deciding whether $\lambda = k$ is NP-complete for planar graphs for k = 8[BKTvL04] [JKM09]

• for
$$k = 4$$

L(2,1)-labeling and Locally Injective Homomorphisms

Fiala and Kratochvíl defined the notion of H(2, 1)-labeling :

- mapping from vertices of G to vertices of a graph H;
- ▶ adjacent vertices in *G* are mapped onto non-adjacent vertices in *H*;
- vertices with a common neighbor in G are mapped onto distinct vertices of H.

They show that :

 \rightarrow H(2,1)-labelings are exactly locally injective homomorphisms from G to \overline{H} .

 \rightarrow L(2, 1)-labeling of span k is a locally injective homomorphism into the complement of the path of length k.

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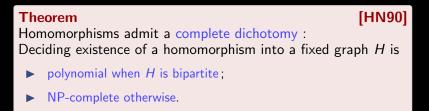
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L(2, 1)-labeling and Locally Injective Homomorphisms

homomorphism : A mapping $f : V(G) \rightarrow V(H)$ is a homomorphism from *G* to *H* if $f(u)f(v) \in E(H)$ for every edge $uv \in E(G)$.



Remark : k-coloring of a graph G corresponds to homomorphism from G to the graph K_k .

homomorphism : A mapping $f : V(G) \rightarrow V(H)$ is a homomorphism from G to H if $f(u)f(v) \in E(H)$ for every edge $uv \in E(G)$.

locally injective homomorphism (LIH) : A homomorphism $f: G \to H$ is locally injective if for every vertex $u \in V(G)$ its neighborhood is mapped injectively into the neighborhood of f(u) in H, i.e., every two vertices having a common neighbor in G are mapped onto disctinct vertices in H.

Theorem

[HKKKL11]

H-locally-injective-homorphism can be solved in time

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[HKKKL11]

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L(2,1)-labeling problem - Exact algorithms

Theorem[HKKKL11]H-locally-injective-homorphism can be solved in time $O^*((\Delta(H) - 1)^n)$

\rightarrow L(2,1)-labeling of span k is a locally injective homomorphism into the complement of the path of length k.

Theorem [HKKKL11] Hence, L(2,1)-labeling problem of span k can be decided in time $O^*((k-1)^n)$

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L(2,1)-labeling problem - Exact algorithms

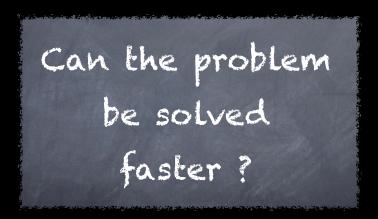
Theorem $L(2, 1)$ -labeling of span 4 : $O(1.3006^n)$	[HKKKL11] (branching)
Theorem $L(2, 1)$ -labeling of span 5 in cubic graphs : $O(1)$	$[GKC10] \\ 1.8613^n) \to O(1.7990^n)$
Theorem $L(2, 1)$ -labeling of min span : $O^*(4^n)$	[Kráľ'06]
Theorem $L(2,1)$ -labeling of min span : $O^*(15^{n/2}) = O(3)$	[HKKKL11] 3.88 ⁿ) (D.P.)
Theorem $L(2, 1)$ -labeling of min span : $O^*(3^n)$	$[CK11]$ (fast ζ transform + IE.)

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A DP based algorithm for L(2, 1)-labeling of min span

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(2) A (Simple) Dynamic Programming Based Algorithm

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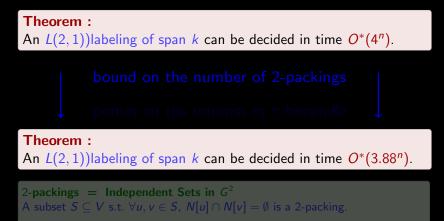
How to compute an L(2, 1)-labeling of span k by Dynamic Programming?

First, we show the following :

Theorem : An L(2, 1) labeling of span k can be decided in time $O^*(4^n)$.

How to compute an L(2, 1)-labeling of span k by Dynamic Programming?

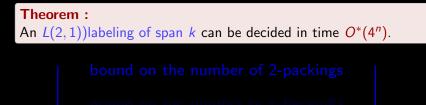
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(2-packing \equiv set of vertices pairwise at distance greater than 2.)

How to compute an L(2,1)-labeling of span k by Dynamic Programming?

First, we show the following :



Theorem : An L(2, 1) labeling of span k can be decided in time $O^*(3.88^n)$.

2-packings = Independent Sets in G^2 A subset $S \subseteq V$ s.t. $\forall u, v \in S$, $N[u] \cap N[v] = \emptyset$ is a 2-packing.

(2-packing \equiv set of vertices pairwise at distance greater than 2.)

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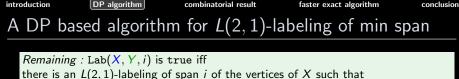
A DP based algorithm for L(2, 1)-labeling of min span

Remaining : Let G = (V, E) be a graph. An L(2, 1)-labeling of span k asks to find a labeling f of G such that :

- ▶ for all $\{u, v\} \in E \implies |f(u) f(v)| \ge 2$;
- ▶ for all $u, v \in V$ s.t. $dist(u, v) = 2 \implies f(u) \neq f(v)$.

 $\forall i \in \{0, 1, \dots, k\}$ and $\forall X, Y \subseteq V$ such that $X \cap Y = \emptyset$, we define the boolean variable Lab(X, Y, i).

Lab(X, Y, i) is true iff there is an L(2, 1)-labeling of span i of the vertices of X such that the vertices of $N(Y) \cap X$ have label at most i - 1.



the vertices of $N(Y) \cap X$ have label at most i - 1.

It is not difficult to check that

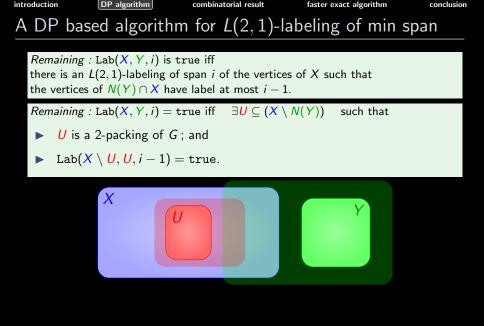
▶
$$Lab(\emptyset, Y, i) \leftarrow true \forall Y, \forall i;$$

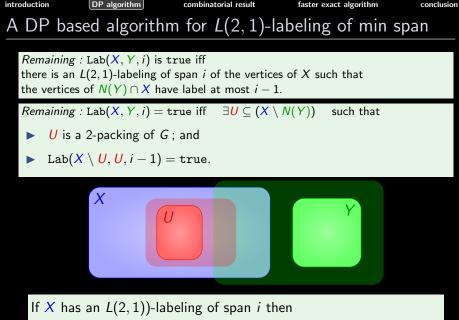
Then, Lab(X, Y, i) is computed by considering the sets X and Y by increasing order of cardinality, and by increasing value of i:

$$Lab(X, Y, i) = true iff \quad \exists U \subseteq (X \setminus N(Y))$$
 such that

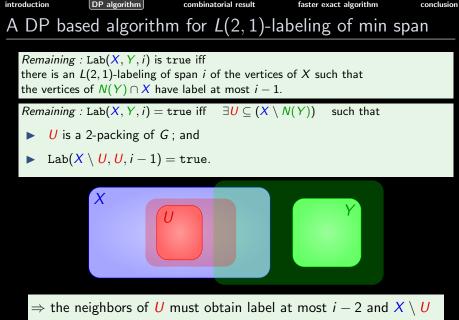
▶
$$Lab(X \setminus U, U, i-1) = true.$$

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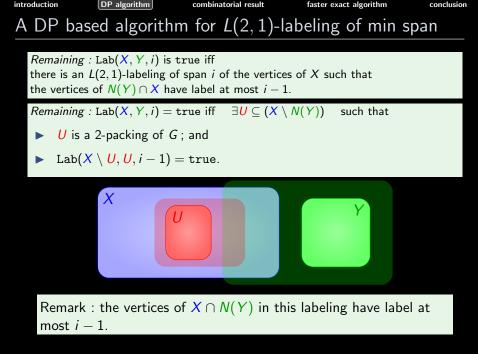




there is a (possibly empty) set $U \subseteq X \setminus N(Y)$ of vertices having label *i*. This set is a 2-packing of *G*.



 \Rightarrow the neighbors of U must obtain label at most i - 2 and $X \setminus U$ must have an L(2, 1)-labeling of span at most i - 1. If a such labeling exists then $Lab(X \setminus U, U, i - 1) = true$.



Running-time analysis :

Lab(X, Y, i) is computed for all $X, Y \subseteq V$ such that $X \cap Y = \emptyset$, and for all $i \in \{0, 1, ..., k\}$.

For each X, Y, we compute all sets $U \subseteq X$ being 2-packings of G.

$$k \cdot \sum_{x=0}^{n} \left(\binom{n}{x} \sum_{y=0}^{n-x} \binom{n-x}{y} \sum_{u=0}^{x} \binom{x}{u} \right)$$

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A DP based algorithm for L(2, 1)-labeling of min span

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$$k \cdot \sum_{x=0}^{n} \left(\binom{n}{x} \sum_{y=0}^{n-x} \binom{n-x}{y} \sum_{u=0}^{x} \binom{x}{u} \right)$$
$$= k \cdot \sum_{x=0}^{n} \left(\binom{n}{x} 2^{n-x} 2^{x} \right)$$
$$= k \cdot 2^{n} \cdot 2^{n}$$

Theorem : Computing an L(2,1) of span k can be obtain in time $O^*(4^n)$.

[HKKKL11]

A DP based algorithm for L(2, 1)-labeling of min span

By using a bound on the number of 2-packing of a certain size,

Theorem

Let u_k be the number of 2-packings of size k in a connected graph. Then, (n/2)

 $u_k \leq \binom{n/2}{k} \cdot 2^k$

$$u_k = 0$$
 for $k > n/2$

we are able to prove that :

Theorem : An L(2,1) of span k can be obtain in time $O^*(4^n) \rightsquigarrow O^*(3.8730^n)$.

[improving upon Král's result]

Note : These results can be extended to L(p, q)-labelings.

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An auxiliary combinatorial result

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2-Packings and Proper Pairs

Like *independent sets* are heavily related to colorings, it seems that 2-*packings* are related to L(2, 1)-labelings.

Theorem : An L(2, 1) of span k can be obtain in time $O^*(2.6488^n)$.

But in fact we need another combinatorial object :

Proper Pairs

... and we need a bound on its maximum number in a graph.

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... and Proper Pairs

Definition A pair (S, X) of subsets of V is a proper pair if $S \cap X = \emptyset$ and S is a 2-packing.

Definition

The number of proper pairs in a graph G is given by

$$pp(G) = \sum_{2- ext{packings } S} 2^{n-|S|}$$

Let $pp(n) = \max pp(G)$ be the maximum number of proper pairs in a connected graph with *n* vertices.

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Let $pp(n) = \max pp(G)$ be the maximum number of proper pairs in a connected graph with *n* vertices.

Theorem

$2.6117^n \le pp(n) \le 2.6488^n$

(will be very useful in the next)

trod	

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... and Proper Pairs

Proof.1/2Let G = (V, E) be a connected graph.Fact 1. If S is a 2-packing, then S is also a 2-packing of $G = (V, E \setminus e)$, for any edge e. \Rightarrow we can assume that G is a tree.

Fact 2. Suppose that there are two leaves which have a common neighbor. Every 2-packing in G is also a 2-packing in H.



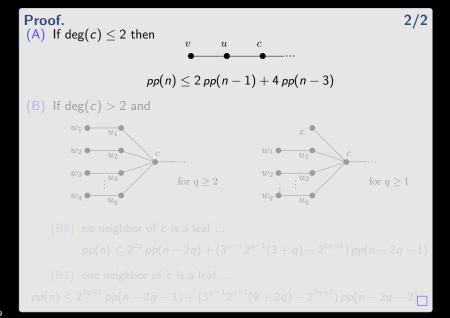
 \Rightarrow we can assume that there are no two or more leaves with a common neighbor

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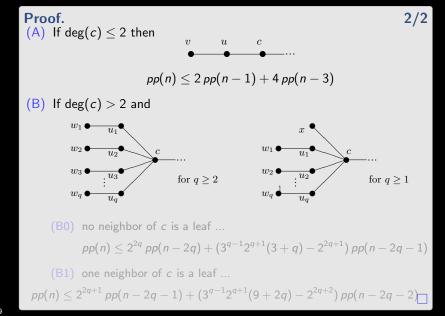




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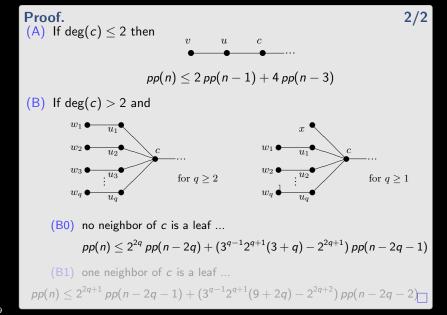




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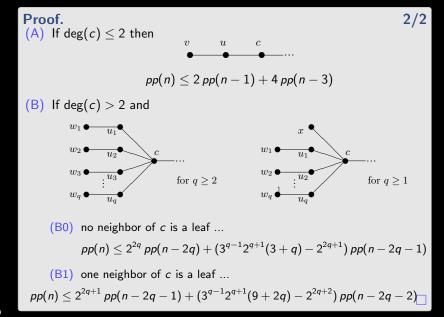


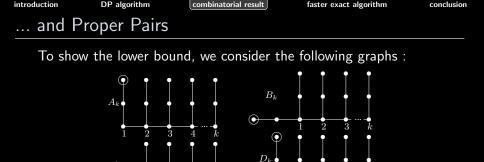


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 $\begin{bmatrix} a_k = 2b_{k-1} + 4a_{k-1} \\ b_k = 2c_k + 2d_k \\ c_k = 2a_k + 12d_{k-1} \end{bmatrix}$

 $d_k = 4d_{k-1} + 12a_{k-1}$

 $2.6117^n \le pp(n) \le 2.6488^n$

2

 C_k

Theorem

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 $\frac{1}{2}$ $\frac{1}{3}$ $\frac{1}{k}$

An Exact Exponential-Time Algorithm

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One key ingredient of our algorithm

Main idea : Use algebraic manipulations similar to

fast matrix multiplication

Assume that A and B are $2^k \times 2^k$ matrices.

$$A = \begin{pmatrix} A_{1,1} & A_{1,2} \\ A_{2,1} & A_{2,2} \end{pmatrix} \quad B = \begin{pmatrix} B_{1,1} & B_{1,2} \\ B_{2,1} & B_{2,2} \end{pmatrix} \quad C = \begin{pmatrix} C_{1,1} & C_{1,2} \\ C_{2,1} & C_{2,2} \end{pmatrix}$$

where

$$\begin{array}{l} C_{1,1} = A_{1,1} \cdot B_{1,1} + A_{1,2} \cdot B_{2,1} \\ C_{1,2} = A_{1,1} \cdot B_{1,2} + A_{1,2} \cdot B_{2,2} \\ C_{2,1} = A_{2,1} \cdot B_{1,1} + A_{2,2} \cdot B_{2,1} \\ C_{2,2} = A_{2,1} \cdot B_{1,2} + A_{2,2} \cdot B_{2,2} \end{array}$$

Thus, 8 matrix multiplications of $2^{k-1} \times 2^{k-1}$ matrices are necessary :

$$T(n) = 8 \cdot T(n/2) = O(n^3)$$

One key ingredient of our algorithm

$$A = \begin{pmatrix} A_{1,1} & A_{1,2} \\ A_{2,1} & A_{2,2} \end{pmatrix} \quad B = \begin{pmatrix} B_{1,1} & B_{1,2} \\ B_{2,1} & B_{2,2} \end{pmatrix} \quad C = \begin{pmatrix} C_{1,1} & C_{1,2} \\ C_{2,1} & C_{2,2} \end{pmatrix}$$

$$\begin{array}{l} \text{trassen [Stra69]}:\\ M_1 = (A_{1,1} + A_{2,2}) \cdot (B_{1,1} + B_{2,2})\\ M_2 = (A_{2,1} + A_{2,2}) \cdot B_{1,1}\\ M_3 = A_{1,1} \cdot (B_{1,2} - B_{2,2})\\ M_4 = A_{2,2} \cdot (B_{2,1} - B_{1,1})\\ M_5 = (A_{1,1} + A_{1,2}) \cdot B_{2,2}\\ M_6 = (A_{2,1} - A_{1,1}) \cdot (B_{1,1} + B_{1,2})\\ M_7 = (A_{1,2} - A_{2,2}) \cdot (B_{2,1} + B_{2,2}) \end{array}$$

and

By St

$$C_{1,1} = M_1 + M_4 - M_5 + M_7$$

$$C_{1,2} = M_3 + M_5$$

$$C_{2,1} = M_2 + M_4$$

$$C_{2,2} = M_1 - M_2 + M_3 + M_6$$

Then,

$$T(n) = 7 \cdot T(n/2) = O(n^{2.807})$$

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Our approach

Our algorithm uses Dynamic Programming

We reduce the number of operations (like in Strassen's algo)

We use a representation for partial L(2, 1)-labelings

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Representation of partial L(2, 1)-labelings

Span 1

Table T_1



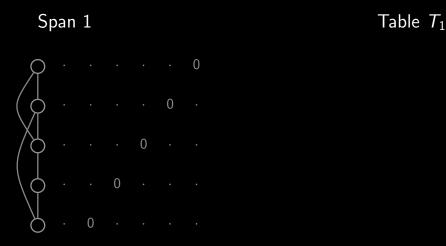


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Representation of partial L(2, 1)-labelings



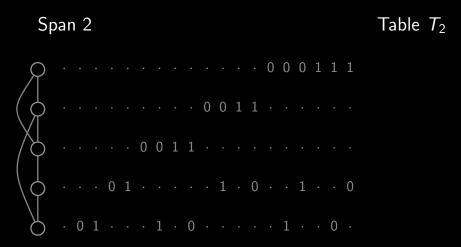
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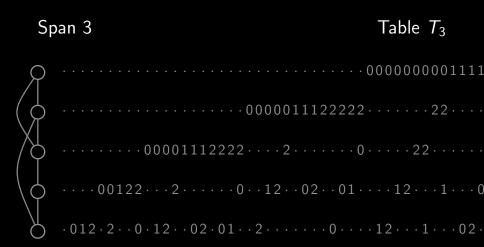
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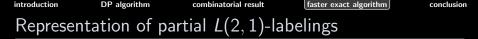
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Representation of partial L(2,1)-labelings





Jump to a compact representation

Table T_{ℓ} contains a vector $\vec{a} \in \{0, \overline{0}, 1, \overline{1}\}^n$ if and only if there is a partial labeling $\varphi \colon V \to \{0, \dots, \ell\}$ such that :

▶ $a_i = 0$ iff v_i is not labeled by φ and there is no neighbor u of v_i with $\varphi(u) = \ell$

▶ $a_i = \overline{0}$ iff v_i is not labeled by φ and there is a neighbor u of v_i with $\varphi(u) = \ell$

$$\bullet \quad \mathsf{a}_i = \mathbf{1} \quad \text{iff} \quad \varphi(\mathsf{v}_i) < \ell$$

$$\blacktriangleright \quad a_i = \overline{1} \quad \text{ iff } \quad \varphi(\mathsf{v}_i) = \ell$$



Representation of partial L(2, 1)-labelings



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Computing the tables

How to compute table $T_{\ell+1}$ from table T_{ℓ} ?

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Computing the tables

Let $P \subseteq \{0,1\}^n$ be the encodings of all 2-packings of G. Formally, $\vec{p} \in P \Leftrightarrow \exists$ a 2-packing $S \subseteq V$ such that $\forall i, p_i = 1$ iff $v_i \in S$.

We compute $T_{\ell+1}$ from $T_{\ell} \oplus P$. We define the partial function \oplus : $\{0, \overline{0}, 1, \overline{1}\} \times \{0, 1\} \rightarrow \{0, 1, \overline{1}\}$:

We generalize \oplus to vectors :

$$a_1a_2\ldots a_n\oplus b_1b_2\ldots b_n = egin{cases} (a_1\oplus b_1)\ldots (a_n\oplus b_n) & ext{if }\oplus ext{ is defined} \\ ext{undefined} & ext{otherwise} \end{cases}$$

introduction	DP algorithm	combinatorial result	faster exact algorithm	conclusion
Comput	ing the table	S		

Then $T_{\ell} \oplus P$ is already almost the same as $T_{\ell+1}$:

 $ec{a} \in \mathcal{T}_{\ell+1}$ iff there is an $ec{a'} \in \mathcal{T}_\ell \oplus P$ such that

▶ $a_i = 0$ iff $a'_i = 0$ and there is no $v_j \in N(v_i)$ with $a'_i = \overline{1}$

$$\blacktriangleright \quad a_i = \overline{0} \text{ iff } a_i' = 0 \text{ and there is a } v_j \in N(v_i) \text{ with } a_j' = \overline{1}$$

$$a_i = 1 \text{ iff } a'_i = 1$$

Б

$$a_i = \overline{1} \text{ iff } a'_i = \overline{1}$$

combinatorial result

faster exact algorithm

conclusion

Computing efficiently the tables

What remains is to find a method to compute $T_\ell\oplus P$

DP algorithm

combinatorial result

faster exact algorithm

conclusion

Computing efficiently the tables

What remains is to find a method to compute $T_\ell\oplus P$



DP algorithm

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conclusion

Computing efficiently the tables

Definition

$$A_w = \{ \vec{v} \mid w \cdot v \in A \}$$

$$egin{array}{rll} A\oplus B&=&0((A_0\cup A_{\overline{0}})\oplus B_0)\ &&\cup&1((A_1\cup A_{\overline{1}})\oplus B_0)\ &&\cup&\overline{1}(A_0\oplus B_1) \end{array}$$

introdu	uction		DP al	gorithm		comb	inatoria	al resul	t	fas	ter exac	t algorit	hm	c	conclusio	on
Co	mpu	iting	; efl	ficie	ntly	the	tal	bles	5							
									_							
					\oplus	0	0	1								
					0	0	0	1	1							
					1	$\overline{1}$	\sim	—	—							
\oplus	00	00	01	01	00	00	<u>0</u> 1	01	10	1 0	11	$1\overline{1}$	10	10	$\overline{1}1$	11
00 01																

10

introdu	uction		DP a	algorithm		combi	natoria	al resul	t	fast	er exact	t algorit	hm	C	onclusio	n
Co	mpı	uting	g e	fficie	ntly	the	tal	oles)							
					⊕ 0	$\left \begin{array}{c} 0 \\ 0 \\ \overline{1} \end{array} \right $	<mark>0</mark> 0		<mark>1</mark> 1							
					1	1	2	—	—							
\oplus	00	00	01	01	00	$\overline{00}$	01	01	10	1 0	11	$1\overline{1}$	10	$\overline{10}$	11	$\overline{11}$
00 01 10																
11																

introduction

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Computing efficiently the tables

\oplus	0	$\overline{0}$	1	1
		0	1	1
1	$\overline{1}$	\sim	_	_

for two adjacent vertices

for two adjacent vertices

		$0\overline{0}$	01	01	00	00	$\overline{0}1$	$\overline{01}$	1 0	10	11	$1\overline{1}$	1 0	10	$\overline{1}1$	$\overline{11}$
00																
01			_	—			—	—			_	—			_	—
10									_	_	_	_	—	_	—	_
11	_															

combinatorial result

faster exact algorithm

conclusion

Computing efficiently the tables

		$\overline{0}$		
0			1	1
1	$\overline{1}$	\sim		_

for two adjacent vertices

for two adjacent vertices

		$0\overline{0}$	01	01	$\overline{0}0$	$\overline{00}$	$\overline{0}1$	01	1 0	1 0	11	$1\overline{1}$	1 0	1 0	$\overline{1}1$	$\overline{11}$
00																
01		\sim	—	—		2	—	—		2	—	—		2	—	—
10					\sim	2	~	\sim	_	_	_	_	_	_	_	—
11	-															

DP algorithm

combinatorial result

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conclusion

Computing efficiently the tables

\oplus	0	$\overline{0}$	1	$\overline{1}$
0	0	0	1	1
1	$\overline{1}$	\sim	—	—

for two adjacent vertices

for two adjacent vertices

\oplus	00	$0\overline{0}$	01	01	$\overline{0}0$	$\overline{00}$	$\overline{0}1$	01	1 0	$1\overline{0}$	11	$1\overline{1}$	1 0	10	$\overline{1}1$	$\overline{11}$
00	00	00	01	01	00	00	01	01	10	10	11	11	10	10	11	-
01	$0\overline{1}$	2	_	_	$0\overline{1}$	2	_	_	$1\overline{1}$	2	_	_	$1\overline{1}$	2	_	_
10	$\overline{1}0$	$\overline{1}0$	$\overline{1}1$	$\overline{1}1$	\sim	\sim	\sim	\sim	_	—	—	_	_	_	—	

introd	uction

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Computing efficiently the tables

\oplus	0	$\overline{0}$	1	$\overline{1}$
0	0	0	1	1
1	$\overline{1}$	\sim	_	_

for two adjacent vertices

for two adjacent vertices

\oplus	00	$0\overline{0}$	01	01	$\overline{0}0$	$\overline{00}$	$\overline{0}1$	01	1 0	1 0	11	$1\overline{1}$	1 0	10	$\overline{1}1$	$\overline{11}$
00	00	00	01	01	00	00	01	01	10	10	11	11	10	10	11	-
01	$0\overline{1}$	2	_	_	$0\overline{1}$	\sim	—	_	$1\overline{1}$	\sim	_	_	$1\overline{1}$	2	_	_
10	$\overline{1}0$	$\overline{1}0$	$\overline{1}1$	$\overline{1}1$	2	2	\sim	\sim	_	_	_	_	_	_	_	

 \rightarrow Prefix $\overline{11}$ cannot appear.

introdu	iction		DP a	lgorithm		com	binator	ial result	t	fast	er exact	t algorit	hm	C	conclusio	on
Co	ՠբւ	utin	g ef	ficie	ently	' the	e ta	bles								
\oplus	00	$0\overline{0}$	01	01	00	$\overline{00}$	<u>0</u> 1	01	1 0	1 0	11	$1\overline{1}$	1 0	$\overline{10}$	$\overline{1}1$	$\overline{11}$
00	00	00	01	01	00	00	01	01	10	10	11	11	10	10	11	_
01	$0\overline{1}$	2	—	_	$0\overline{1}$	2	—	_	$1\overline{1}$	\sim	_	_	$1\overline{1}$	2	_	_
10	$\overline{1}0$	$\overline{1}0$	$\overline{1}1$	$\overline{1}1$	\sim	\sim	\sim	\sim	_	—	_	—	_	_	_	_
11																

 $A \oplus B =$

introdu	iction		DP al	gorithm		com	binatori	al result		fast	er exact	t algorit	hm	C	onclusio	on
Co	mpι	utin	g ef	ficie	ently	' the	e ta	bles								
\oplus	00	00	01	01	00	00	0 1	01	1 0	1 0	11	$1\overline{1}$	1 0	1 0	$\overline{1}1$	$\overline{11}$
00	00	00	01	01	00	00	01	01	10	10	11	11	10	10	11	-
01	$0\overline{1}$	2	_	_	$0\overline{1}$	2	—	—	$1\overline{1}$	2	—	_	$1\overline{1}$	2	—	_
10	$\overline{1}0$	$\overline{1}0$	$\overline{1}1$	$\overline{1}1$	2	\sim	\sim	2	_	_	_	_	_	_	_	_
11																

introdu	uction		DP al	gorithm		com	binator	ial result		fast	er exact	t algorit	hm	C	onclusio	n
Co	ՠբւ	uting	g ef	ficie	ently	' the	e ta	bles								
\oplus	00	00	01	01	00	00	0 1	01	10	1 0	11	$1\overline{1}$	1 0	10	$\overline{1}1$	$\overline{11}$
00	00	00	01	01	00	00	01	01	10	10	11	11	10	10	11	_
01	$0\overline{1}$	\sim	_	_	$0\overline{1}$	\sim	_	_	$1\overline{1}$	\sim	_	_	$1\overline{1}$	\sim	_	—
10	$\overline{1}0$	$\overline{1}0$	$\overline{1}1$	$\overline{1}1$	\sim	\sim	\sim	\sim	_	_	_	_	_	_	_	_
11																

 $A \oplus B = \overline{00((A_{00} \cup A_{0\overline{0}} \cup A_{\overline{00}} \cup \overline{A_{\overline{00}}}) \oplus B_{00})}$ $\cup 01((A_{01} \cup A_{0\overline{1}} \cup A_{\overline{01}} \cup A_{\overline{01}}) \oplus B_{00})$

introdu	iction		DP al	gorithm		com	binatori	al result		fast	er exact	t algorit	hm	C	onclusio	on
Co	ՠբւ	utin	g ef	ficie	ntly	' the	e ta	bles								_
\oplus	00	00	01	01	00	00	<u>0</u> 1	01	1 0	1 0	11	$1\overline{1}$	1 0	10	$\overline{1}1$	$\overline{11}$
00	00	00	01	01	00	00	01	01	10	10	11	11	10	10	11	_
01	$0\overline{1}$	\sim	_	_	$0\overline{1}$	\sim	_	_	$1\overline{1}$	\sim	_	_	$1\overline{1}$	\sim	_	—
10	$\overline{1}0$	$\overline{1}0$	$\overline{1}1$	$\overline{1}1$	\sim	\sim	\sim	\sim	_	_	_	_	_	_	_	_
11																

 $\begin{array}{rcl} A \oplus B = & 00((A_{00} \cup A_{0\overline{0}} \cup A_{\overline{00}} \cup A_{\overline{00}}) \oplus B_{00}) \\ & \cup & 01((A_{01} \cup A_{0\overline{1}} \cup A_{\overline{01}} \cup A_{\overline{01}}) \oplus B_{00}) \\ & \cup & 10((A_{10} \cup A_{\overline{10}} \cup A_{\overline{10}} \cup A_{\overline{10}}) \oplus B_{00}) \end{array}$

introdu	uction		DP al	gorithm		com	nbinator	ial result		fast	er exac	t algorit	hm	c	onclusio	on
Co	mpι	utin	g ef	ficie	ntly	' the	e ta	bles								_
\oplus	00	00	01	01	00	00	01	01	10	1 0	11	$1\overline{1}$	<u>1</u> 0	10	$\overline{1}1$	$\overline{11}$
00	00	00	01	01	00	00	01	01	10	10	11	11	10	10	11	_
01	$0\overline{1}$	\sim	_	_	$0\overline{1}$	\sim	_	_	$1\overline{1}$	\sim	_	_	$1\overline{1}$	\sim	_	—
10	$\overline{1}0$	$\overline{1}0$	$\overline{1}1$	$\overline{1}1$	\sim	\sim	\sim	\sim	_	_	_	_	_	_	_	_
11	-															

- $\cup \quad 01((A_{01}\cup A_{0\overline{1}}\cup A_{\overline{01}}\cup A_{\overline{01}})\oplus B_{00})$
- $\cup \quad 10((A_{10}\cup A_{\overline{10}}\cup A_{\overline{10}}\cup A_{\overline{10}})\oplus B_{00})$
- $\cup \quad 11((A_{11}\cup A_{1\overline{1}}\cup A_{\overline{11}})\oplus B_{00})$

introdu	uction		DP al	gorithm		com	ibinator	ial result		fast	er exact	t algorit	hm	C	onclusio	on
Co	mpι	uting	g ef	ficie	ently	' the	e ta	bles								_
\oplus	00	00	01	$0\overline{1}$	00	00	<u>0</u> 1	01	1 0	10	11	$1\overline{1}$	<u>1</u> 0	$\overline{10}$	$\overline{1}1$	$\overline{11}$
00	00	00	01	01	00	00	01	01	10	10	11	11	10	10	11	_
01	$0\overline{1}$	\sim	_	_	$0\overline{1}$	\sim	—	_	$1\overline{1}$	\sim	_	_	$1\overline{1}$	2	_	_
10	$\overline{1}0$	$\overline{1}0$	$\overline{1}1$	$\overline{1}1$	\sim	\sim	\sim	\sim	_	_	_	_	_	_	_	_
11	-															

- $\cup \quad 01((A_{01}\cup A_{0\overline{1}}\cup A_{\overline{01}}\cup A_{\overline{01}})\oplus B_{00})$
- $\cup \quad 10((A_{10}\cup A_{\overline{10}}\cup A_{\overline{10}}\cup A_{\overline{10}})\oplus B_{00})$
- $\cup \quad 11((A_{11}\cup A_{1\overline{1}}\cup A_{\overline{11}})\oplus B_{00})$
- $\cup \quad \overline{01}((A_{00} \cup A_{\overline{0}0}) \oplus \overline{B_{01}})$

introdu	uction		DP al	gorithm		com	ibinator	ial result		fast	er exact	t algorit	hm	C	onclusio	on
Co	ՠբւ	uting	g ef	ficie	ently	' the	e ta	bles								_
\oplus	00	00	01	$0\overline{1}$	00	00	<u>0</u> 1	01	1 0	10	11	$1\overline{1}$	<u>1</u> 0	$\overline{10}$	$\overline{1}1$	$\overline{11}$
00	00	00	01	01	00	00	01	01	10	10	11	11	10	10	11	_
01	$0\overline{1}$	\sim	_	_	$0\overline{1}$	\sim	—	_	$1\overline{1}$	\sim	_	_	$1\overline{1}$	2	_	_
10	$\overline{1}0$	$\overline{1}0$	$\overline{1}1$	$\overline{1}1$	\sim	\sim	\sim	\sim	_	_	_	_	_	_	_	_
11	-															

- $\cup \quad 01((A_{01}\cup A_{0\overline{1}}\cup A_{\overline{01}}\cup A_{\overline{01}})\oplus B_{00})$
- $\cup \quad 10((A_{10}\cup A_{\overline{10}}\cup A_{\overline{10}}\cup A_{\overline{10}})\oplus B_{00})$
- $\cup \quad 11((A_{\underline{11}}\cup A_{\underline{11}}\cup A_{\underline{11}})\oplus B_{00})$
- $\cup \quad 0\overline{1}((A_{00}\cup A_{\overline{0}0})\oplus B_{01})$
- $\cup \quad 1\overline{1}((A_{10}\cup A_{\overline{1}0})\oplus B_{01})$

introdu	iction		DP al	gorithm		com	ibinator	ial result		fast	er exac	t algorit	hm	C	onclusio	n
Co	ՠբւ	uting	g ef	ficie	ntly	' the	e ta	bles								_
\oplus	00	$0\overline{0}$	01	01	00	$\overline{00}$	0 1	01	1 0	1 0	11	$1\overline{1}$	<u>1</u> 0	$\overline{10}$	$\overline{1}1$	11
00	00	00	01	01	00	00	01	01	10	10	11	11	10	10	11	_
01	$0\overline{1}$	\sim	_	_	$0\overline{1}$	\sim	—	_	$1\overline{1}$	\sim	_	_	$1\overline{1}$	2	_	—
10	$\overline{1}0$	$\overline{1}0$	$\overline{1}1$	$\overline{1}1$	\sim	\sim	\sim	\sim	_	—	_	_	_	_	_	—
11	-															

 $A \oplus B = 00((A_{00} \cup A_{\overline{00}} \cup A_{\overline{00}} \cup A_{\overline{00}}) \oplus B_{00})$

- $\cup \quad 01((A_{01}\cup A_{0\overline{1}}\cup A_{\overline{01}}\cup A_{\overline{01}})\oplus B_{00})$
- $\cup \quad 10((A_{\underline{1}0}\cup A_{\underline{1}\overline{0}}\cup A_{\overline{1}\overline{0}}\cup A_{\overline{1}\overline{0}})\oplus B_{00})$
- $\cup \quad 11((A_{11}\cup A_{1\overline{1}}\cup A_{\overline{11}})\oplus B_{00})$
- $\cup \quad 0\overline{1}((A_{00}\cup A_{\overline{0}0})\oplus B_{01})$
- $\cup \quad 1\overline{1}((A_{10}\cup A_{\overline{1}0})\oplus B_{01})$
- $\cup \quad \overline{10}((A_{00}\cup A_{0\overline{0}})\oplus B_{10})$

introdu	iction		DP al	gorithm		com	ibinator	ial result		fast	er exac	t algorit	hm	C	onclusio	n
Co	ՠբւ	uting	g ef	ficie	ntly	' the	e ta	bles								_
\oplus	00	$0\overline{0}$	01	01	00	$\overline{00}$	0 1	01	1 0	1 0	11	$1\overline{1}$	<u>1</u> 0	$\overline{10}$	$\overline{1}1$	11
00	00	00	01	01	00	00	01	01	10	10	11	11	10	10	11	_
01	$0\overline{1}$	\sim	_	_	$0\overline{1}$	\sim	—	_	$1\overline{1}$	\sim	_	_	$1\overline{1}$	2	_	—
10	$\overline{1}0$	$\overline{1}0$	$\overline{1}1$	$\overline{1}1$	\sim	\sim	\sim	\sim	_	—	_	_	_	_	_	—
11	-															

 $A \oplus B = 00((A_{00} \cup A_{\overline{00}} \cup A_{\overline{00}} \cup A_{\overline{00}}) \oplus B_{00})$

- $\cup \quad 01((A_{01}\cup A_{0\overline{1}}\cup A_{\overline{01}}\cup A_{\overline{01}})\oplus B_{00})$
- $\cup \quad 10((A_{\underline{1}0}\cup A_{\underline{1}\overline{0}}\cup A_{\overline{1}\overline{0}}\cup A_{\overline{1}\overline{0}})\oplus B_{00})$
- $\cup \quad 11((A_{11}\cup A_{1\overline{1}}\cup A_{\overline{11}})\oplus B_{00})$
- $\cup \quad 0\overline{1}((A_{00} \cup A_{\overline{00}}) \oplus B_{01})$
- $\cup \quad 1\overline{1}((A_{10}\cup A_{\overline{1}0})\oplus B_{01})$
- $\cup \quad \overline{10}((A_{00}\cup A_{0\overline{0}})\oplus B_{10})$
- $\cup \quad \overline{11}((A_{01}\cup A_{0\overline{1}})\oplus B_{10})$

introdu	iction		DP al	gorithm		com	ibinator	ial result		fast	er exac	t algorit	hm	C	onclusio	on
Co	mpι	uting	g ef	ficie	ntly	' the	e ta	bles								_
\oplus	00	00	01	01	00	$\overline{00}$	01	01	1 0	1 0	11	$1\overline{1}$	<u>1</u> 0	$\overline{10}$	$\overline{1}1$	$\overline{11}$
00	00	00	01	01	00	00	01	01	10	10	11	11	10	10	11	_
01	$0\overline{1}$	\sim	_	_	$0\overline{1}$	\sim	_	_	$1\overline{1}$	\sim	_	_	$1\overline{1}$	\sim	_	—
10	$\overline{1}0$	$\overline{1}0$	$\overline{1}1$	$\overline{1}1$	\sim	\sim	\sim	\sim	_	_	_	_	_	_	_	_
11																

 $A \oplus B = 00((A_{00} \cup A_{\overline{00}} \cup A_{\overline{00}} \cup A_{\overline{00}}) \oplus B_{00})$

- $\cup \quad 01((A_{01}\cup A_{0\overline{1}}\cup A_{\overline{01}}\cup A_{\overline{01}})\oplus B_{00})$
- $\cup \quad 10((A_{10}\cup A_{\overline{10}}\cup A_{\overline{10}}\cup A_{\overline{10}})\oplus B_{00})$
- $\cup \quad 11((A_{11}\cup A_{1\overline{1}}\cup A_{\overline{11}})\oplus B_{00})$
- $\cup \quad 0\overline{1}((A_{00}\cup A_{\overline{0}0})\oplus B_{01})$
- $\cup \quad 1\overline{1}((A_{10}\cup A_{\overline{1}0})\oplus B_{01})$
- $\cup \quad \overline{10}((A_{00}\cup A_{0\overline{0}})\oplus B_{10})$
- $\cup \quad \overline{11}((A_{01}\cup A_{0\overline{1}})\oplus B_{10})$

Running-time : $T(n) = 8 \cdot T(n-2) = 8^{n/2} < 2.8285^n$

DP algorithm

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Decomposing the graph into connected subgraphs

What about using a \oplus -table for k' = O(1) vertices?

Imagine that a graph can be decomposed into some connected subsets of constant size k' ...

Decomposing the graph into connected subgraphs

Theorem (\star) Let *G* be a connected graph of order *n*. Let k < n be a positive integer.

Then there exist connected subgraphs G_1, G_2, \ldots, G_q of G s.t.

- (i) every vertex of G belongs to at least one of them
- (ii) the order of each of $G_1, G_2, \ldots, G_{q-1}$ is at least k and at most 2k (while for G_q we only require $|V(G_q)| \le 2k$)
- (iii) the sum of the numbers of vertices of $G'_i s$ is at most $n(1+\frac{1}{k})$

Decomposing the graph into connected subgraphs

Proof

1/2

- Consider a DFS-tree T of G rooted at r.
- For every v let T(v) be the subtree rooted in v.
- If $|T(r)| \le 2k$ then add G to the set of desired subgraphs and stop.
- If there is a vertex v such that k ≤ |T(v)| ≤ 2k then add G[V(T(v))] to the set of desired subgraphs and proceed recursively with G \ V(T(v)).

Proof

Decomposing the graph into connected subgraphs

- 2/2
- Otherwise there must be a vertex v such that |T(v)| > 2k and for its every child u, |T(u)| < k.</p>

In such a case find a subset $\{u_1, \ldots, u_i\}$ of children of v such that $k-1 \leq |T(u_1)| + \cdots + |T(u_i)| \leq 2k-1$.

Add $G[\{v\} \cup V(T(u_1)) \cup \cdots \cup V(T(u_i))]$ to the set of desired subgraphs and proceed recursively with $G \setminus (V(T(u_1)) \cup \cdots \cup V(T(u_i))).$

► This procedure terminates after at most ⁿ/_k steps and in each of them we have left at most one vertex of the identified connected subgraph in the further processed graph.

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An exact algorithm

Let $A \subseteq \{0, \overline{0}, 1, \overline{1}\}^n$ and $B \subseteq \{0, 1\}^n$ where n > k'. We compute $A \oplus B$ is the following way :

$$A \oplus B = \bigcup_{\substack{\vec{u} \in \{0, \bar{0}, 1, \bar{1}\}^{k'} \\ \vec{v} \in \{0, 1\}^{k'} \\ \text{s.t. } \vec{u} \oplus \vec{v} \text{ is defined}}} (\vec{u} \oplus \vec{v}) (A_{\vec{u}} \oplus B_{\vec{v}})$$
$$= \bigcup_{\substack{\vec{v} \in \{0, 1\}^{k'} \\ \vec{w} \in I \cap 1}} \left[\left(\bigcup_{\substack{\vec{u} \in \{0, \bar{0}, 1, \bar{1}\}^{k'} \\ \vec{v} \in \vec{v} = \vec{w}}} A_{\vec{u}} \right) \oplus B_{\vec{v}} \right]$$

Remark :

Computation can be omitted whenever $\left(\bigcup_{\vec{u} \in \{0, \vec{0}, 1, \vec{1}\}^{k'}} A_{\vec{u}}\right)$ is empty. s.t. $\vec{u} \oplus \vec{v} = \vec{w}$

How many pairs \vec{v}, \vec{w} are there s.t. there is at least one \vec{u} with $\vec{u} \oplus \vec{v} = \vec{w}$?

If \vec{v} is fixed, then $v_i = 1 \Rightarrow w_i = \overline{1}$.

Thus, for a fixed \vec{v} there are at most $2^{k'-||\vec{v}||}$ many \vec{w} 's, where $||\vec{v}||$ denotes the number of positions *i* such that $v_i = 1$.

The total number of pairs \vec{v}, \vec{w} such that $\vec{w} = \vec{v} \oplus \vec{u}$ for some \vec{u} is therefore at most

$$\sum_{ec{v} \in \{0,1\}^{k'}} 2^{k' - ||ec{v}||} \le pp(k')$$

How many pairs \vec{v}, \vec{w} are there s.t. there is at least one \vec{u} with $\vec{u} \oplus \vec{v} = \vec{w}$?

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The total number of pairs \vec{v}, \vec{w} such that $\vec{w} = \vec{v} \oplus \vec{u}$ for some \vec{u} is therefore at most

$$\sum_{ec{v} \in \{0,1\}^{k'}} 2^{k' - ||ec{v}||} \le pp(k')$$

How many pairs \vec{v}, \vec{w} are there s.t. there is at least one \vec{u} with $\vec{u} \oplus \vec{v} = \vec{w}$?

If \vec{v} is fixed, then $v_i = 1 \Rightarrow w_i = \overline{1}$.

Thus, for a fixed \vec{v} there are at most $2^{k'-||\vec{v}||}$ many \vec{w} 's, where $||\vec{v}||$ denotes the number of positions *i* such that $v_i = 1$.

The total number of pairs \vec{v}, \vec{w} such that $\vec{w} = \vec{v} \oplus \vec{u}$ for some \vec{u} is therefore at most

$$\sum_{ec{ au} \in \{0,1\}^{k'}} 2^{k' - ||ec{ au}||} \le pp(k')$$

How many pairs \vec{v}, \vec{w} are there s.t. there is at least one \vec{u} with $\vec{u} \oplus \vec{v} = \vec{w}$?

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By Theorem (*), the total length of the vectors is $n' \le n(1+1/k)$. In each recursive computation :

- Prepare up to pp(k') many pairs of sets of vectors of length n' k'
- ▶ Recursively compute ⊕ on these pairs
- From the result, compute $T_{\ell+1}$ in linear time
- The size of B is at most $O(n2^{n'})$ bits
- ► The size of A is at most O(npp(n')) bits : the 1̄'s form a 2-packing and there are only two possibilities (1 or 0/0̄) for the other nodes.

Thus the running-time is given by

 $T(n) \leq O(n \cdot pp(n') + pp(k') \cdot T(n' - k'))$

where $k \leq k' \leq 2k$.

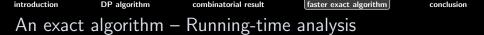
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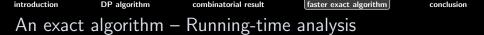
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Choosing constant k big enough :

 $T(n) = O(2.6488^n)$



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Conclusion

1 Definitions and Known Results

2 A (Simple) Dynamic Programming Based Algorithm

3 A Combinatorial Result

(4) A Faster Exact Exponential-Time Algorithm

5 Conclusion

Conclusion

Combinatorial result : number of proper pairs $2.6117^n \le pp(n) \le 2.6488^n$



Exact exponential-time algorithm for L(2, 1)-labelings $O(2.6488^{n})$

Interesting questions :

- Does inclusion/exclusion or subset convolution can achieve a $O(2^n)$ -time algorithm?
- ▶ Is it possible to find a 2-approx in $O(c^n)$ with c < 2?
- ▶ In [GY92], it is conjectured that $\lambda(G) < \Delta(G)^2$. It is still not fully resolved. It has been proved for graphs of large maximum degree [HRS08].

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Merci!





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