

Tight bounds for counting colorings parameterised by cutwidth via matrix ranks

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(slides by Carla Groenland and Isja Mannens)
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Joint work with Isja Mannens, Jesper Nederlof and Krisztina Szilágyi
STACS 2022

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- Suppose it divides global solutions in families A, B of partial solutions
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[CLN'18] $rk(M_{match}^k) \geq 4^k \rightarrow$ tight lower bound for #HAM CYC/pw.

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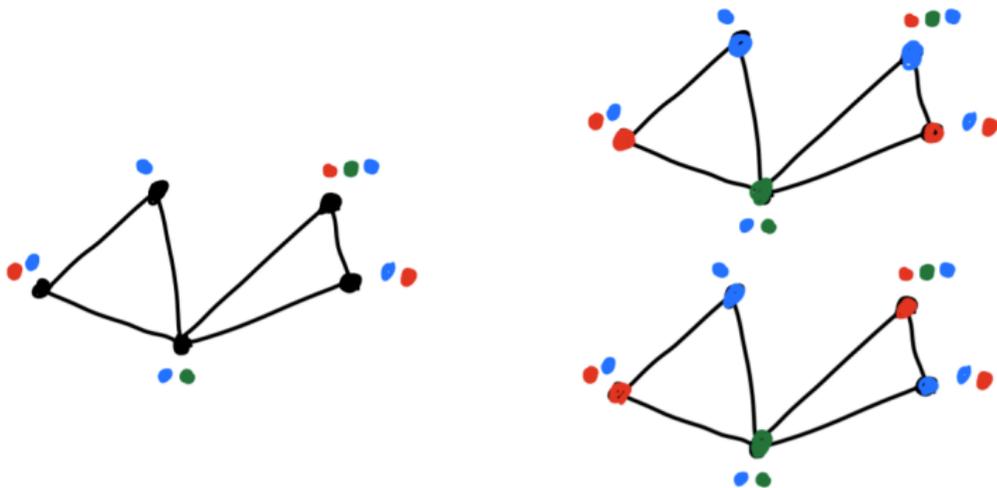
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q -list coloring

Given: Graph $G = (V, E)$ and for each $v \in V$, a list $L(v) \subseteq \{1, \dots, q\}$.

Want: $c(v) \in L(v)$ for all $v \in V$ such that $c(u) \neq c(v)$ for $uv \in E$.



q -LIST COL MOD p

Given. G graph, lists $L(v) \subseteq \{1, \dots, q\}$ for all $v \in V(G)$, $k \in \mathbb{F}_p$.

Output. What is the number of list colorings of G modulo p ?

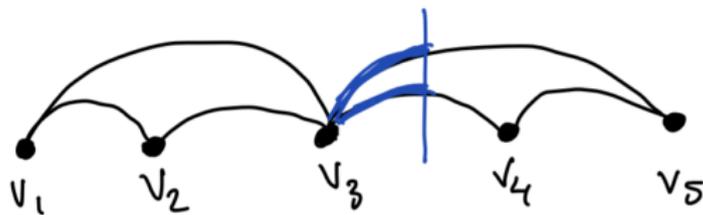
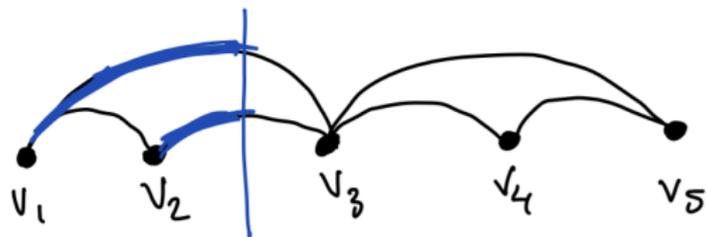
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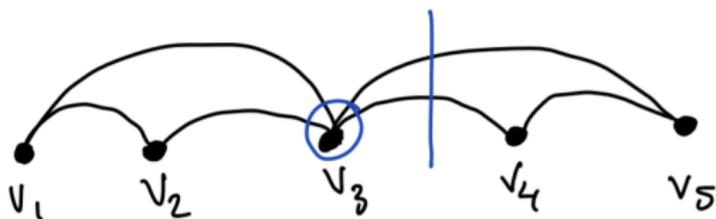
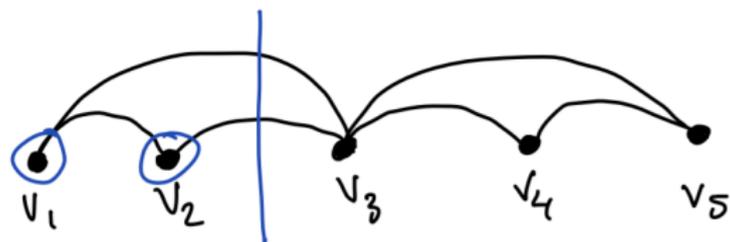
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Cutwidth



$\max_{\sigma} \min_i \# \text{ edges crossing } i\text{th cut}$

Pathwidth



$-1 + \max_{\sigma} \min_i \# \text{ left endpoints of edges crossing } i\text{th cut}$

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Strong Exponential Time Hypothesis (SETH)

For all $\epsilon > 0$, there is some $k \geq 3$ such that k -SAT cannot be solved in time $O((2 - \epsilon)^n)$.

q -coloring parameterised by width

q -COL: does a given graph admit a q -coloring?

Treewidth. Dynamic programming: $q^{\text{tw}} \text{poly}(n)$.

Also extends to $\#q$ -LIST COL.

No $(q - \epsilon)^{\text{tw}} \text{poly}(n)$ algorithm under SETH.

(Lokshtanov, Marx and Saurabh, 2011)

Cutwidth. Randomised algorithm: $2^{\text{ctw}} \text{poly}(n)$.

Does not extend to $\#q$ -LIST COL.

No $(2 - \epsilon)^{\text{ctw}} \text{poly}(n)$ algorithm under SETH.

(Nederlof, Jansen, 2018)

Present paper builds on some insights from this paper

Our result

For n -vertex graphs of cutwidth ctw , there is an algorithm running in time

$$\begin{cases} q^{ctw} \text{poly}(n) & \text{if } p \text{ does not divide } q - 1, \\ (q - 1)^{ctw} \text{poly}(n) & \text{if } p \text{ divides } q - 1. \end{cases}$$

Furthermore, under SETH there is no $(q - \epsilon)^{ctw} \text{poly}(n)$ resp. $(q - 1 - \epsilon)^{ctw} \text{poly}(n)$ algorithms in these cases.

Coloring compatibility matrix

$\text{col}(X)$ = set of list q -colorings of $G[X]$.

Bipartite graph (X, Y, E) , $x \in \text{col}(X)$ and $y \in \text{col}(Y)$.

Compatible: $x \sim y$ if $x(u) \neq y(v)$ for all $uv \in E$.

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$\text{rk}_p(M)$ depends on whether p divides $q - 1$.

Rank of coloring compatibility matrix

Bipartite graph (X, Y, E) , for $x \in \text{col}(X)$ and $y \in \text{col}(Y)$ q -colorings

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$rk_p(M) \leq (q - 1)^{|E|}$ if p divides $q - 1$ and $\leq q^{|E|}$ otherwise.

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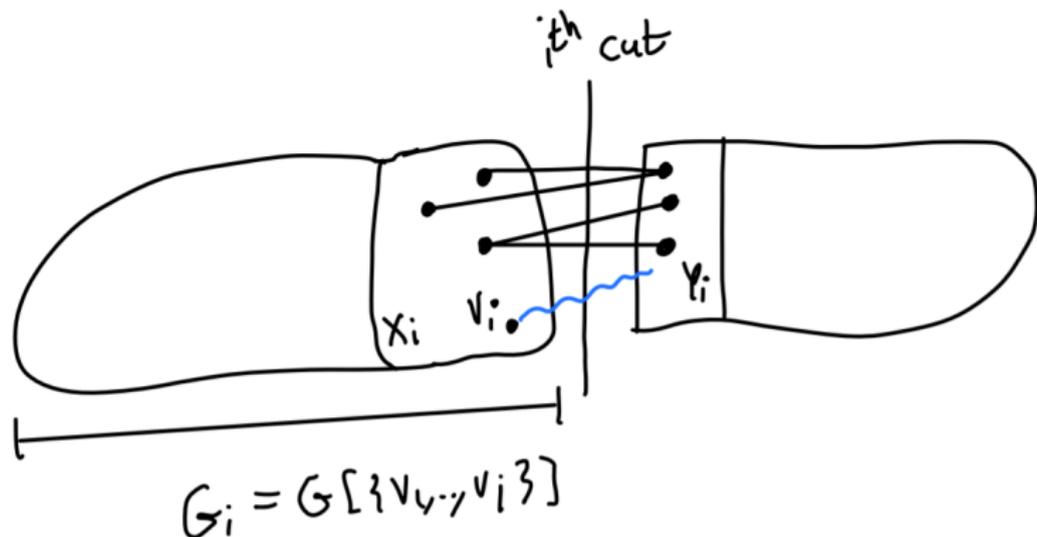
$\text{rk}_p(M) \leq (q-1)^{|E|}$ if p divides $q-1$ and $\leq q^{|E|}$ otherwise.

$$\begin{array}{l} x(u) = 1 \\ x(u) = 2 \\ x(u) = 3 \\ x(u) = 4 \end{array} \begin{pmatrix} y(v) = 1 & y(v) = 2 & y(v) = 3 & y(v) = 4 \\ 0 & 1 & 1 & 1 \\ 1 & 0 & 1 & 1 \\ 1 & 1 & 0 & 1 \\ 1 & 1 & 1 & 0 \end{pmatrix}$$

Cutwidth order v_1, \dots, v_n .

$G_i = G[\{v_1, \dots, v_i\}]$, the i th cut gives a bipartite graph (L_i, R_i) .

$X_i = L_i \cup \{v_i\}$, $Y_i = N[X_i] = R_i$.



Dynamic programming

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For $x \in \text{col}(X_i)$,

$$\begin{aligned} T_i[x] &= \text{number of extensions of } x \text{ to } G_i \\ &= \text{number of } c \in \text{col}(G_i - X_i) \text{ with } c \sim x \\ &= \sum_{c \in \text{col}(G_i - X_i): c \sim x} T_{i-1}[z|_{X_{i-1}}]. \end{aligned}$$

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Table size $|\text{col}(X_i)| = q^{|X_i|} \leq q^{\text{ctw}+1}$.

Idea. Compute smaller table 'representative' of bigger table.

Coloring compatibility matrix M_i of bipartite graph on between X_i and Y_i .
For $y \in \text{col}(Y_i)$,

$$\sum_{x \in \text{col}(X_i)} T_i[x] M_i[x, y] = (T_i^t M_i)[y]$$

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T'_i is M_i -**representative** for T_i if $T_i^t M_i \equiv_p (T'_i)^t M_i$.

Reducing the table size

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If we know $T'_i[x] = 0$, we do not need to compute it.

Linear algebra \implies there exists such T'_i with $|\text{supp}(T'_i)| \leq \text{rank}_p(M_i)$.

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1 : a		0	1	1		1 : $a + c$		0	1	1
2 : b		1	0	1		2 : $b + c$		1	0	1
3 : c		1	1	0		3 : 0		1	1	0

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- Can we efficiently compute T'_i ?
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Maintain $R \subseteq X_i$ such that $T'_i[x] = 0$ if $x(v) = q$ for some $v \in R$.

T'_i fully reduced if $\{v \in X_i : \deg(v) = 1\} \subseteq R$.

Number of $x : X \rightarrow \{1, \dots, q\}$ with $x(v) \neq q$ for all $v \in R$ is at most

$$(q - 1)^{|R|} q^{|X| - |R|}.$$

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For each x we have a table entry $T_i[x]$.

Since $|X \setminus R| \leq \frac{1}{2}(ctw - |R|)$, this is at most $(q - 1)^{ctw}$.

Algorithm. Initialise for $i = 1$. For $i = 2, \dots, n$,

- Ensure T'_{i-1} fully reduced.
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Lemma 1. If $v \in X_i$ degree 1, then can compute T' representative for T with set of reduced vertices $R \cup \{v\}$ in time $O((q-1)^{|R|} q^{|X_i|-|R|})$.

(Proof of Lemma 1 uses p divides $q-1$.)

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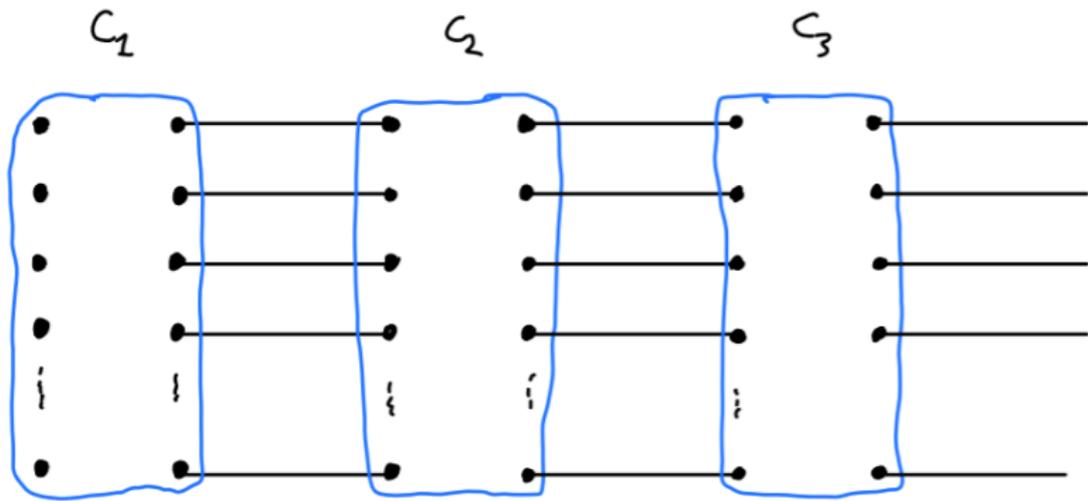
Lemma 2. If T'_{i-1} representative for T_{i-1} and fully reduced, then can compute T'_i representative for T_i in time $O((q-1)^{ctw})$.

Under SETH, $\#\text{CSP}(q, r) \bmod p$ cannot be solved in $(q - \epsilon)^n \text{poly}(n, m)$ for some r [Lampis, '20 (and others?)]

n variables, m constraints

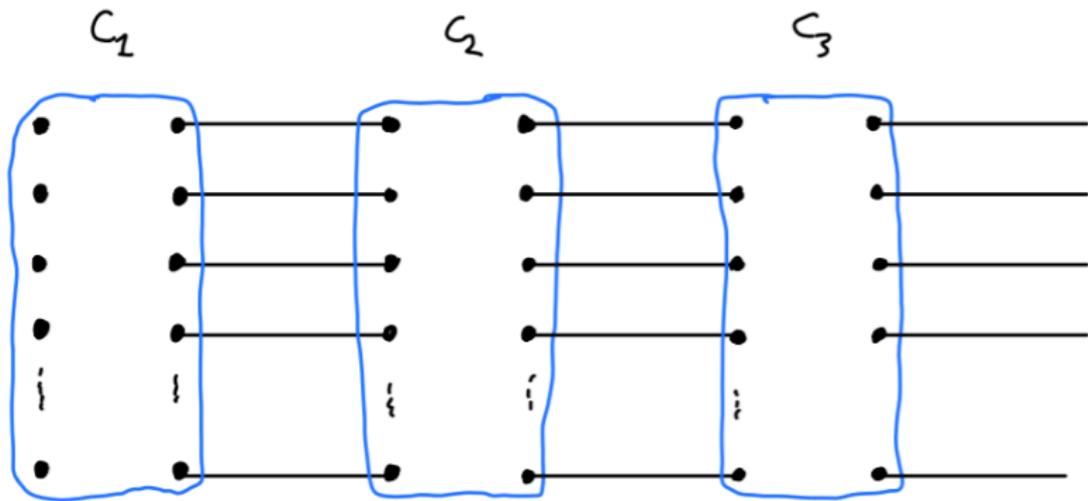
Constraints $\{1, \dots, q\}^r \rightarrow \{0, 1\}$ depend on at most r variables.

Count number of satisfying assignments mod p .

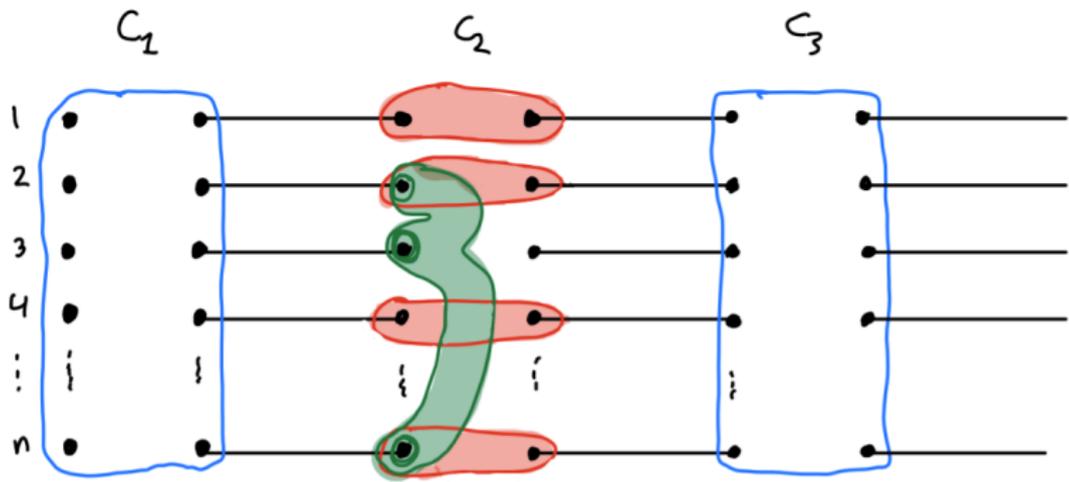


One column per constraint; one row per variable.

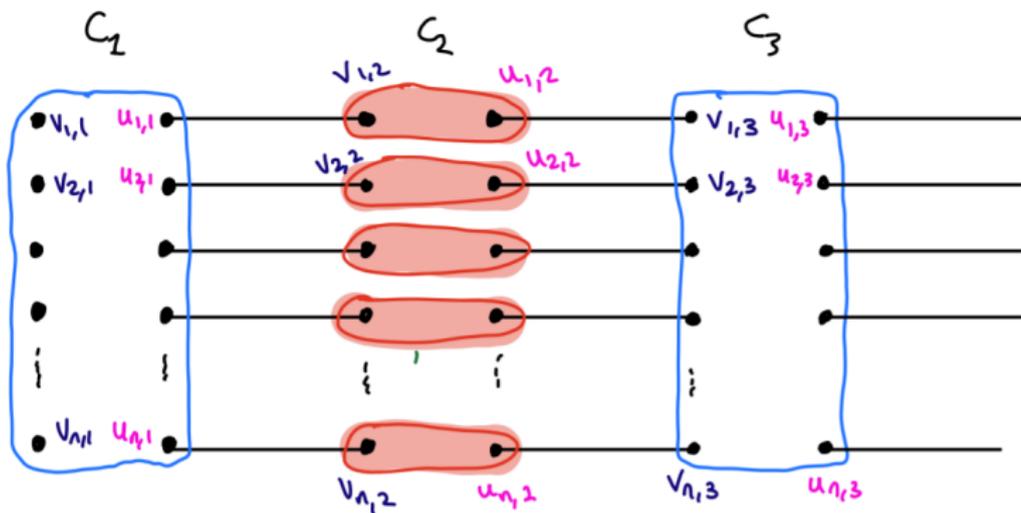
Cutwidth = $n + O(1)$, number of vertices = $\text{poly}(n, m)$.



Want: number of q -colorings equals number of satisfying assignments.
 "Identify a q -coloring with an assignment; check and copy."



Constraint 2 depends on variables 2, 3, n



Colour of $v_{i,2}$ = colour of $v_{i,3}$ for all i

Exploiting invertibility



Fix c_2, c_3 colorings of $v_{1,2}$ and $v_{1,3}$ respectively. The number of extensions of these colorings to the red graph equals

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$$\sum_{c'_2} f[c_2, c'_2] M[c'_2, c_3] = fM = M^{-1}M = I,$$

if we can set $f[c_2, c'_2] = M^{-1}[c'_2, c_2]$.

Counting connected edge sets

Given $G = (V, E)$, how many $X \subseteq E(G)$ are there for which (V, X) is connected?

Tutte polynomial can link this to number of q -colorings.

\implies under SETH there is no $(p - \epsilon)^{ctw} \text{poly}(n)$ algorithm.

'Correct' running times: $p^{ctw} \text{poly}(n)$, $p^{pw} \text{poly}(n)$, $p^{tw} \text{poly}(n)$.

Summary

Running time for counting modulo p the number of q -list colorings of n -vertex graphs of cutwidth ctw is

$$\begin{cases} q^{ctw} \text{poly}(n) & \text{if } p \text{ does not divide } q - 1, \\ (q - 1)^{ctw} \text{poly}(n) & \text{if } p \text{ divides } q - 1. \end{cases}$$

The rank over \mathbb{F}_p of the matrix $J_q - I_q$ (zeros on the diagonal, ones everywhere else) is

$$\begin{cases} q & \text{if } p \text{ does not divide } q - 1, \\ q - 1 & \text{if } p \text{ divides } q - 1. \end{cases}$$