

Degree Sequence Optimization and Sparse Integer Programming

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Given a graph H and a function $f_i : \{0, 1, \dots, d_i(H)\} \rightarrow \mathbb{Z}$
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Example: with $f_i(x) = (x-1)^2$ for all i can decide if H has a perfect matching

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NP-hard: H has a nonempty cubic subgraph if and only if for some i the optimal value of DSO with $f_i(x) = (x-3)^2$ and $f_j(x) = x(x-3)^2$ for all $j \neq i$ is zero

Some Special Cases

Complete Graphs

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This answers a **30 year long open question** (**Colbourn et al. 1986**)

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Proof: Involved dynamic programming with states being 7-tuples

Convex Functions and General Factors

The general **Factor** problem is:

Given a graph H and a subset B_i of $\{0, 1, \dots, d_i(H)\}$ for each i ,
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Reduces to DSO with $f_i(x) := 0$ for x in B_i and $f_i(x) := 1$ for x not in B_i

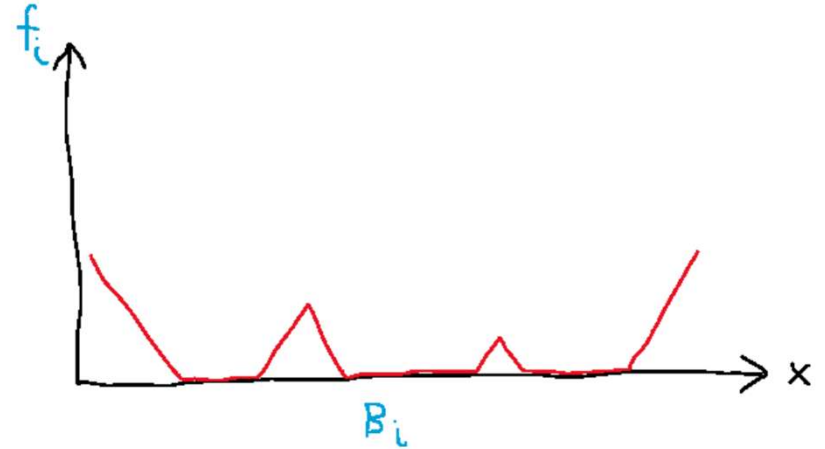
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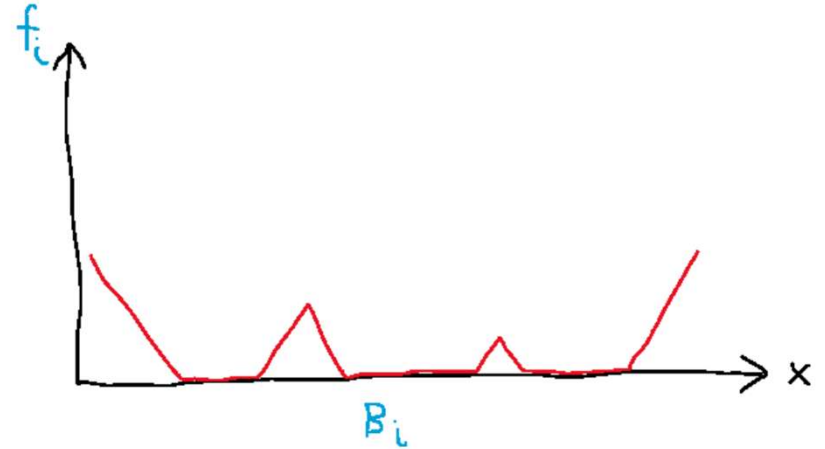
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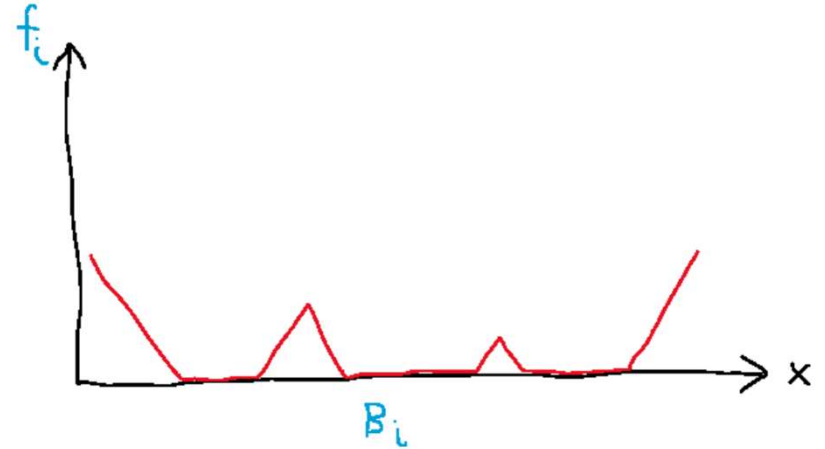
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Proof: Reduces to **weighted matching** on a suitable **larger graph**

Bounded Tree Width or Depth

Theorem (Onn 2022) For any fixed k , can solve DSO in polynomial time for any functions f_i and any graph H with $\text{tw}(H) \leq k$

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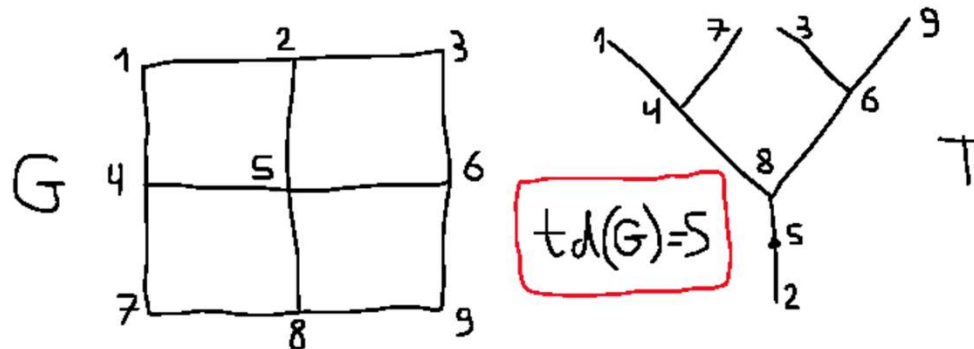
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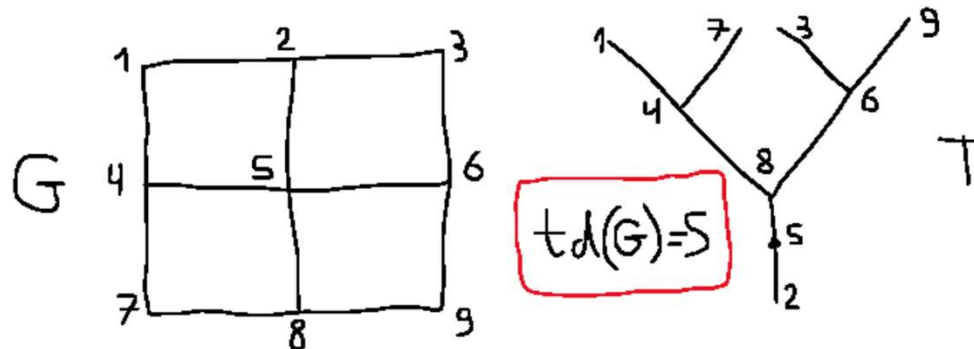


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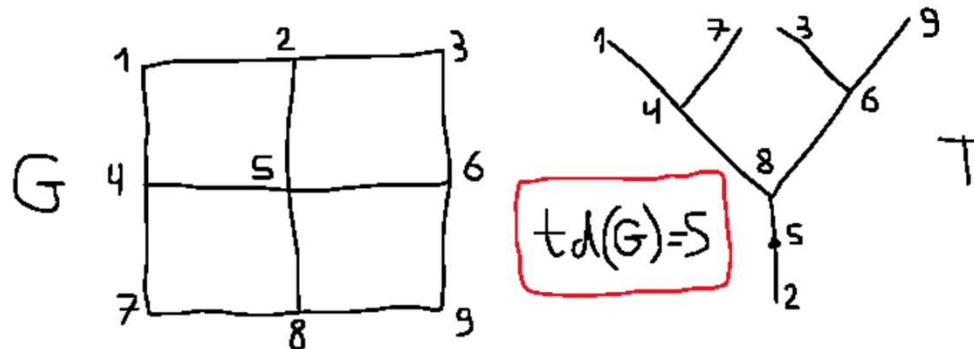
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The **tree-depth** of G is the **smallest** height of a rooted tree valid for G

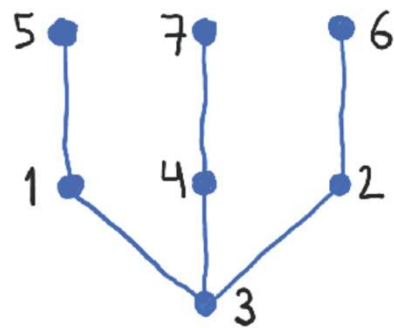


Sparse Integer Programming

Tree-Depth of a Matrix

The graph of $m \times n$ matrix A is the graph $G(A)$ on $[n]$ with j, k an edge iff $A_{i,j}, A_{i,k}$ are nonzero for some i . The tree-depth of A is $td(A) := td(G(A))$

$$A = \begin{matrix} & \begin{matrix} 1 & 2 & 3 & 4 & 5 & 6 & 7 \end{matrix} \\ \begin{matrix} \times & 0 & \times & 0 & \times & 0 & 0 \\ 0 & 0 & \times & \times & 0 & 0 & \times \\ 0 & \times & \times & 0 & 0 & \times & 0 \end{matrix} \end{matrix}$$



$$td(A) = 3$$

Sparse Integer Programming

$$\text{IP: } \min \{ cx : Ax = b, l \leq x \leq u, x \in \mathbb{Z}^n \}$$

A integer $m \times n$ matrix, l, u, c in \mathbb{Z}^n , b in \mathbb{Z}^m

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Specifically, there are functions $g(\alpha, d)$, $h(d)$ so IP is solvable in time:

$$g(\alpha, d) \text{poly}(n) \text{ when } d = \text{td}(A)$$

$$(\alpha + 1)^{h(d)} \text{poly}(n) \text{ when } d = \text{td}(A^T)$$

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Koutecky-Levin-Onn 2018, Eisenbrand-Hunkenschroder-Klein-Koutecky-Levin-Onn 2019

Koutecky-Onn 2021, "Sparse Integer Programming is FPT", Bulletin EATCS

One Application: N-Fold and Block Shaped IP

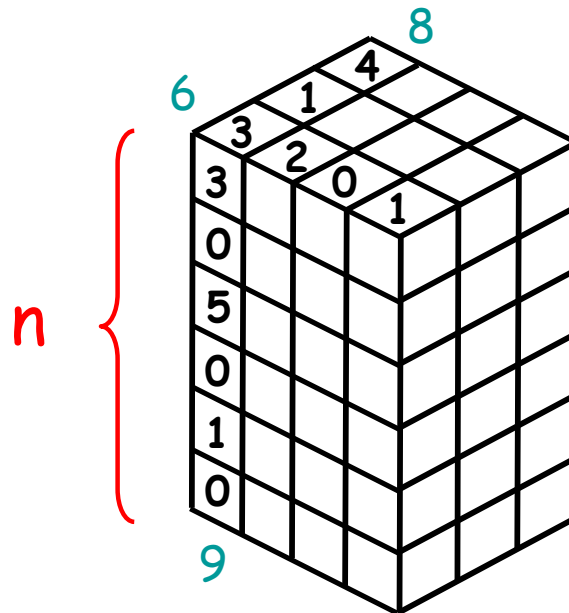
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$$A = \begin{pmatrix} A_{1,1} & A_{1,2} & A_{1,3} & A_{1,4} & A_{1,5} & A_{1,6} & A_{1,7} & A_{1,8} \\ A_{2,1} & A_{2,2} & A_{2,3} & A_{2,4} & A_{2,5} & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & A_{2,6} & A_{2,7} & A_{2,8} \\ A_{3,1} & A_{3,2} & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & A_{3,3} & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & A_{3,4} & A_{3,5} & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & A_{3,6} & A_{3,7} & A_{3,8} \\ A_{4,1} & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & A_{4,2} & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & A_{4,3} & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & A_{4,4} & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & A_{4,5} & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & A_{4,6} & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & A_{4,7} & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & A_{4,8} \end{pmatrix}$$

Corollary: Block shaped IP and N-fold IP are solvable in FPT time

In Particular: Multiway Tables

Optimization in **FPT time** over $m_1 \times \dots \times m_k \times n$ tables with given **margins** (and **multicommodity flows**), where this line started by **De Loera - Onn**:



Proof Sketch

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3. It can be shown that few Graver-best steps suffice to reach optimum

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Proof: The matrix A of the following IP has parameters $a=n-1$ and $d=\text{td}(A^T)=\text{td}(H)+1$ so solvable in polynomial time $n^{h(k+1)}\text{poly}(n)$:

$$\begin{aligned} \min \quad & \sum_{i=1}^n \sum_{j=0}^{d_i(H)} f_i(j) y_{ij} \\ \sum_{e \in \delta_i(H)} x_e - \sum_{j=0}^{d_i(H)} j y_{ij} &= 0 \\ \sum_{j=0}^{d_i(H)} y_{ij} &= 1 \\ 0 \leq x_e, y_{ij} \leq 1, & \text{ integer} \end{aligned}$$

Colored Degree Sequence Optimization

Adding suitable constraints to this IP, we can even solve the **colored** version of **DSO**, where the edges are **colored** by **p** colors and we need to find a subgraph having prescribed number of edges of each color:

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A special case of this problem is the notorious **exact matching problem** for which a **randomized** algorithm is known but not a **deterministic** one