

# Mind the Gap.

## Doubling Constant Parametrization of Weighted Problems: TSP, Max-Cut, and More

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# Disclaimer

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- I'm doing a PhD in *database systems* (3rd year).
- Focus: making databases robust with theory.
  - Query speedups on  Microsoft's Fabric via join size lower bounds.
  - Using  $\ell_p$ -norms, reverse inner-product inequalities - happy to discuss!

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  - Query speedups on  Microsoft's Fabric via join size lower bounds.
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- Submitting in TCS whenever I see an (exciting) result missing.
  - Latest: Approximate min-sum subset convolution [WAOA'24].
- Happy for feedback in terms of TCS terminology!

## Algebraic Improvements

- Björklund: Hamiltonicity in time  $O^*(1.66^n)$  [FOCS 2010].
- "The 4 Scandinavians": Fast Subset Convolution in time  $O^*(2^n)$  [STOC 2007].
- Williams: Max-Cut in time  $O^*(2^{\omega n/3})$  [ICALP 2004].
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# My Motivation

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## "Small Weights"

- (Almost) every single paper has an extra theorem to support *small weights* instances.

## "Small Weights" (cont'ed)

We also note that our algorithm can be used to solve TSP with integer weights via self-reducibility at the cost of a runtime blow-up by roughly a factor of the sum of all edges' weights.

**Theorem 3** *There is a Monte Carlo algorithm finding the weight of the lightest TSP tour in a positive integer edge weighted graph on  $n$  vertices in  $O^*(w1.657^n)$  time, where  $w$  is the sum of all weights, with error probability exponentially small in  $n$ .*

**Figure 1:** Björklund: How to solve TSP via Hamiltonicity.

## "Small Weights" (cont'ed)

THEOREM 3. *The subset convolution over the integer max-sum (min-sum) semiring can be computed in  $\tilde{O}(2^n M)$  time, provided that the range of the input functions is  $\{-M, -M + 1, \dots, M\}$ .*

**Figure 2:** The four Scandinavians: How to solve Min-Sum Subset Convolution.

# My Motivation: What Everyone Is Using

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## The "Small Weights" Case

- Solve the problem over the  $(\min, +)$  semiring by embedding the input weights into the polynomial ring.
- Run the new algorithm + extract the coefficients & exponents of the solution polynomial.

# My Motivation: What Everyone Is Using

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→ Downside: Pseudo-polynomial factors in the running time.

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## The "Small Weights" Case

- Solve the problem over the  $(\min, +)$  semiring by embedding the input weights into the polynomial ring.
- Run the new algorithm + extract the coefficients & exponents of the solution polynomial.

→ Downside: Pseudo-polynomial factors in the running time.

*Nevertheless*, if the input weights are **polynomially bounded**, the running time for the weighted problem = unweighted problem (under  $O^*$ -notation).

*Q: Are polynomially bounded weights the **only** case where algebraic algorithms avoid the pseudo-polynomial overhead?*

## My Motivation: Research Question

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*Q: Are polynomially bounded weights the **only** case where algebraic algorithms avoid the pseudo-polynomial overhead?*

*A: Apparently **not**.*

## Main Result (informal)

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*If the input weights have **small doubling**, then we can avoid the pseudo-polynomial overhead.*

Additive combinatorics got pretty popular in the last decade.

## **Hype: Some Results**

- Clustered 3SUM via the (constructive) BSG theorem [CL, STOC'15].
- 3SUM on Sidon sets [JX, STOC'23].
- Subset Sum & Integer Programs under small doubling via the constructive Freiman's theorem [RW, ESA'24].
- Shaving a  $O(\sqrt{n})$ -time factor for Subset Sum [BFN, SODA'25].

## Small Doubling

**Sumset.** For a set  $A$ , define

$$A + A = \{a + b \mid a, b \in A\}$$

**Doubling constant.**

$$C(A) = \frac{|A + A|}{|A|}$$

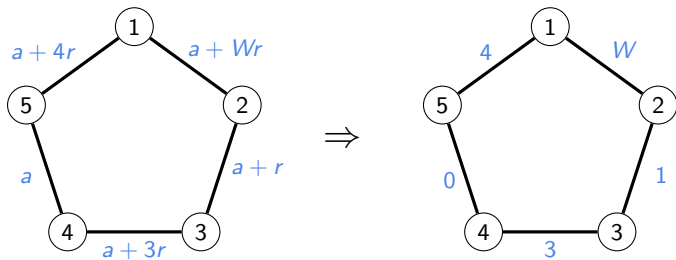
**Example.**

$$A = \{2, 4, 6, 8\} \quad A + A = \{4, 6, 8, 10, 12, 14, 16\}$$

$$\Rightarrow C(A) = 7/4 \approx 2.$$

**Key intuition:** small doubling  $\Rightarrow$  strong *additive structure*.

## Intuition: TSP with Arithmetic Progressions



Original weights  $w \in [0, a + Wr]$

New weights  $w' \in [0, W]$

$$w' = \frac{w - a}{r}$$

Recover original cost:  $r \cdot \text{cost}' + n \cdot a$ .

# Beyond Arithmetic Progressions: GAPs

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## Definition.

$$G = \{x_1 l_1 + \dots + x_d l_d \mid l_i \in [0, L_i]\}$$

- $x_1, \dots, x_d$ : generators (step sizes).
- $d$ : dimension.
- $L_1, \dots, L_d$ : GAP bounds.

## Example.

$$G = \{2l_1 + 5l_2 \mid l_1 \in [0, 3], l_2 \in [0, 2]\}$$

**Special case:**  $d = 1$  gives a simple AP.

## Small Doubling $\rightarrow$ GAP

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**Freiman's Theorem [Fre73].** If a finite set  $A$  satisfies

$$|A + A| \leq C|A|,$$

then  $A$  is contained in a GAP

$$G = \{x_1 l_1 + \cdots + x_d l_d \mid l_i \in [0, L_i]\}$$

of dimension  $d = d(C)$  and size  $|G| \leq v(C)|A|$ .

Both  $d(C)$  and  $v(C)$  depend only on  $C$ .

# Constructive Freiman's Theorem

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## Question

- Can we find the parameters of the GAP?

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**Constructive Freiman's Theorem [RW24].** Let  $A$  be a set of  $n$  integers with  $|A + A| \leq C|A|$ . Then, there exists an  $\tilde{O}_C(n)$ -time algorithm that, with probability  $1 - n^{-\gamma}$  for an arbitrarily large constant  $\gamma > 0$ , returns a GAP

$$G = \{x_1 \ell_1 + x_2 \ell_2 + \cdots + x_{d(C)} \ell_{d(C)} \mid \forall i, \ell_i \in [L_i]\} \supseteq A$$

with dimension  $d(C)$  and volume  $v(C)|A|$ .

# Meta Algorithm: Key Idea

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## Sketch (instantiated for TSP)

Let  $A$  be the input weights.

→ All possible solutions are contained in  $nA = A + \dots + A$  ( $n$  times).

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3. Encode the weights  $w'$  via  $G'$ .
  - Note:  $|G'|$  is polynomial w.r.t.  $O_C$  since  $|G'| \leq n^{d(C)} v(C) |A|$ .

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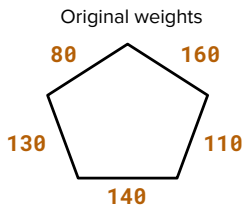
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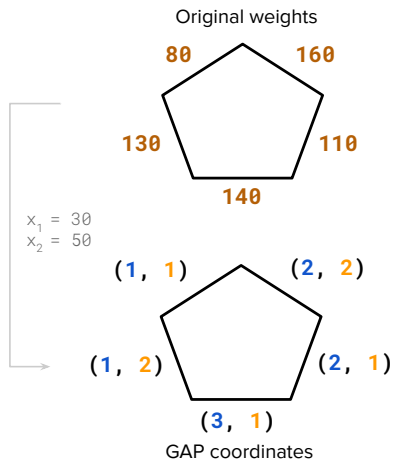
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4. Solve the problem on  $w'$  with the algebraic algorithm.
5. Decode the solution.

## Key Idea: Visualization

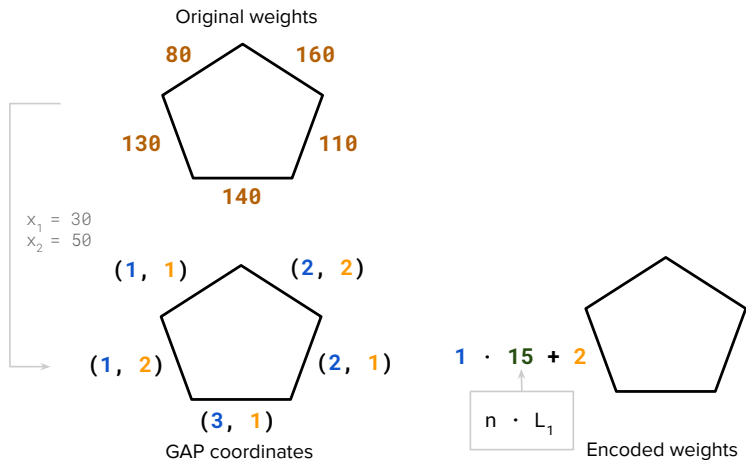
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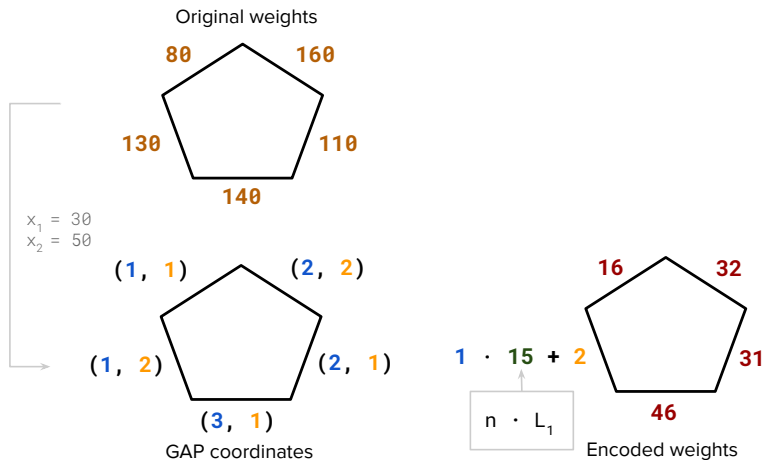
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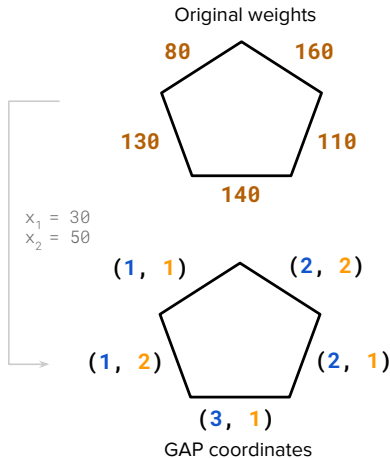
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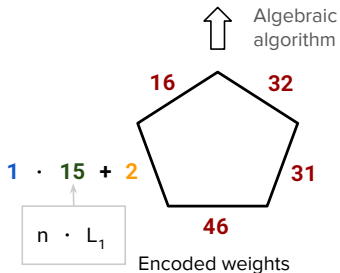
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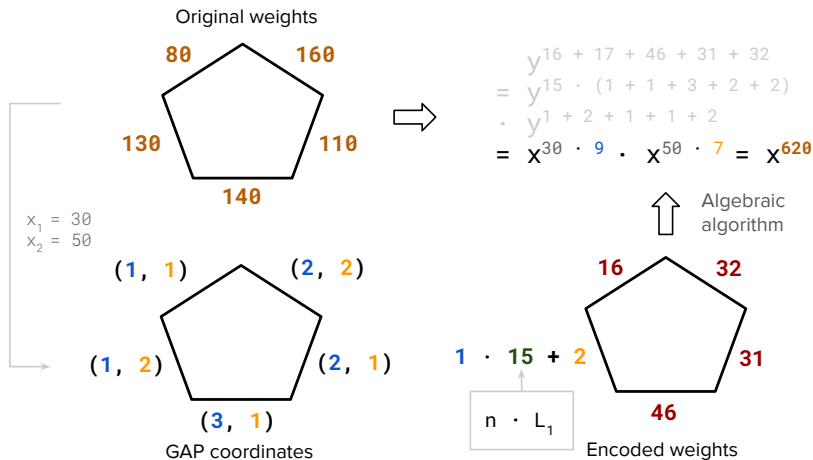
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$$= y^{16 + 17 + 46 + 31 + 32}$$
$$= y^{15} \cdot (1 + 1 + 3 + 2 + 2)$$
$$\cdot y^{1 + 2 + 1 + 1 + 2}$$



# Key Idea: Visualization



# Meta Algorithm: Formalization

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Let  $C\text{-}\mathbf{P}_w$  be the problem where the input weights have doubling constant  $C$ . Example:

- $C\text{-TSP}$  is TSP with  $|w(E) + w(E)| \leq C|w(E)|$ .

**Theorem.** If problem  $\mathbf{P}_w$  satisfies property  $\phi$  and the unweighted version can be solved in time  $O(T(n))$  by an algebraic algorithm  $\mathcal{A}$ , then  $C\text{-}\mathbf{P}_w$  can be solved in time  $O^*(T(n))$ .

## Meta Algorithm: Property $\phi$

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**Property  $\phi$ .** Let  $\mathbf{P}_w$  be a weighted problem and  $\mathbf{P}$  its unweighted counterpart. Let  $I$  be an instance of  $\mathbf{P}_w$  of size  $n$ ,  $w(I)$  its set of weights, and  $W = \max w(I)$ . Then, we say that  $\mathbf{P}_w$  has property  $\phi$  if the following hold:

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1. Any feasible solution to  $I$  is the total weight of a set  $S \subseteq w(I)$ ,
2. There is an algebraic algorithm  $\mathcal{A}$  that solves  $\mathbf{P}$  in  $O(T(n))$ -time and  $\mathbf{P}_w$  in  $O^*(T(n) \cdot W)$ -time, and any intermediate solution to  $I$  produced by  $\mathcal{A}$  is the total weight of a polynomial-size multi-set with support in  $w(I)$ .

# Meta Algorithm: Instantiations

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| <b>P</b>         | <b>P<sub>w</sub></b>           | <b>T(n)</b>      |
|------------------|--------------------------------|------------------|
| Hamiltonicity    | TSP                            | $1.66^n$         |
| Max-Cut          | Weighted Max-Cut               | $2^{\omega n/3}$ |
| <i>k</i> -Clique | Edge-weighted <i>k</i> -Clique | $n^{\omega k/3}$ |

Your favorite problem?

## Summary & Future Work

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- When the set of input weights has **small doubling**, TSP can be solved in the same time as Hamiltonicity.
- Meta algorithm  $\rightarrow$  application to other weighted problems, e.g., weighted Max-Cut, edge-weighted  $k$ -Clique, weighted Steiner tree.

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### Poly-time problems?

- Min-plus convolution—unweighted:  $O(n \log n)$ , weighted:  $O(n^2)$ .
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### Beyond Freiman's theorem

- Currently, Freiman's theorem limited to  $C = O(1)$ .  
What if  $C = O(\alpha(n))$ ?



Gregory A Freiman.

**Foundations of a structural theory of set addition.**

*Translation of Math. Monographs, 37, 1973.*



Tim Randolph and Karol Wegrzycki.

**Parameterized Algorithms on Integer Sets with Small Doubling: Integer Programming, Subset Sum and  $k$ -SUM.**

In Timothy M. Chan, Johannes Fischer, John Iacono, and Grzegorz Herman, editors, *32nd Annual European Symposium on Algorithms, ESA 2024, September 2-4, 2024, Royal Holloway, London, United Kingdom*, volume 308 of *LIPIcs*, pages 96:1–96:19. Schloss Dagstuhl - Leibniz-Zentrum für Informatik, 2024.

doi:10.4230/LIPIcs.ESA.2024.96.