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informatik

Sublinear Algorithms

Lecture 10: Applications II – (Modular) Subset Sum



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Recap: Sparse Convolution

Sparse Polynomial Multiplication:

For $P(X) = \sum_{i=0}^{d-1} p_i X^i$ and $Q(X) = \sum_{i=0}^{d-1} q_i X^i$

their product $R = P \cdot Q$ has coefficients $r_i = \sum_{j=0}^i p_j q_{i-j}$

Given P, Q as lists of non-zero coefficients,

we can compute the coefficients of R in time $\tilde{O}(\text{out})$ where $\text{out} = \#\{i \mid r_i \neq 0\}$

Out-of-bounds values
are interpreted as 0

\tilde{O} hides factor $\text{polylog}(d)$

Sparse Convolution:

For $u, v \in \mathbb{R}^d$, their convolution is the vector $u * v$ with $(u * v)_i = \sum_{j=0}^i u_j v_{i-j}$

Given u, v as lists of non-zero entries,

$u * v$ can be computed in time $\tilde{O}(\text{out})$ where $\text{out} = \#\{i \mid (u * v)_i \neq 0\}$

Recap: Sparse Convolution

Sparse Polynomial Multiplication:

For $P(X) = \sum_{i=0}^{d-1} p_i X^i$ and $Q(X) = \sum_{i=0}^{d-1} q_i X^i$

their product $R = P \cdot Q$ has coefficients $r_i = \sum_{j=0}^i p_j q_{i-j}$

Given P, Q as lists of non-zero coefficients,

we can compute the coefficients of R in time $\tilde{O}(\text{out})$ where $\text{out} = \#\{i \mid r_i \neq 0\}$

Sumset:

For $A, B \subseteq \{0, 1, \dots, d\}$, their sumset is $A + B = \{a + b \mid a \in A, b \in B\}$

Given A, B as lists of all elements,

$A + B$ can be computed in time $\tilde{O}(\text{out})$ where $\text{out} = |A + B|$

w.h.p.

$A + B$ corresponds to the non-zero coefficients of $(\sum_{a \in A} x^a)(\sum_{b \in B} x^b)$

Subset Sum

Goal: $\tilde{O}(\underline{\text{out}})$ time

Given a set X of n positive numbers, compute all sums of subsets of X

$$\text{out} = |\mathcal{S}(X)|$$

Subset Sum:

Given $X = \{x_1, \dots, x_n\} \subseteq \{1, \dots, d\}$

Compute set of all subset sums:

$$\mathcal{S}(X) := \left\{ \sum_{y \in Y} y \mid Y \subseteq X \right\}$$

$A + B = \{a + b \mid a \in A, b \in B\}$ in time $\tilde{O}(\text{out})$

$$X = \{2, 3, 5\}$$

$$\mathcal{S}(X) = \{0, 2, 3, 5, \cancel{7}, \cancel{8}, \cancel{10}\}$$

Subset Sum

Given a set X of n positive numbers, compute all sums of subsets of X

Subset Sum:

Given $X = \{x_1, \dots, x_n\} \subseteq \{1, \dots, d\}$

Compute set of all subset sums:

$$\mathcal{S}(X) := \{\sum_{y \in Y} y \mid Y \subseteq X\}$$

Chain-like: $O(n \cdot \text{out})$

1) Initialize $S_0 := \{0\}$

2) For $i = 1, \dots, n$:

$$S_i := \underbrace{S_{i-1}}_{\text{blue}} \cup \underbrace{(S_{i-1} + x_i)}_{\text{blue}}$$

3) Return S_n

[Bellman'57]

$$S_{i-1} \cup (S_{i-1} + x_i) = \underbrace{S_{i-1} + \{0, x_i\}}_{\text{blue}}$$

$$\mathcal{S}(X) = \{0, x_1\} + \dots + \{0, x_n\}$$

$$S_i = \mathcal{S}(\{x_1, \dots, x_i\})$$

compute S_i in time

$$\underbrace{O(|S_i|)}_{\text{blue}} = \underbrace{O(|S_n|)}_{\text{blue}} = \underbrace{O(\text{out})}_{\text{blue}}$$

$$A \bigoplus B = \{a + b \mid a \in A, b \in B\} \text{ in time } \tilde{O}(\text{out})$$

Subset Sum

Given a set X of n positive numbers, compute all sums of subsets of X

Subset Sum:

Given $X = \{x_1, \dots, x_n\} \subseteq \{1, \dots, d\}$

Compute set of all subset sums:

$$\mathcal{S}(X) := \{\sum_{y \in Y} y \mid Y \subseteq X\}$$

$$S_{i-1} \cup (S_{i-1} + x_i) = S_{i-1} + \{0, x_i\}$$

$$\mathcal{S}(X) = \{0, x_1\} + \dots + \{0, x_n\}$$

$A + B = \{a + b \mid a \in A, b \in B\}$ in time $\tilde{O}(\text{out})$

Chain-like: $\tilde{O}(n \cdot \text{out})$

- 1) Initialize $S_0 := \{0\}$
- 2) For $i = 1, \dots, n$:

$$S_i := \underbrace{S_{i-1}}_{\text{blue arrow}} + \{0, x_i\}$$

- 3) Return S_n

$$S_i = \mathcal{S}(\{x_1, \dots, x_i\})$$

compute S_i in time

$$\tilde{O}(|S_i|) = \tilde{O}(|S_n|) = \tilde{O}(\text{out})$$

Subset Sum

Given a set X of n positive numbers, compute all sums of subsets of X

Tree-like: $\tilde{O}(n \cdot \text{out})$

- 1) Initialize $S_{0,i} := \{0, x_i\}$ for all i
- 2) For $\ell = 1, \dots, \log n$: For $i = 1, \dots, n/2^\ell$:

$$S_{\ell,i} := \underbrace{S_{\ell-1,2i-1} + S_{\ell-1,2i}}$$

- 3) Return $S_{\log n, 1} = \underline{\mathcal{S}(X)}$ Assumes that n is power of 2

Chain-like: $\tilde{O}(n \cdot \text{out})$

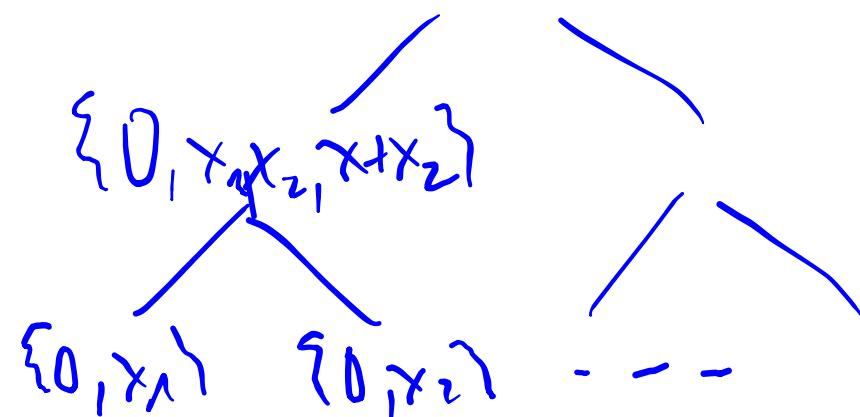
- 1) Initialize $S_0 := \{0\}$
- 2) For $i = 1, \dots, n$:

$$S_i := \underbrace{S_{i-1} + \{0, x_i\}}$$

- 3) Return S_n

$$\mathcal{S}(X) = \{0, x_1\} + \dots + \{0, x_n\} \quad |||$$

$$A + B = \{a + b \mid a \in A, b \in B\} \text{ in time } \tilde{O}(\text{out})$$



Subset Sum

Given a set X of n positive numbers, compute all sums of subsets of X

Tree-like: $\tilde{O}(n \cdot \text{out})$

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- 3) Return $S_{\log n, 1}$ Assumes that n is power of 2

$$\mathcal{S}(X) = \{0, x_1\} + \dots + \{0, x_n\}$$

$$A + B = \{a + b \mid a \in A, b \in B\} \text{ in time } \tilde{O}(\text{out})$$

Lem: For any non-empty A, B

$$|A + B| \geq \underline{|A| + |B| - 1}$$

Proof:

Sort $A = \{a_1, \dots, a_n\}, B = \{b_1, \dots, b_m\}$. Then:

$$a_1 + b_1 < \dots < \boxed{a_n + b_1} < \dots < a_n + b_m \parallel$$

(A \setminus \{a_n\}) \cup \{a_n + b_1\} \cup (B \setminus \{b_1\})

$$a_1 \mid A \mid$$

$$a_i + b_1$$

$$a_n + b_1$$

$$(B)$$

Subset Sum

Given a set X of n positive numbers, compute all sums of subsets of X

Tree-like: $\tilde{O}(n \cdot \text{out})$

- 1) Initialize $S_{0,i} := \{0, x_i\}$ for all i
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- 3) Return $S_{\log n, 1}$ Assumes that n is power of 2

$$\mathcal{S}(X) = \{0, x_1\} + \dots + \{0, x_n\}$$

$$A + B = \{a + b \mid a \in A, b \in B\} \text{ in time } \tilde{O}(\text{out})$$

Lem: For any non-empty A, B

$$|A + B| \geq |A| + |B| - 1$$

Cor: For any non-empty A_1, \dots, A_k

$$|A_1| + \dots + |A_k| \leq \underbrace{|A_1 + \dots + A_k|}_{\text{blue underline}} + k - 1 \quad \underbrace{\text{blue underline}}$$

Proof:

$$\begin{aligned} \underbrace{|A_1 + \dots + A_k|}_{\text{blue underline}} &\geq \underbrace{|A_1 + \dots + A_{k-1}|}_{\text{blue underline}} + \underbrace{|A_k|}_{\text{blue underline}} - 1 \\ &\geq \underbrace{|A_1 + \dots + A_{k-2}|}_{\text{blue underline}} + \underbrace{|A_{k-1}|}_{\text{blue underline}} + \underbrace{|A_k|}_{\text{blue underline}} - 2 \\ &\geq \underbrace{|A_1|}_{\text{blue underline}} + \dots + \underbrace{|A_k|}_{\text{blue underline}} - (k-1) \end{aligned}$$

Subset Sum

Given a set X of n positive numbers, compute all sums of subsets of X

Tree-like: ~~$\tilde{O}(n \times \text{out})$~~

$\tilde{O}(\text{out})$

- 1) Initialize $S_{0,i} := \{0, x_i\}$ for all i
- 2) For $\ell = 1, \dots, \log n$: For $i = 1, \dots, n/2^\ell$:

$$S_{\ell,i} := S_{\ell-1,2i-1} + S_{\ell-1,2i}$$

- 3) Return $S_{\log n, 1}$ Assumes that n is power of 2

$$\mathcal{S}(X) = \{0, x_1\} + \dots + \{0, x_n\}$$

$$A + B = \{a + b \mid a \in A, b \in B\} \text{ in time } \tilde{O}(\text{out})$$

Lem: For any non-empty A, B

$$|A + B| \geq |A| + |B| - 1$$

Cor: For any non-empty A_1, \dots, A_k

$$|A_1| + \dots + |A_k| \leq |A_1 + \dots + A_k| + k - 1$$

Running Time:

$$\begin{aligned} \tilde{O}\left(\sum_{\ell} \sum_i |S_{\ell,i}|\right) &= \tilde{O}\left(\sum_{\ell} \left(out + \frac{n}{2^\ell}\right)\right) \\ &= \tilde{O}(out + n) = \tilde{O}(\text{out}) \end{aligned}$$

Subset Sum

Given a set X of n positive numbers, compute all sums of subsets of X

Tree-like: ~~$\tilde{O}(n \cdot \text{out})$~~ $\tilde{O}(\text{out})$

- 1) Initialize $S_{0,i} := \{0, x_i\}$ for all i
- 2) For $\ell = 1, \dots, \log n$: For $i = 1, \dots, n/2^\ell$:

$$S_{\ell,i} := S_{\ell-1,2i-1} + S_{\ell-1,2i}$$

- 3) Return $S_{\log n, 1}$

Assumes that n is power of 2

Chain-like: $\tilde{O}(n \cdot \text{out})$

- 1) Initialize $S_0 := \{0\}$
 - 2) For $i = 1, \dots, n$:
- $$S_i := S_{i-1} + \{0, x_i\}$$
- 3) Return S_n

$$\mathcal{S}(X) = \{0, x_1\} + \dots + \{0, x_n\}$$

$$A + B = \{a + b \mid a \in A, b \in B\} \text{ in time } \tilde{O}(\text{out})$$

Output-sensitive algorithm for all Subset Sums
follows from Sparse Convolution,
which is solved via sublinear techniques

Modular Subset Sum

Goal: $\tilde{O}(out)$ time

Given a set X of n positive numbers and integer m , compute all sums of subsets of X modulo m

Modular Subset Sum:

Given $X \subseteq \{1, \dots, d\}$ of size n

Compute $\mathcal{S}_m(X) := \mathcal{S}(X) \bmod m$

$$= \left\{ \underbrace{\left(\sum_{y \in Y} y \right)}_{\text{mod } m} \mid \underbrace{Y \subseteq X} \right\}$$

$$\mathcal{S}(X) = \{0, x_1\} + \dots + \{0, x_n\}$$

$$A + B = \{a + b \mid a \in A, b \in B\} \text{ in time } \tilde{O}(out)$$

$$X = \{2, 3, 5\} \quad \mathcal{S}(X) \bmod 7 = \{0, 1, 2, 3, 5\}$$

Modular Subset Sum

Given a set X of n positive numbers and integer m , compute all sums of subsets of X modulo m

Modular Subset Sum:

Given $X \subseteq \{1, \dots, d\}$ of size n

Compute $\mathcal{S}_m(X) := \mathcal{S}(X) \bmod m$

$$= \left\{ \left(\sum_{y \in Y} y \right) \bmod m \mid Y \subseteq X \right\}$$

- 1) Compute $C := A + B$
- 2) Return $\{c \bmod m \mid c \in C\}$

$$\mathcal{S}_m(X) = \{0, x_1\} +_m \dots +_m \{0, x_n\}$$

$A +_m B = \{(a + b) \bmod m \mid a \in A, b \in B\}$ in time $\tilde{\mathcal{O}}(\text{out})$

Chain-like: $\tilde{\mathcal{O}}(n \cdot \text{out})$

1) Initialize $S_0 := \{0\}$

2) For $i = 1, \dots, n$:

$$S_i := S_{i-1} +_m \{0, x_i\}$$

3) Return S_n

$$S_i = \mathcal{S}_m(\{x_1, \dots, x_i\})$$

compute S_i in time
 $\tilde{\mathcal{O}}(|S_i|) = \tilde{\mathcal{O}}(|S_n|) = \tilde{\mathcal{O}}(\text{out})$

Modular Subset Sum

Given a set X of n positive numbers and integer m , compute all sums of subsets of X modulo m

Tree-like: $\tilde{O}(n \cdot \text{out})$

- 1) Initialize $S_{0,i} := \{0, x_i\}$ for all i
- 2) For $\ell = 1, \dots, \log n$: For $i = 1, \dots, n/2^\ell$:

$$S_{\ell,i} := S_{\ell-1,2i-1} +_m S_{\ell-1,2i}$$

- 3) Return $S_{\log n, 1}$

Chain-like: $\tilde{O}(n \cdot \text{out})$

- 1) Initialize $S_0 := \{0\}$
- 2) For $i = 1, \dots, n$:
$$S_i := S_{i-1} +_m \{0, x_i\}$$
- 3) Return S_n

$$\mathcal{S}_m(X) = \{0, x_1\} +_m \dots +_m \{0, x_n\}$$

$$A +_m B = \{(a + b) \bmod m \mid a \in A, b \in B\} \text{ in time } \tilde{O}(\text{out})$$

Modular Subset Sum

Given a set X of n positive numbers and integer m , compute all sums of subsets of X modulo m

Tree-like: $\tilde{O}(n \cdot \text{out})$

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$$S_{\ell,i} := S_{\ell-1,2i-1} +_m S_{\ell-1,2i}$$
- 3) Return $S_{\log n, 1}$

Lem: For any non-empty A, B

$$|A + B| \geq |A| + |B| - 1$$

$|A +_m B| \geq \underline{|A| + |B| - 1}$
is wrong in general!

e.g. $A = B = \{0, \dots, m-1\}$

$$\mathcal{S}_m(X) = \{0, x_1\} +_m \dots +_m \{0, x_n\}$$

$$A +_m B = \{(a + b) \bmod m \mid a \in A, b \in B\} \text{ in time } \tilde{O}(\text{out})$$

Modular Subset Sum

Given a set X of n positive numbers and integer m , compute all sums of subsets of X modulo m

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- 3) Return $S_{\log n, 1}$

$$\mathcal{S}_m(X) = \{0, x_1\} +_m \dots +_m \{0, x_n\}$$

$$A +_m B = \{(a + b) \bmod m \mid a \in A, b \in B\} \text{ in time } \tilde{O}(\text{out})$$

Approaches to time $\tilde{O}(\text{out})$:

1) via sketching

2) via additive combinatorics:

Kneser's Thm:

$$|A +_m B| \geq |A| + |B| - |\text{Sym}(A + B)|$$

$$\text{where } \text{Sym}(Z) = \{x \mid Z +_m x = Z\}$$

3) via dynamic strings

[Mehlhorn, Sundar, Uhrig 1997]

Modular Subset Sum by Sketching

Data Structure $\mathcal{D}(\nu)$: maintains vector $\nu \in \mathbb{R}^m$ s.t.

- **Initialize:** set $\nu = (0, \dots, 0)$
- **Read**(i): return ν_i
- **Write**(i, s): set $\nu_i := s$
- **TestZero:** checks if $\nu = (0, \dots, 0)$; correct w.h.p.
- **Support:** computes $\text{supp}(\nu)$; correct w.h.p.
- **Linear:** given access to $\mathcal{D}(u), \mathcal{D}(\nu)$ and α, β , we can simulate $\mathcal{D}(\alpha u + \beta \nu)$
- **Rotation:** given access to $\mathcal{D}(\nu)$ and x , we can simulate $\mathcal{D}(\text{rot}_x(\nu)) = \mathcal{D}(\nu_x, \dots, \nu_{\cancel{x}-1}, \nu_0, \dots, \nu_{x-1})$

} time $\tilde{\mathcal{O}}(1) = \text{polylog}(m)$

} time $\tilde{\mathcal{O}}(|\text{supp}(\nu)|)$ w.h.p.

} time: constant-factor overhead

Modular Subset Sum by Sketching

Goal: Update S_{i-1} to S_i in time $\tilde{O}(|S_i \setminus S_{i-1}|)$

Then total time is $\tilde{O}(\sum_i |S_i \setminus S_{i-1}|)$

$$\begin{aligned}&= \tilde{O}(\sum_i (|S_i| - |S_{i-1}|)) \\&= \tilde{O}(|S_n| - |S_0|) \\&= \tilde{O}(out)\end{aligned}$$

Chain-like: $\tilde{O}(n \cdot out)$

1) Initialize $S_0 := \{0\}$

2) For $i = 1, \dots, n$:

$$S_i := S_{i-1} +_m \{0, x_i\}$$

3) Return S_n

$$A +_m B = \{(a + b) \bmod m \mid a \in A, b \in B\}$$

Modular Subset Sum by Sketching

Goal: Update S_{i-1} to S_i in time $\tilde{O}(|S_i \setminus S_{i-1}|)$

Note $S_i = S_{i-1} \cup (S_{i-1} +_m x_i)$

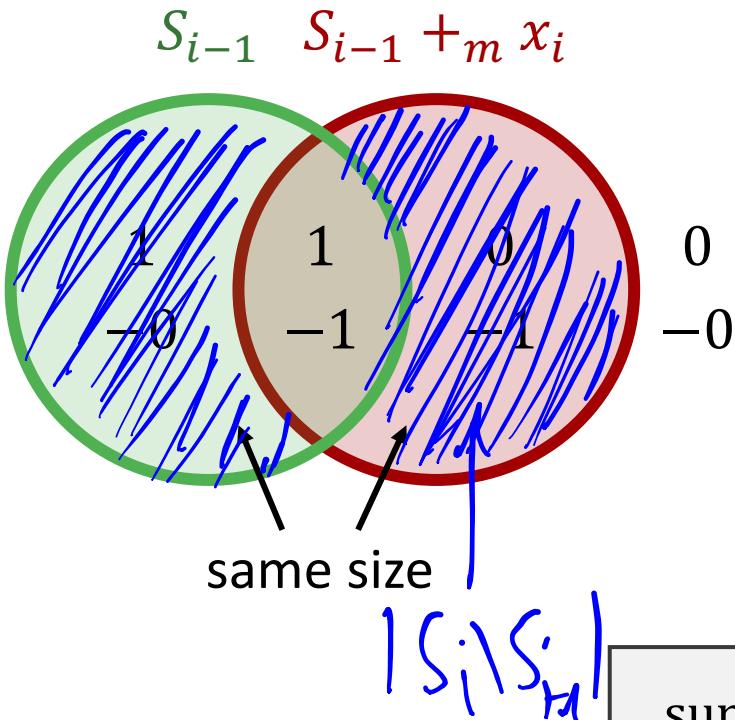
$$\nu := \mathbf{1}[S_{i-1}] - \mathbf{1}[S_{i-1} +_m x_i]$$

$$S_i = \underline{S_{i-1}} \cup \underline{\text{supp}(\nu)}$$

$$|\text{supp}(\nu)| = 2\underline{|S_i \setminus S_{i-1}|}$$

Refined Goal: Compute $\text{supp}(\nu)$ in time $\tilde{O}(|\text{supp}(\nu)|)$

$$\text{Total time is } \tilde{O}(\sum_i \underline{|S_i \setminus S_{i-1}|}) = \tilde{O}(\underline{\text{out}})$$



Chain-like: $\tilde{O}(n \cdot \text{out})$

1) Initialize $S_0 := \{0\}$

2) For $i = 1, \dots, n$:

$$S_i := S_{i-1} +_m \{0, x_i\}$$

3) Return S_n

indicator vector $\mathbf{1}[A]$:

$$\text{supp}(\nu) = \{i \mid \nu_i \neq 0\}$$

$$\mathbf{1}[A]_a = \begin{cases} 1 & \text{if } a \in A \\ 0 & \text{otherwise} \end{cases}$$

$$A +_m B = \{(a + b) \bmod m \mid a \in A, b \in B\}$$

Modular Subset Sum by Sketching

Data Structure $\mathcal{D}(\nu)$: maintains vector $\nu \in \mathbb{R}^m$ s.t.

- **Initialize:** set $\nu = (0, \dots, 0)$
 - **Read**(i): return ν_i
 - **Write**(i, s): set $\nu_i := s$
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 - **Linear:** given access to $\mathcal{D}(u), \mathcal{D}(\nu)$ and α, β , we can simulate $\mathcal{D}(\alpha u + \beta \nu)$
 - **Rotation:** given access to $\mathcal{D}(\nu)$ and x , we can simulate $\mathcal{D}(\text{rot}_x(\nu)) = \mathcal{D}(\nu_x, \dots, \nu_{d-1}, \nu_0, \dots, \nu_{x-1})$
-
- time $\tilde{O}(1)$
- time $\tilde{O}(|\text{supp}(\nu)|)$ w.h.p.
- time: constant-factor overhead

Modular Subset Sum by Sketching

$$\nu := \mathbf{1}[S_{i-1}] \ominus \mathbf{1}[S_{i-1} +_m x_i]$$

$$S_i = S_{i-1} \cup \underline{\text{supp}(\nu)} \parallel$$

$$|\text{supp}(\nu)| = 2|S_i \setminus S_{i-1}|$$

Refined Goal: Compute $\text{supp}(\nu)$ in time $\tilde{O}(|\text{supp}(\nu)|)$

Total time is $\tilde{O}(\sum_i |S_i \setminus S_{i-1}|) = \tilde{O}(out)$

Modular Subset Sum: $\tilde{O}(out)$

Maintain $\mathcal{D} = \mathcal{D}(\mathbf{1}[S_i])$

1) // Initialize $S_0 := \{0\}$

$\mathcal{D}.\text{Initialize}$ and $\mathcal{D}.\text{Write}(0,1)$

2) For $i = 1, \dots, n$:

// here: $\mathcal{D} = \mathcal{D}(\mathbf{1}[S_{i-1}])$

Simulate $\mathcal{D}' :=$

$\mathcal{D}(\mathbf{1}[S_{i-1}] - \mathbf{1}[S_{i-1} +_m x_i])$

$T := \mathcal{D}'.\text{Support}$

For each $x \in T$: $\mathcal{D}.\text{Write}(x, 1)$

Modular Subset Sum by Sketching

Data Structure $\mathcal{D}(\nu)$: maintains vector $\nu \in \mathbb{R}^m$ s.t.

- **Initialize:** set $\nu = (0, \dots, 0)$
- **Read(i):** return ν_i
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- **Linear:** given access to $\mathcal{D}(u), \mathcal{D}(v)$ and α, β , we can simulate $\mathcal{D}(\alpha u + \beta v)$
- **Rotation:** given access to $\mathcal{D}(\nu)$ and x , we can simulate $\mathcal{D}(\text{rot}_x(\nu)) = \mathcal{D}(\nu_x, \dots, \nu_{d-1}, \nu_0, \dots, \nu_{x-1})$

Modular Subset Sum: $\tilde{\mathcal{O}}(\text{out})$

Maintain $\mathcal{D} = \mathcal{D}(\mathbf{1}[S_i])$

1) // Initialize $S_0 := \{0\}$

$\mathcal{D}.\text{Initialize}$ and $\mathcal{D}.\text{Write}(0, 1)$

2) For $i = 1, \dots, n$:

// here: $\mathcal{D} = \mathcal{D}(\mathbf{1}[S_{i-1}])$

Simulate $\mathcal{D}' :=$

$\mathcal{D}(\mathbf{1}[S_{i-1}] - \mathbf{1}[S_{i-1} +_m x_i])$

$T := \mathcal{D}'.\text{Support}$

For each $x \in T$: $\mathcal{D}.\text{Write}(x, 1)$

Implementation of \mathcal{D}

Data Structure $\mathcal{D}(\nu)$: maintains vector $\nu \in \mathbb{R}^m$ s.t.

- ✓ **Initialize:** set $\nu = (0, \dots, 0)$
- ✓ **Read(i):** return ν_i
- ✓ **Write(i, s):** set $\nu_i := s$
- **TestZero:** checks if $\nu = (0, \dots, 0)$; correct w.h.p.
- **Support:** computes $\text{supp}(\nu)$; correct w.h.p.
- ✓ **Linear:** given access to $\mathcal{D}(u), \mathcal{D}(\nu)$ and α, β , we can simulate $\mathcal{D}(\alpha u + \beta \nu)$
- **Rotation:** given access to $\mathcal{D}(\nu)$ and x , we can simulate $\mathcal{D}(\text{rot}_x(\nu)) = \mathcal{D}(\nu_x, \dots, \nu_{d-1}, \nu_0, \dots, \nu_{x-1})$

Implementation of $\mathcal{D}(\nu)$:

For $b = 1, \dots, \log m$:

Pick set P_b of $\Theta(\log m)$ random primes in the range $\Theta(2^b \log^2 m)$

$$P := \bigcup_b P_b$$

$$\text{Store } \nu \in \mathbb{R}^m$$

Sparse representation
 ω is m -th root of unity

Store $\text{fold}(\nu, p) \in \mathbb{R}^p$ for each $p \in P$

$$\text{fold}(\nu, p)_i = \sum_{j:j \bmod p=i} \nu_j$$

Store $\text{fold}'(\nu, p) \in \mathbb{C}^p$ for each $p \in P$

$$\text{fold}'(\nu, p)_i = \sum_{j:j \bmod p=i} \nu_j \omega^j$$

Implementation of \mathcal{D}

- **Rotation:** given $\mathcal{D}(\nu)$ and x , simulate $\mathcal{D}(\nu')$
for $\nu' := \text{rot}_x(\nu) = (\nu_x, \dots, \nu_{d-1}, \nu_0, \dots, \nu_{x-1})$

$$\nu'_j = \nu_{j+m^x}$$

$$\begin{aligned}\text{fold}(\nu', p)_i &= \sum_{j:j \bmod p = i} \nu_{j+m^x} \\ &= \sum_{j:j \bmod p = i+m^x} \nu_j \\ &= \text{fold}(\nu, p)_{i+m^x}\end{aligned}$$

$$\begin{aligned}\text{fold}'(\nu', p)_i &= \sum_{j:j \bmod p = i} \nu_{j+m^x} \cdot \omega^j \\ &= \sum_{j:j \bmod p = i+m^x} \nu_j \cdot \omega^{\underline{j-x}} \\ &= \text{fold}'(\nu, p)_{i+m^x} \cdot \underline{\omega^{-x}}\end{aligned}$$

Implementation of $\mathcal{D}(\nu)$:

For $b = 1, \dots, \log m$:

Pick set P_b of $\Theta(\log m)$ random primes in the range $\Theta(2^b \log^2 m)$

$$P := \bigcup_b P_b$$

Store $\nu \in \mathbb{R}^m$ ω is m -th root of unity

Store $\text{fold}(\nu, p) \in \mathbb{R}^p$ for each $p \in P$

$$\text{fold}(\nu, p)_i = \sum_{j:j \bmod p = i} \nu_j$$

Store $\text{fold}'(\nu, p) \in \mathbb{C}^p$ for each $p \in P$

$$\text{fold}'(\nu, p)_i = \sum_{j:j \bmod p = i} \nu_j \omega^j$$

Implementation of \mathcal{D}

- **TestZero**: checks if $v = (0, \dots, 0)$; correct w.h.p.

- 1) For each $p \in P$:
- 2) For $\Theta(\log^2 m)$ random $r \in \{0, \dots, p - 1\}$:
- 3) If $\text{fold}(v, p)_r \neq 0$: return $v \neq \vec{0}$
- 4) return $v = \vec{0}$

If $v \neq \vec{0}$: let $b = \log|\text{supp}(v)|$, fix $p \in P_b$

Each $v_j \neq 0$ is **isolated** with probability $\Omega(1)$, since

$$\mathbb{E} \left[\begin{array}{l} \#j' \neq j \text{ s.t. } v_{j'} \neq 0 \text{ and} \\ (j - j') \bmod p = 0 \end{array} \right] \leq \frac{|\text{supp}(v)| \cdot \log m}{\#\text{primes in } \Theta(2^b \log^2 m)} \stackrel{\leq 1}{\underset{\approx 1}{\approx}}$$

*prime
Activations*

Implementation of $\mathcal{D}(v)$:

For $b = 1, \dots, \log m$:

Pick set P_b of $\Theta(\log m)$ random primes in the range $\Theta(2^b \log^2 m)$

$$P := \bigcup_b P_b$$

Store $v \in \mathbb{R}^m$ ω is m -th root of unity

Store $\text{fold}(v, p) \in \mathbb{R}^p$ for each $p \in P$

$$\text{fold}(v, p)_i = \boxed{\sum_{j:j \bmod p=i} v_j} = \vee_j$$

Store $\text{fold}'(v, p) \in \mathbb{C}^p$ for each $p \in P$

$$\text{fold}'(v, p)_i = \sum_{j:j \bmod p=i} v_j \omega^j$$

Implementation of \mathcal{D}

- **TestZero**: checks if $v = (0, \dots, 0)$; correct w.h.p.

- 1) For each $p \in P$:
- 2) For $\Theta(\log^2 m)$ random $r \in \{0, \dots, p - 1\}$:
- 3) If $\text{fold}(v, p)_r \neq 0$: return $v \neq \vec{0}$
- 4) return $v = \vec{0}$

If $v \neq \vec{0}$: let $b = \log|\text{supp}(v)|$, fix $p \in P_b$

Each $v_j \neq 0$ is **isolated** with probability $\Omega(1)$

W. prob. $\Omega(1)$, an $\Omega(1)$ -fraction of all $v_j \neq 0$ are isolated

W. prob. $\Omega(1)$, an $\Omega\left(\frac{1}{\log^2 m}\right)$ -fraction of all r 's yield $v \neq \vec{0}$

Implementation of $\mathcal{D}(v)$:

For $b = 1, \dots, \log m$:

Pick set P_b of $\Theta(\log m)$ random primes in the range $\Theta(2^b \log^2 m)$

$$P := \bigcup_b P_b$$

Store $v \in \mathbb{R}^m$ ω is m -th root of unity

Store $\text{fold}(v, p) \in \mathbb{R}^p$ for each $p \in P$

$$\text{fold}(v, p)_i = \sum_{j:j \bmod p=i} v_j$$

Store $\text{fold}'(v, p) \in \mathbb{C}^p$ for each $p \in P$

$$\text{fold}'(v, p)_i = \sum_{j:j \bmod p=i} v_j \omega^j$$

Implementation of \mathcal{D}

- **Support:** computes $\text{supp}(\nu)$; correct w.h.p.

First suppose we know $b = \log|\text{supp}(\nu)|$

PromiseSupport(b):

- 1) Initialize $S = \emptyset$, $u = (0, \dots, 0)$
- 2) For each $p \in P_b$: For $i = 0, \dots, p - 1$:
- 3) $z := \text{fold}'(\nu, p)_i / \text{fold}(\nu, p)_i$
- 4) $x := \text{Invert}(z) // \omega^j \mapsto j$
- 5) If $v_x \neq 0$: $S := S \cup \{x\}$, $u_x := v_x$
- 5) Return (S, u)

Each $v_j \neq 0$ is **isolated**
by some p w.h.p.

Implementation of $\mathcal{D}(\nu)$:

For $b = 1, \dots, \log m$:

Pick set P_b of $\Theta(\log m)$ random
primes in the range $\Theta(2^b \log^2 m)$

$$P := \bigcup_b P_b$$

Store $\nu \in \mathbb{R}^m$ ω is m -th root of unity

Store $\text{fold}(\nu, p) \in \mathbb{R}^p$ for each $p \in P$

$$\text{fold}(\nu, p)_i = \sum_{j:j \bmod p = i} v_j$$

Store $\text{fold}'(\nu, p) \in \mathbb{C}^p$ for each $p \in P$

$$\text{fold}'(\nu, p)_i = \sum_{j:j \bmod p = i} v_j \omega^j$$

Implementation of \mathcal{D}

- **Support:** computes $\text{supp}(\nu)$; correct w.h.p.

First suppose we know $b = \log|\text{supp}(\nu)|$

Now if we do not know b :

1) For $b = 1, \dots, \log m$:

3) $(S, u) := \text{PromiseSupport}(b)$

4) Build $\mathcal{D}(u)$

5) If $\mathcal{D}(u - v).\text{TestZero}$: return S

Can only terminate with correct vector $u = v$

Terminates at the latest for $b = \log|\text{supp}(\nu)|$

Implementation of $\mathcal{D}(\nu)$:

For $b = 1, \dots, \log m$:

Pick set P_b of $\Theta(\log m)$ random primes in the range $\Theta(2^b \log^2 m)$

$$P := \bigcup_b P_b$$

Store $v \in \mathbb{R}^m$ ω is m -th root of unity

Store $\text{fold}(\nu, p) \in \mathbb{R}^p$ for each $p \in P$

$$\text{fold}(\nu, p)_i = \sum_{j:j \bmod p=i} \nu_j$$

Store $\text{fold}'(\nu, p) \in \mathbb{C}^p$ for each $p \in P$

$$\text{fold}'(\nu, p)_i = \sum_{j:j \bmod p=i} \nu_j \omega^j$$

More Material

| *Modular Subset Sum by Sketching:*

[Axiotis, Backurs, Tzamos „Fast Modular Subset Sum using Linear Sketching“ 2019]

See also:

[Bringmann „A Near-Linear Pseudopolynomial Time Algorithm for Subset Sum“ 2017]

[Bringmann, Nakos „Top-k Convolution and the Quest for Near-Linear Output-Sensitive Subset Sum“ 2020]

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See you on Thursday!