

# Classical and Quantum Algorithms for Variants of Subset-Sum via Dynamic Programming

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## The Problems

### SUBSET-SUM

**Input:** Multiset  $\{a_1, \dots, a_n\}$  and target  $m$

**Output:**  $S \subseteq \{1, \dots, n\}$  such that  $\sum_{i \in S} a_i = m$

### SHIFTED-SUMS

**Input:** Multiset  $\{a_1, \dots, a_n\}$  and shift  $s$

**Output:**  $S_1 \neq S_2 \subseteq \{1, \dots, n\}$  s.t.  $\sum_{i \in S_1} a_i + s = \sum_{i \in S_2} a_i$

Special cases of SHIFTED-SUMS:

#### EQUAL-SUMS

**Input:** Set  $\{a_1, \dots, a_n\}$

**Output:**  $S_1 \neq S_2 \subseteq \{1, \dots, n\}$  s.t.  $\sum_{i \in S_1} a_i = \sum_{i \in S_2} a_i$

#### PIGEONHOLE EQUAL-SUMS

**Input:** Set  $\{a_1, \dots, a_n\}$  s.t.  $\sum_{i=1}^n a_i < 2^n - 1$

**Output:**  $S_1 \neq S_2 \subseteq \{1, \dots, n\}$  s.t.  $\sum_{i \in S_1} a_i = \sum_{i \in S_2} a_i$

## Known Results

### Classical results:

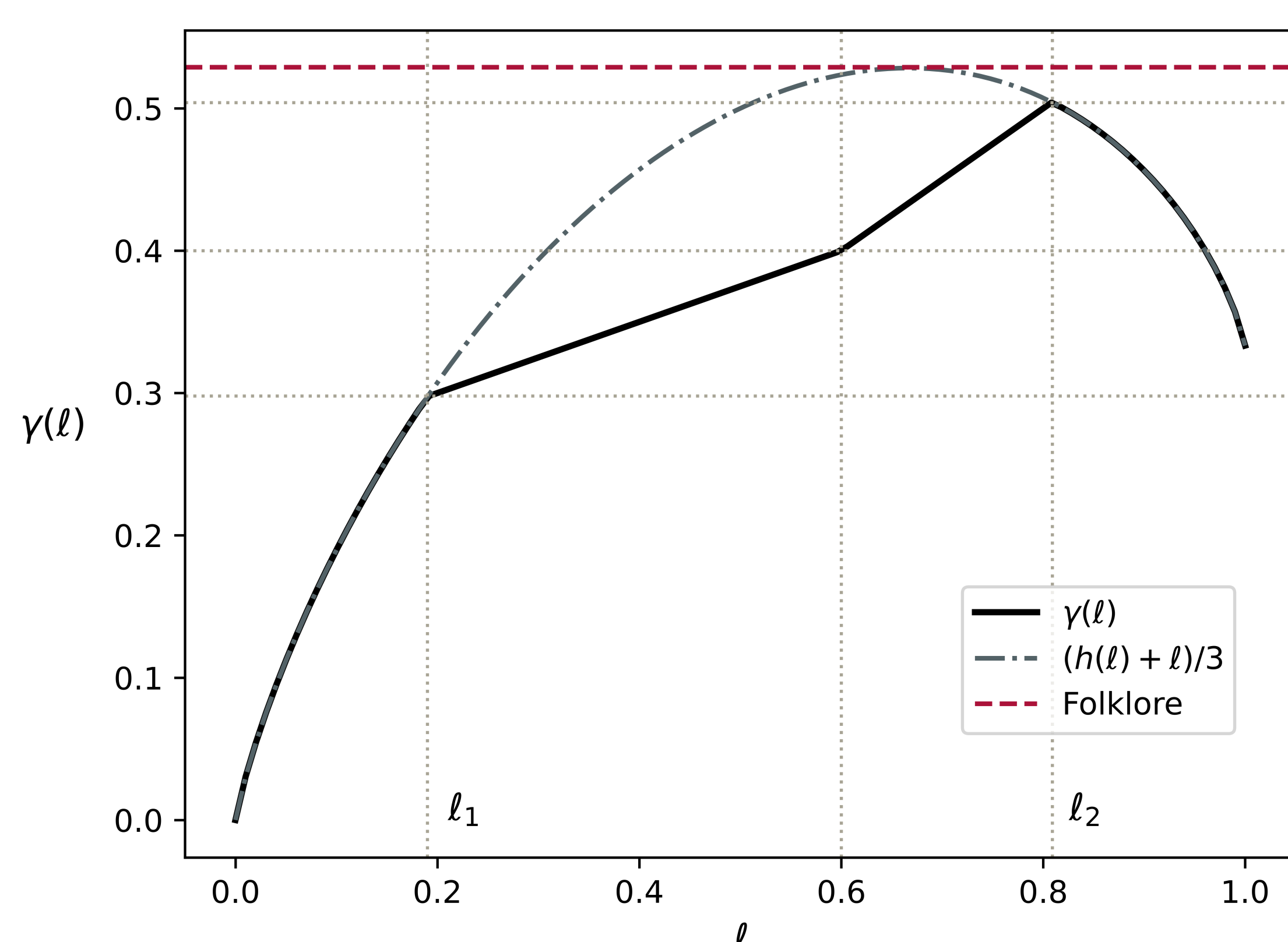
- SUBSET-SUM  $\in \tilde{O}(2^{n/2})$  [HS'74] Meet in the Middle
- SUBSET-SUM  $\in \tilde{O}(n + m)$  [Bri'17]
- SETH  $\Rightarrow \forall \epsilon > 0, \exists \delta > 0$ , SUBSET-SUM  $\notin O(m^{1-\epsilon} 2^{n\delta})$  [ABHS'19]
- EQUAL-SUMS  $\in \tilde{O}(3^{n/2}) \leq O(2^{0.793n})$  [Woe'08] MiM
- EQUAL-SUMS  $\in O(2^{0.773n})$  [MNPW'19]

### Quantum results:

- SUBSET-SUM  $\in \tilde{O}(2^{n/3})$  [BJLM'13] MiM
- EQUAL-SUMS  $\in \tilde{O}(3^{n/3}) \leq O(2^{0.529n})$  MiM

## Our Results

	Classical	Quantum
SUBSET-SUM (not MiM)	$\tilde{O}(2^{n/2})$	$\tilde{O}(2^{n/3})$
SHIFTED-SUMS	$O(2^{0.773n})$	$O(2^{0.504n})$
PIGEONHOLE EQUAL-SUMS	$O(2^{n/2})$	$O(2^{0.4n})$



## Dynamic Programming Datastructure

$$T_{p,k} = \left\{ S \subseteq \{1, \dots, n\} : \sum_{i \in S} a_i \equiv k \pmod{p} \right\}$$

1 - Compute the size  $|T_{p,k}|$  by **dynamic programming**:

Let  $t_p[i, j] = |\{S \subseteq \{1, \dots, i\} : \sum_{i \in S} a_i \equiv j \pmod{p}\}|$ .

The table  $t_p[i, j]$  can be constructed in time  $O(np)$ :

$$t_p[i, j] = t_p[i-1, j] + t_p[i-1, (j - a_i) \bmod p].$$

2 - Given random access to the table  $t_p$ , we construct an **oracle** that gives access to the elements of  $T_{p,k}$  in time  $O(n)$  (for a certain total order  $<$  over  $T_{p,k}$ ).

## Main Ideas for Quantum SHIFTED-SUMS

**Maximum solution size,  $\ell \in (0, 1)$ :**

$$\ell n = \max \{ |S_1| + |S_2| : S_1, S_2 \text{ disjoint solutions} \}$$

Suppose that  $\ell$  is known. Then:

- If  $\ell < 0.190$  or  $\ell > 0.809$ , use **quantum MiM**.
- If  $0.190 \leq \ell \leq 0.809$ , use **representation technique**.

**Example when  $\ell = 3/5$ :**

- 1 - Choose a **random prime**  $p \in \{2^{2n/5}, \dots, 2^{2n/5+1}\}$ .
- 2 - Choose a **random index**  $k \in [0, p-1]$ .
- 3 - Search for a solution in  $T_{p,k}$  using the **dynamic programming oracle** and Ambainis' **Collision Finding**.

$\rightarrow$  Expected bin size  $|T_{p,k}| \approx 2^{3n/5}$ ,

- Sufficiently **large** to contain a solution.
- Sufficiently **small** to keep the cost of search low:  $\tilde{O}(2^{2n/5})$ .

**Additional ideas when  $\ell < 3/5$ :**

- Choose  $p \in \{2^{(1+\ell)n/4}, \dots, 2^{(1+\ell)n/4+1}\}$ .
- **Multiple solutions** in a random  $T_{p,k}$ .
- New **faster** Collision Finding for multiple collisions.

**Additional ideas when  $\ell > 3/5$ :**

- Choose  $p \in \{2^{2n/5}, \dots, 2^{2n/5+1}\}$ .
- Solution in a random  $T_{p,k}$  with **exp. small probability**.
- Use **Variable-Time Amplitude Amplification** to boost the success probability of sampling a good  $T_{p,k}$ .