## CS 506 – Class Notes

Scriber: Mohsen Dehghankar

October 1st 2025

## 1 Simon's Problem

Our goal is to solve the following problem:

**Simon's problem** Given a black-box function  $f: \{0,1\}^n \to \{0,1\}^n$  with the guarantee that:

• There exists a vector  $\vec{s} \in \{0,1\}^n$ , such that  $\vec{s} \neq \vec{0}$  and we have:

$$f(\vec{x}) = f(\vec{x} \oplus \vec{s})$$

for all  $\vec{x} \in \{0,1\}^n$ 

The goal is to find  $\vec{s}$  using (a few) queries to f.

### 1.1 Classical Result on Hardness

In the classical setting, where no quantum queries are allowed, one can show the following claim:

Claim: Let  $\mathcal{A}$  be a randomized algorithm that solves Simon's problem with probability of success at least  $\frac{2}{3}$ . Then, it should make  $\Omega(2^{\frac{n}{3}})$  queries to function f. In other words, it's exponential to n.

However, here, in this context, we are allowed to have *Quantum Queries*. That's why this classical hardness result is not applied to what we are going to discuss.

### 2 Preliminaries

We start by giving some preliminaries on vector spaces from linear algebra. Then, we define Generalized Simon's problem and give an intuition on how to solve it with linear number of queries (linear to dimension n).

#### 2.1 Standard Vector Space

**Definition** The set  $\mathbb{Z}_2^n$  denotes the vector space consisting of all binary vectors  $\vec{x} \in \{0,1\}^n$ .

• Operations: Vector addition is defined component-wise modulo 2, denoted by  $\oplus$ . For  $\vec{x} = (x_1, \dots, x_n)$  and  $\vec{y} = (y_1, \dots, y_n)$ ,

$$\vec{x} \oplus \vec{y} \equiv (x_1 \oplus y_1, \dots, x_n \oplus y_n)$$
, where  $\oplus$  is addition modulo 2.

- **Dimension:** The dimension of  $\mathbb{Z}_2^n$  is n, corresponding to the number of basis vectors.
- Standard Basis: A canonical basis for  $\mathbb{Z}_2^n$  is given by

$$e_1 = (1, 0, 0, \dots, 0), \quad e_2 = (0, 1, 0, \dots, 0), \quad \dots, \quad e_n = (0, 0, 0, \dots, 1).$$

## 2.2 Example of a Vector Space

Consider the set  $S = {\vec{0}, \vec{s}}$ . The span of S is

$$\operatorname{span}(S) = \{\vec{0}, \vec{s}\},\$$

which is the same as S itself. Hence, S is a subspace of  $\mathbb{Z}_2^n$ .

• Basis of S: The only nonzero basis vector is  $\vec{s}$ , so the dimension of S is 1.

$$\vec{0} = \vec{s} + \vec{s},\tag{1}$$

$$\vec{s} = \vec{s}.\tag{2}$$

Now consider the orthogonal subspace

$$S^{\perp} = \{ \vec{z} \in \mathbb{Z}_2^n \mid \vec{z} \cdot \vec{s} = 0 \},$$

where  $\cdot$  denotes the inner product modulo 2.

- The dimension of  $S^{\perp}$  is n-1.
- In general, for a subspace  $S \subseteq \mathbb{Z}_2^n$ ,

$$\dim(\mathbb{Z}_2^n) = \dim(S) + \dim(S^{\perp}).$$

• Thus,  $S^{\perp}$  has n-1 basis vectors, say  $\{\vec{b}_1, \ldots, \vec{b}_{n-1}\}$ , where

$$\vec{b}_i \cdot \vec{s} = 0 \quad \forall i < n - 1,$$

and the basis vectors  $\vec{b}_i, \vec{b}_j$  are mutually orthogonal (with respect to the modulo 2 inner product).

# 3 A Bridge Problem

We now consider the following problem: suppose you are given the subspace  $S^{\perp}$ , as defined above. Your goal is to recover the corresponding vector  $\vec{s}$  such that

$$S^{\perp} = \{ \vec{z} \in \mathbb{Z}_2^n \mid \vec{z} \cdot \vec{s} = 0 \}.$$

Assuming you are allowed to sample from  $S^{\perp}$ , how can you determine  $\vec{s}$ ?

Suppose that we sample uniformly at random vectors from  $S^{\perp}$ , n-1 times, and we get vectors  $\vec{c_1}, \cdots \vec{c_{n-1}}$  and suppose that  $c_i$  are mutually *independent* (\* what is the probability of this happening?).

#### 3.1 Uniform Sampling

One approach is to sample n-1 vectors  $\vec{c}_1, \ldots, \vec{c}_{n-1}$  uniformly from  $S^{\perp}$ . If these vectors are linearly independent (discussed later), they form a basis for  $S^{\perp}$ . The corresponding vector  $\vec{s}$  is then the unique vector orthogonal to all of them.

**Formulation.** We need to solve the system of equations

$$\vec{c_i} \cdot \vec{s} = 0 \pmod{2}, \quad \forall i = 1, \dots, n-1.$$

This can be expressed in matrix form as:

$$M = \begin{bmatrix} \vec{c}_1^\top \\ \vec{c}_2^\top \\ \vdots \\ \vec{c}_{n-1}^\top \end{bmatrix}, \qquad s = \begin{bmatrix} s_1 \\ s_2 \\ \vdots \\ s_n \end{bmatrix}, \qquad Ms \equiv \vec{0} \pmod{2}.$$

**Solution.** Solving this linear system over  $\mathbb{Z}_2$  can be done using Gaussian elimination in  $O(n^3)$  time. The nontrivial solution obtained gives the desired vector  $\vec{s}$ .

#### 3.1.1 Probability of Independence of Sampled Vectors

**Problem.** What is the probability that the sampled vectors  $\vec{c}_1, \dots, \vec{c}_{n-1}$  from  $S^{\perp}$  are linearly independent?

Step 1: Conditional probability. The probability can be decomposed as

 $P[\text{all independent}] = P[\vec{c}_2 \text{ independent} \mid \vec{c}_1] \cdot P[\vec{c}_3 \text{ independent} \mid \vec{c}_1, \vec{c}_2] \cdots P[\vec{c}_{n-1} \text{ independent} \mid \vec{c}_1, \dots, \vec{c}_{n-2}].$ 

Step 2: Single step. Suppose  $T_{k-1} = \{\vec{c}_1, \dots, \vec{c}_{k-1}\}$  is linearly independent. Then the span of  $T_{k-1}$  contains  $2^{k-1}$  vectors. Since there are in total  $2^{n-1}$  possible vectors in  $S^{\perp}$ , the probability that a newly chosen vector  $\vec{c}_k$  lies outside this span (and is therefore independent) is

$$P[\vec{c}_k \text{ independent} \mid T_{k-1}] = \frac{2^{n-1} - 2^{k-1}}{2^{n-1}}.$$

Step 3: Full product. Thus, the overall probability that n-1 sampled vectors are linearly independent is

$$P[\text{success}] = \prod_{k=2}^{n-1} \left( 1 - \frac{2^{k-1}}{2^{n-1}} \right) = \left( 1 - \frac{1}{2} \right) \left( 1 - \frac{1}{4} \right) \cdots \left( 1 - \frac{1}{2^{n-2}} \right)$$

Here, we skipped k = 1, because for the first vector  $\vec{c_1}$ , the probability is always 1.

**Step 4: Limit.** As  $n \to \infty$ , this product converges to a constant known as the *binary q-series* (or 2-series). As a result, we get:

$$\lim_{n\to\infty} P[\text{success}] > 0.288.$$

#### 3.1.2 How to Avoid Complicated Calculations

In general, for any  $0 < a_i \le 1$ , we have the inequality

$$(1-a_1)(1-a_2)\cdots(1-a_t) > 1-(a_1+a_2+\cdots+a_t).$$

**Application.** Applying this bound to P[success], we obtain:

$$P[\text{success}] = \left(1 - \frac{1}{2}\right) \left(1 - \frac{1}{2^2}\right) \cdots \left(1 - \frac{1}{2^{n-2}}\right) \ge \left(1 - \frac{1}{2}\right) \cdot \left(1 - \sum_{i=2}^{n-2} \frac{1}{2^i}\right)$$

**Simplification.** We can obtain an upper bound for the right term based on the geometric series:

$$\sum_{i=2}^{n-2} \frac{1}{2^i} \le \sum_{i=2}^{\infty} \frac{1}{2^i} = \frac{\frac{1}{4}}{1 - \frac{1}{2}} = \frac{1}{2}$$

So we have:

$$P[\text{success}] \ge (1 - \frac{1}{2}) \cdot (1 - \frac{1}{2}) = \frac{1}{4} = 0.25$$

### 3.2 How to Sample

We randomly sample n-1 vectors from  $S^{\perp}$ . But how can we sample from  $S^{\perp}$  without knowing  $\vec{s}$  in the first place? This step cannot be carried out classically in a straightforward way, it requires a quantum algorithm!

## 4 Generalized Simon's Problem

Consider a black-box function

$$f: \{0,1\}^n \to \{0,1\}^n$$

which implicitly defines a hidden subspace  $S \subseteq \mathbb{Z}_2^n$ . The promise is that for all  $\vec{x}, \vec{y} \in \mathbb{Z}_2^n$ ,

$$f(\vec{x}) = f(\vec{y}) \iff \left(\vec{x} = \vec{y} \text{ or } \vec{x} - \vec{y} \in S\right).$$

Goal. The task is to determine a basis for the subspace S using as few (quantum) queries to f as possible.

#### Result.

• The number of quantum queries required is

$$O(n - \dim(S)).$$

• This follows the same intuition we get from the previous Bridge Problem.