

QUANTUM LEARNING SEMINAR

LECTURE 3: SINGLE QUERY LEARNING

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Introduction

As we noted in the first lecture, quantum computation is a relatively young subject, and quantum learning is even younger. The goal of this lecture is to explain that there is a conjecture about quantum learning dating back to '93. That is, 1893!

CONJECTURE (Hadamard [1]): For $N \equiv 0 \pmod{4}$ there is a concept class of size N such that quantum learning from membership queries has sample complexity 1.

Concept classes

In the last two lectures we introduced the concept class

$$\mathcal{G}^n = \{g_a : \mathbb{Z}_N \rightarrow \mathbb{Z}_2 \mid a \in \mathbb{Z}_N \text{ and } g_a(x) = \delta_{xa}\},$$

and found that Grover's algorithm [3] provides a learning algorithm with sample complexity $O(\sqrt{N})$ for concepts in \mathcal{G}^n . Figure 3.1 shows a typical concept in \mathcal{G}^n for the case $N = 8$ ($n = 3$).

Now consider a different concept class:

$$\mathcal{BV}^n = \{f_a : \mathbb{Z}_2^n \rightarrow \mathbb{Z}_2 \mid a \in \mathbb{Z}_2^n \text{ and } f_a(x) = a \cdot x \pmod{2}\},$$

(named after Bernstein and Vazirani who first investigated this set in the context of quantum computing [4]). Figure 3.2 shows the first four concepts in \mathcal{BV}^3 . Notice that except for $a = 0$, $|f_a^{-1}(1)| = N/2$, in contrast to Grover concepts each of which contains exactly 1 element. There are still $N = 2^n$ concepts in \mathcal{BV}^n , just as there are in \mathcal{G}^n . It is easy to see, however, that the classical complexity of learning from \mathcal{BV}^n is $O(\log N)$. A learning algorithm can exploit the *structure* of \mathcal{BV}^n , which is best illustrated not by graphing the

concepts as in Figure 3.2, but as in Figure 3.3 where the possible inputs are the vertices of a (hyper)cube. The membership oracle need only be queried about the n basis vectors of \mathbb{Z}_2^n . A deterministic algorithm that does this, in the format of Figure 1.2 is

Algorithm DBVⁿ.

0. Set $X \leftarrow \mathbb{Z}_2^n$.
1. Set $x \leftarrow 1 = 0 \dots 01 \in \mathbb{Z}_2^n$.
2. While $x < 2^n$,
 - 2a. Set hypothesis to f_x .
 - 2b. Evaluate $f_x(x)$.
 - 2c. Query the membership oracle about x .
 - 2d. If $f_x(x) = \text{MO}(x)$ then adjust $X \leftarrow X \cap f_x^{-1}(1)$ else adjust $X \leftarrow X \cap f_x^{-1}(0)$.
 - 2e. Adjust $x \leftarrow 2x$.
3. Set $\{x\} \leftarrow X$; output f_x .

This algorithm identifies the bits of a one by one, reducing the set X of possible a s by half as each bit is identified. After n iterations of step **2**, X consists of a single element, a , which determines the concept learned, f_a .

Quantum learning

Classically \mathcal{BV}^n is a much easier concept class from which to learn than is \mathcal{G}^n . This is also true quantum mechanically: The first step is to prepare the same query state (1.1) as in Grover's algorithm. Submitting it to the membership oracle causes the unitary transformation:

$$\begin{aligned} \sum_x \frac{1}{\sqrt{N}} |x\rangle \frac{1}{\sqrt{2}} (|0\rangle - |1\rangle) &\mapsto \sum_x \frac{1}{\sqrt{N}} |x\rangle (-1)^{f_a(x)} \frac{1}{\sqrt{2}} (|0\rangle - |1\rangle) \\ &= \sum_x \frac{1}{\sqrt{N}} |x\rangle (-1)^{a \cdot x} \frac{1}{\sqrt{2}} (|0\rangle - |1\rangle). \end{aligned} \quad (3.1)$$

The first tensor factor of the resulting vector is one of $N = 2^n$ different vectors, depending on the concept f_a . For $N = 8$ they are the columns of the matrix

$$A = \frac{1}{\sqrt{8}} \begin{matrix} a = & 000 & 001 & 010 & 011 & 100 & 101 & 110 & 111 \\ \begin{pmatrix} + & + & + & + & + & + & + & + \\ + & - & + & - & + & - & + & - \\ + & + & - & - & + & + & - & - \\ + & - & - & + & + & - & - & + \\ + & + & + & + & - & - & - & - \\ + & - & + & - & - & + & - & + \\ + & + & - & - & - & - & + & + \\ + & - & - & + & - & + & + & - \end{pmatrix}, \end{matrix} \quad (3.2)$$

where $+$ and $-$ denote $+1$ and -1 , respectively.

It is easy to check that the columns of A are orthogonal for any N : $A^\dagger A = I$. Thus these vectors define an orthonormal basis and hence, according to the definition from the previous lecture, a projective measurement on \mathbb{C}^N . Since the first tensor factor on the right hand side of (3.1) is exactly one of these basis vectors (depending on the concept f_a), a projective measurement in this basis will observe a with probability 1. Thus, for concepts in \mathcal{BV}^n , the sample complexity of quantum learning from membership queries is 1.

Hadamard matrices

This is clearly equally true for any concept class consisting of concepts that are *orthogonal* in the sense that the matrix corresponding to (3.2) is orthogonal. The first part of the following definition is standard; the second part may not be.

DEFINITION. An $N \times N$ matrix H with elements in $\{\pm 1\}$ that satisfies $H^T H = NI$ (and consequently also $HH^T = NI$) is called a *Hadamard matrix*. Correspondingly, if a concept class \mathcal{C} of maps $X \rightarrow \mathbb{Z}_2$ satisfies

$$\sum_{x \in X} (-1)^{c(x)} (-1)^{c'(x)} = 0$$

for all $c \neq c' \in \mathcal{C}$, we will call \mathcal{C} a *Hadamard concept class*.

It is an immediate consequence of our definitions that any Hadamard concept class has sample complexity 1 for quantum learning from a membership oracle. As we will see in a subsequent lecture, the classical sample complexity is $\Omega(\log N)$, as it is for \mathcal{BV}^n . These statements are only interesting, of course, if there are Hadamard concept classes other than the family \mathcal{BV}^n . These are not so easy to find, but mathematicians have been looking for the corresponding Hadamard matrices since 1867 when Sylvester discovered the family $\begin{pmatrix} + & + \\ + & - \end{pmatrix}^{\otimes n}$ [5]—corresponding to \mathcal{BV}^n . Some time afterwards Hadamard proved:

THEOREM (Hadamard [2]). *If H is an $N \times N$ Hadamard matrix, then $N \in \{1, 2\}$ or $N \equiv 0 \pmod{4}$.*

Proof. Multiplying any row or column by -1 leaves a Hadamard matrix Hadamard. By doing so appropriately we can change the first row and first column to all $+1$ s; a Hadamard matrix in this form is called *normalized*. Permuting the columns of a Hadamard matrix also leaves it Hadamard. Thus we can arrange the columns so that the first three rows have the form

$$\begin{array}{cccccccc} + & \text{-----} & & & & & & + \\ + & \text{-----} & + & - & \text{-----} & & & - \\ + & - & + & - & - & - & + & - & + & - & - & - \\ \underbrace{\hspace{1.5cm}}_{i+s} & \underbrace{\hspace{1.5cm}}_{j-s} & \underbrace{\hspace{1.5cm}}_{k+s} & \underbrace{\hspace{1.5cm}}_{l-s} \end{array}$$

Since these rows are mutually orthogonal,

$$\begin{aligned}i + j - k - l &= 0 \\i - j + k - l &= 0 \\i - j - k + l &= 0,\end{aligned}$$

which implies $i = j = k = l$. Thus $N = 4k$, if $N \geq 3$. ■

This result suggested the conjecture with which we started this lecture; the standard phrasing, of course, is:

CONJECTURE (Hadamard [1]). *There is an $N \times N$ Hadamard matrix for all $N \equiv 0 \pmod{4}$.*

This conjecture is sometimes attributed to Paley [6], but in 1893 Hadamard wrote: “*J’ai formé des déterminants réels pour $n = 12$ et $n = 20$, sans avoir pu néanmoins reconnaître d’une façon certaine s’il en existe chaque fois que n est divisible par 4.*” [1], so it is clear that he thought it might be true. As of 2002, the smallest multiple of 4 for which no Hadamard matrix is known is $N = 428$ [7]. Since the Sylvester matrix $\begin{pmatrix} + & + \\ + & - \end{pmatrix}^{\otimes n}$ has dimension 2^n , this implies that there must be other ways to construct Hadamard matrices. We describe the simplest, which produces the $N = 12$ matrix found by Hadamard (I haven’t checked his example for $N = 20$.), as the smallest of an infinite family of matrices different from the Sylvester matrices, next.

Paley’s construction

We begin with some number theoretic preliminaries. Let $p > 2$ be a prime number.

DEFINITION. Let $0 < a \in \mathbb{Z}$. The elements of $\{a^2 \pmod{p}\}$ are called *quadratic residues mod p* .

To compute the number of quadratic residues, note that since

$$(a + p)^2 = a^2 + 2ap + p^2 \equiv a^2 \pmod{p},$$

we need only consider $0 \leq a < p$. Furthermore, since

$$(p - a)^2 = p^2 - 2pa + a^2 \equiv a^2 \pmod{p},$$

we need only consider $0 \leq a \leq \frac{p-1}{2}$. But $1^2, 2^2, \dots, \left(\frac{p-1}{2}\right)^2 \pmod{p}$ are all distinct since

$$\begin{aligned}a^2 \equiv b^2 \pmod{p} &\implies p \mid a^2 - b^2 = (a + b)(a - b) \\ &\implies a = b.\end{aligned}$$

Therefore there are $\frac{p-1}{2}$ quadratic residues mod p . The other $\frac{p-1}{2}$ nonzero elements of \mathbb{Z}_p are called *nonresidues*. 0 is neither a residue nor a nonresidue.

DEFINITION. The *Legendre symbol* is defined to be

$$\left(\frac{a}{p}\right) = \begin{cases} +1 & \text{if } a \text{ is a residue;} \\ -1 & \text{if } a \text{ is a nonresidue;} \\ 0 & \text{if } a = 0. \end{cases}$$

DEFINITION. For $p \equiv -1 \pmod{4}$, the *Jacobsthal matrix* is the $p \times p$ matrix Q with elements $q_{ij} = \left(\frac{j-i}{p}\right)$, for $i, j \in \{0, \dots, p-1\}$.

EXAMPLE. For $p = 11$,

$$Q = \begin{pmatrix} 0 & + & - & + & + & + & - & - & - & + & - & - \\ - & 0 & + & - & + & + & + & - & - & - & + & - \\ - & - & 0 & + & - & + & + & + & - & - & - & + \\ + & - & - & 0 & + & - & + & + & + & - & - & - \\ - & + & - & - & 0 & + & - & + & + & + & - & - \\ - & - & + & - & - & 0 & + & - & + & + & + & - \\ + & - & - & - & + & - & - & 0 & + & - & + & + \\ + & + & - & - & - & + & - & - & 0 & + & - & + \\ + & + & + & - & - & - & + & - & - & 0 & + & - \\ - & + & + & + & - & - & - & + & - & - & 0 & + \\ + & - & + & + & + & - & - & - & + & - & - & 0 \end{pmatrix}.$$

Notice that $Q^T = -Q$. This is true for every Jacobsthal matrix; it is a consequence of the Legendre symbol being a character. Another consequence is that

$$QQ^T = -Q^2 = pI - \mathbf{1}\mathbf{1}^T,$$

where $\mathbf{1}$ is the $p \times 1$ vector of all +1s.

Now let H be the $p+1 \times p+1$ matrix:

$$H = \begin{pmatrix} 1 & \mathbf{1}^T \\ \mathbf{1} & Q - I \end{pmatrix}. \quad (3.3)$$

This is Paley's (Type I) construction [6], which was apparently discovered by Gilman a few years earlier [8]. It produces a Hadamard matrix, since

$$HH^T = \begin{pmatrix} p+1 & \mathbf{0}^T \\ \mathbf{0} & \mathbf{1}\mathbf{1}^T + (Q - I)(Q - I)^T \end{pmatrix},$$

where $\mathbf{0}$ is the $p \times 1$ zero vector. The lower right block is

$$\mathbf{1}\mathbf{1}^T + QQ^T - Q^T - Q + I = \mathbf{1}\mathbf{1}^T + pI - \mathbf{1}\mathbf{1}^T + I = (p+1)I,$$

where we have used the two facts about Jacobsthal matrices noted in the Example above. Thus $HH^T = (p + 1)I$, which means H is a Hadamard matrix.

The $p = 11$ Jacobsthal matrix of the Example above produces the unique 12×12 Hadamard matrix by this construction. For $p = 3$ and $p = 7$, Paley’s/Gilman’s construction reproduces the 4×4 and 8×8 Sylvester matrices. For $p = 31$, however, it produces a 32×32 Hadamard matrix that is distinct from the 32×32 Sylvester matrix. The number of distinct Hadamard matrices is known currently only up to $N = 28$ [9]:

$$\begin{array}{rcccccccc} N = & 1 & 2 & 4 & 8 & 12 & 16 & 20 & 24 & 28 \\ \# = & 1 & 1 & 1 & 1 & 1 & 5 & 3 & 60 & 487 \end{array}$$

Each distinct Hadamard matrix defines a Hadamard concept class in which quantum learning with access to a membership oracle succeeds with sample complexity 1. The concept class \mathcal{BV}^n corresponding to the $2^n \times 2^n$ Sylvester matrices consists of concepts of the form “numbers x such that $x \cdot a = 1 \pmod{2}$ ”. If the rows and columns of (3.3) are labelled from 0 to p , the corresponding concept class is “numbers x such that $x \neq 0$ and $x - a$ is a square mod p for $a \neq 0$ ”. In a paper that inspired this lecture, van Dam has combined the Paley/Gilman construction and Paley’s (Type II) construction for primes $p \equiv 1 \pmod{4}$ into the *Shifted Legendre Symbol* problem and proved that it has sample complexity 2 [10].

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