## Lecture 15

**Lemma 6.** Let  $x \in \{0,1\}^k$  and  $|x\rangle = |x_1\rangle \dots |x_k\rangle$  be a k-qubit state. Then

$$H^{\otimes k} |x\rangle = \frac{1}{\sqrt{2^k}} \sum_{y \in \{0,1\}^k} (-1)^{x \cdot y} |y\rangle, \qquad (98)$$

where  $H^{\otimes k} := H \otimes \cdots \otimes H$  (k times) and  $x \cdot y := \sum_{i=1}^k x_i y_i$ .

*Proof.* We have

$$H^{\otimes k} |x\rangle = H |x_1\rangle \otimes \cdots \otimes H |x_n\rangle$$

$$= \frac{1}{\sqrt{2}} (|0\rangle + (-1)^{x_1} |1\rangle) \otimes \cdots \otimes \frac{1}{\sqrt{2}} (|0\rangle + (-1)^{x_1} |1\rangle) \qquad \text{Eq. (39) (Phase kickback)}$$

$$= \frac{1}{\sqrt{2^k}} \sum_{y_1, \dots, y_k \in \{0,1\}} (-1)^{x_1 y_1 + \dots + x_k y_k} |y_1\rangle |y_2\rangle \dots |y_k\rangle \qquad \text{think about phase for fixed } y$$

$$= \frac{1}{\sqrt{2^k}} \sum_{y \in \{0,1\}^k} (-1)^{x \cdot y} |y\rangle ,$$

as required.

**Lemma 7.** Let  $K \in \mathbb{N}$ . Suppose  $z_1, \ldots, z_K \leftarrow \mathbb{F}_2^k$ . Then the probability that the dimension of the span of the  $z_i s$ , i.e., the dimension of the subspace

$$V := \{a_1 z_1 + \dots + a_K z_K \mid a_1, \dots, a_K \in \mathbb{F}_2\} \le \mathbb{F}_2^k \tag{99}$$

is k is at least  $1 - 2^{k-K}$ .

Based on [StackExchange post].

*Proof.* Let  $A \in \mathbb{F}_2^{K \times k}$  denote the matrix whose rows are the  $z_i$ s. The dimension of V is the same as the row-rank (dimension of the span of the rows) of A, which is equal to the column-rank of A by a standard fact in linear algebra. Now, the column-rank of A is k if and only if the kernel of A is  $\{0\}$  by the rank-nullity theorem, where the kernel of A is defined by

$$\ker(A) := \{ x \in \mathbb{F}_2^k \mid Ax = 0 \}. \tag{100}$$

Since the  $z_i$ s are chosen uniformly from  $\mathbb{F}_2^k$ , A is a uniformly random matrix in  $\mathbb{F}_2^{K \times k}$ . In the following, the probability is over  $A \leftarrow \mathbb{F}_2^{K \times k}$ .

$$\begin{split} \Pr[\ker(A) \neq \{0\}] &= \Pr[\exists x \in \mathbb{F}_2^k, x \neq 0, Ax = 0] & \text{definition} \\ &\leq \sum_{x \in \mathbb{F}_2^k, x \neq 0} \Pr[Ax = 0] & \text{union bound} \\ &= (2^k - 1) \frac{1}{2^K} & Ax \text{ is unif. random in } \mathbb{F}_2^K, \text{ e.g., suppose } x_k = 1 \\ &\leq \frac{2^k}{2^K}. \end{split}$$

Therefore  $\Pr[\dim(V) = k] = \Pr[\ker(A) = \{0\}] \ge 1 - 2^{k-K}$ .

**Lemma 8.** Let  $K \in \mathbb{N}$  and  $0 \neq a \in \mathbb{F}_2^k$ . Let  $z_1, \ldots, z_K \in \mathbb{F}_2^k$  (arbitrary) be such that  $\forall i \in [K]$ ,  $a \cdot z_i = 0 \mod 2$ . Then the dimension of the span of the  $z_i s$  is at most k-1.

*Proof.* It suffices to prove that the dimension of the following subspace is k-1:

$$U := \{ z \in \mathbb{F}_2^k \mid a \cdot z = 0 \}. \tag{101}$$

Note that U is the kernel of the  $1 \times k$  matrix  $A := (a_1, \dots, a_k)$ . Now, the column-rank of A is 1 since  $a \neq 0$ . Therefore, by the rank-nullity theorem,  $\dim(U) = k - 1$ .

**Remark 8.** In the case  $x \in D_1$ , a slight modification of the algorithm above can also recover a: choose K large enough (how large?) such that in the case  $x \in D_1$ , we have d = k-1 whp; collect the k-1 linearly independent vectors  $z^{(1)}, \ldots z^{(k-1)} \in \mathbb{F}_2^k$  into the rows of a matrix  $A \in \mathbb{F}_2^{(k-1) \times k}$  and compute the kernel of A, which will have size a. a is the non-zero element. Moreover, note that, since  $a = 2^k$ , we can identify  $\{0, 1, \ldots, n-1\}^n$  with  $\{0, 1, \ldots, n-1\}^{\frac{n}{2}}$ .

Therefore, we also have an  $O(\log(n))$  quantum algorithm for the following query problem:

$$\operatorname{Simon}_{n}' : D' \subseteq \{0, 1, \dots, n-1\}^{\mathbb{F}_{2}^{k}} \to \mathbb{F}_{2}^{k}$$

$$(102)$$

where  $x \in D'$  if and only if there exists an  $a \in \mathbb{F}_2^k - \{0^k\}$  such that  $\forall s, t \in \mathbb{F}_2^k$ ,  $x(s) = x(t) \iff s \in \{t, t+a\}$  (addition as defined in the group  $\mathbb{Z}_2^k$ , i.e., component-wise addition), and  $\operatorname{Simon}_n'(x)$  outputs the a (period) associated with x. (Writing it this way is to allow for direct comparison with the order finding problem at the heart of Shor's algorithm later.)

**Proposition 8.**  $Q(\operatorname{Simon}_n) = \Theta(\sqrt{n})$ 

*Proof.* Upper bound Randomized query algorithm for computing Collision<sub>n</sub>. n is even.

Note that the following description can be formally phrased in terms of a distribution over decision trees (how?). Given input  $x \in D_0 \dot{\cup} D_1$ :

Sample a uniformly random subset  $\{i_1, \ldots, i_m\} \subseteq [n]$  of size m. Query  $x_{i_1}, \ldots, x_{i_m}$ , if there is a collision, i.e.,  $i_a \neq i_b$  with  $a, b \in [m]$ , such that  $x_{i_a} = x_{i_b}$ , then output 1, else output 0.

How large of a  $m \le n/2$  do we need to pick? (Note if m > n/2, guaranteed to find a collision.) If  $x \in D_0$ , then will never observe a collision, so always correct in this case. So the probability of error is the probability that no collision is observed if  $x \in D_1$ . Comment: first expression: for visual aid, consider a complete bipartite graph with n/2 vertices in each part.

$$\frac{n(n-2)(n-4)\dots(n-2(m-1))/m!}{\binom{n}{m}} = 1 \cdot \left(1 - \frac{1}{n-1}\right) \cdot \left(1 - \frac{2}{n-2}\right) \dots \left(1 - \frac{m-1}{n-m+1}\right) \\ \leq \exp\left(-\sum_{i=1}^{m-1} \frac{i}{n-i}\right) \leq \exp\left(-\sum_{i=1}^{m-1} \frac{i}{n}\right) = \exp\left(-\frac{m(m-1)}{2n}\right) \leq \exp\left(-\frac{(m-1)^2}{2n}\right).$$

Therefore the probability of error is  $\leq \epsilon$  if

$$\exp\left(-\frac{(m-1)^2}{2n}\right) \le \epsilon \iff m \ge \sqrt{2n\ln(1/\epsilon)} + 1. \tag{103}$$

Therefore,  $R_{\epsilon}(\text{Collision}_n) \leq \min(\sqrt{2n\ln(1/\epsilon)} + 1, n/2)$ . So  $R(\text{Collision}_n) \leq O(\sqrt{n})$ .