

## Lecture 13

Comment: HW1 Q4: good to redo in Dirac notation if you used matrices because when there are more than two qubits (say, on exam), direct matrix manipulation is really inefficient. HW1 Q2 interpretation: *non-adaptive* deterministic query complexity of NAND tree on  $n$  variables is  $n$ . If adaptive, answer is still  $\Omega(n)$ , but need an enhanced argument. Consider  $(x_1 \wedge x_2) \vee (\neg x_1 \wedge x_3)$ , which needs 3 non-adaptive queries but only 2 adaptive ones.

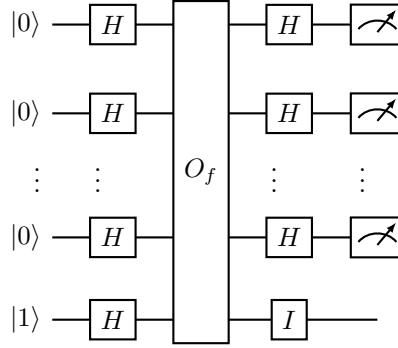
Recall the Deutsch-Jozsa problem: given  $f: \{0,1\}^n \rightarrow \{0,1\}$ , ask balanced or constant. Classically, need  $2^{n-1} + 1$  queries  $f$  to solve with certainty. Quantumly, it turns out that only 1 (quantum) query to  $f$  suffices.

Recall the quantum oracle for  $f$  is given by

$$O_f |x\rangle |b\rangle \rightarrow |x\rangle |b \oplus f(x)\rangle, \quad (74)$$

where  $x \in \{0,1\}^n, b \in \{0,1\}$ .

The quantum circuit for the Deutsch-Jozsa algorithm is shown below:



Output “constant” if the measurement outcome is  $0^n$  and “balanced” if the measurement outcome is not  $0^n$ .

To make the analysis easier, we will change the  $I$  to an  $H$  (which does not affect the measurement distribution as per a lemma from 2 lectures ago), introduce the quantum phase oracle and give a general expression for how multiple Hadamard gates act on a computational basis state.

**Definition 13.** The quantum phase oracle for  $f$  is given by

$$U_f |x\rangle |b\rangle = (-1)^{b \cdot f(x)} |x\rangle |b\rangle, \quad (75)$$

where  $x \in \{0,1\}^n, b \in \{0,1\}$ .

The following lemma will be used in the next two propositions.

**Lemma 1.** For  $b \in \{0,1\}$ , we have

$$H |b\rangle = \frac{1}{\sqrt{2}}(|0\rangle + (-1)^b |1\rangle). \quad (76)$$

*Proof.* By direct calculation. □

**Proposition 5** (Phase kickback trick). We have  $U_f = (\mathbb{1}_{2^n} \otimes H) O_f (\mathbb{1}_{2^n} \otimes H)$ .

*Proof.* Write  $\mathbb{1}$  for  $\mathbb{1}_{2^n}$ . Then

$$\begin{aligned} |x\rangle |b\rangle &\xrightarrow{\mathbb{1} \otimes H} |x\rangle \frac{1}{\sqrt{2}}(|0\rangle + (-1)^b |1\rangle) && \text{by Eq. (76)} \\ &\xrightarrow{O_f} \frac{1}{\sqrt{2}} |x\rangle (|f(x)\rangle + (-1)^b |f(x) \oplus 1\rangle) \\ &\xrightarrow{\mathbb{1} \otimes H} \frac{1}{\sqrt{2}} |x\rangle \left( \frac{1}{\sqrt{2}}(|0\rangle + (-1)^{f(x)} |1\rangle) + (-1)^b \frac{1}{\sqrt{2}}(|0\rangle + (-1)^{f(x) \oplus 1} |1\rangle) \right) \\ &= \frac{1}{2} |x\rangle ((1 + (-1)^b) |0\rangle + (-1)^{f(x)} (1 - (-1)^b) |1\rangle) \\ &= (-1)^{f(x) \cdot b} |x\rangle |b\rangle, \end{aligned}$$

as required. □

**Proposition 6.** For all  $x \in \{0, 1\}^n$ ,

$$H^{\otimes n} |x\rangle = \frac{1}{\sqrt{2^n}} \sum_{y \in \{0, 1\}^n} (-1)^{x \cdot y} |y\rangle, \quad (77)$$

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

*Proof.* We have

$$\begin{aligned} H^{\otimes n} |x\rangle &= H |x_1\rangle \otimes \cdots \otimes H |x_n\rangle \\ &= \frac{1}{\sqrt{2}} (|0\rangle + (-1)^{x_1} |1\rangle) \otimes \cdots \otimes \frac{1}{\sqrt{2}} (|0\rangle + (-1)^{x_n} |1\rangle) \quad \text{by Eq. (76)} \\ &= \frac{1}{\sqrt{2^n}} \sum_{y \in \{0, 1\}^n} (-1)^{x \cdot y} |y\rangle, \end{aligned}$$

as required □

Analysis of the Deutsch-Jozsa algorithm.

$$\begin{aligned} |0^n\rangle |1\rangle &\xrightarrow{H^{\otimes n}} \frac{1}{\sqrt{2^n}} \sum_x |x\rangle |1\rangle \\ &\xrightarrow{U_f} \frac{1}{\sqrt{2^n}} \sum_x (-1)^{f(x)} |x\rangle |1\rangle \\ &\xrightarrow{H^{\otimes n}} \frac{1}{\sqrt{2^n}} \sum_x (-1)^{f(x)} \sum_y (-1)^{x \cdot y} |y\rangle |1\rangle \\ &= \frac{1}{2^n} \sum_y \left( \sum_x (-1)^{f(x) + x \cdot y} \right) |y\rangle |1\rangle \end{aligned}$$

Therefore the probability of measuring  $y = 0^n$  is

$$\Pr[0^n] = \frac{1}{2^{2n}} \left( \sum_x (-1)^{f(x)} \right)^2. \quad (78)$$

If  $f$  is constant, then  $\Pr[0^n] = 1$ . If  $f$  is balanced, then  $\Pr[0^n] = 0$ .

**Remark 2.** Does DJ mean we have a real (unconditional) provable exponential quantum speedup? **No.** To compute  $f$  in the real world, a description of  $f: \{0, 1\}^n \rightarrow \{0, 1\}$  as a circuit<sup>4</sup> must be given. But once we have such a description of  $f$ , we may be able to solve the problem faster than time  $2^{n-1} + 1$  by analyzing the circuit diagram rather than only evaluating (aka querying) the circuit. **Comment:** will say a bit more about this next lecture.

If we are *forced* to only query the circuit, then yes, DJ gives an exponential quantum speedup between quantum and randomized query complexity (if we also demand certain correctness). Being forced to *only* query the circuit is the essential assumption of the query model of computation. The real-world model of computation (aka Turing model) does not make this assumption, so DJ does *not* constitute a provable exponential quantum speedup in the real world.

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<sup>4</sup>If a circuit sounds abstract, think of this as a program or description. These become circuits once you compile them to run on hardware.