

# Quantum Computing 104: The Deutsch-Jozsa Algorithm

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March 24, 2025

Before diving into quantum algorithms, we first explored quantum gates, entanglement, and teleportation. Now, we transition into quantum algorithms by introducing the Deutsch-Jozsa algorithm, one of the first to demonstrate quantum advantage.

## 1 The Deutsch Algorithm

The Deutsch algorithm serves as a foundational proof-of-concept rather than a practical tool, providing a key step in quantum algorithm development. It serves as an important stepping stone in the development of quantum algorithms and provides a clear illustration of fundamental quantum principles.

### 1.1 Mathematical Formulation

Consider a Boolean function  $f : \{0, 1\} \rightarrow \{0, 1\}$ . The Deutsch problem asks whether  $f$  is constant ( $f(0) = f(1)$ ) or balanced ( $f(0) \neq f(1)$ ).

The Deutsch problem is solved using the Deutsch algorithm, which was later generalized to the Deutsch-Jozsa algorithm.

### 1.2 Quantum Oracle

The function  $f(x)$  is implemented as a quantum oracle, a unitary operator  $U_f$  that acts on quantum states. The oracle's action is defined as:

$$U_f |x\rangle |y\rangle = |x\rangle |y \oplus f(x)\rangle$$

where  $\oplus$  represents addition modulo 2 (XOR operation).

### 1.3 Superposition

The algorithm uses superposition to evaluate both  $f(0)$  and  $f(1)$  simultaneously.

### 1.4 Hadamard Gates

Hadamard gates generate and manipulate superposition states, enabling quantum parallelism. They are represented by the matrix:

$$H = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$$

And they act on a single qubit as:

$$H|0\rangle = \frac{1}{\sqrt{2}}(|0\rangle + |1\rangle)$$

$$H|1\rangle = \frac{1}{\sqrt{2}}(|0\rangle - |1\rangle)$$

## 1.5 Phase Kickback

The oracle's action introduces a phase factor that encodes the function's property (constant or balanced). Specifically, when the ancilla qubit is initialized to  $|-\rangle = \frac{1}{\sqrt{2}}(|0\rangle - |1\rangle)$ , the oracle acts as:

$$U_f|x\rangle|-\rangle = (-1)^{f(x)}|x\rangle|-\rangle$$

This phase factor  $(-1)^{f(x)}$  is crucial for the algorithm to distinguish between constant and balanced functions.

## 1.6 Algorithm Steps

1. Initialize two qubits:  $|0\rangle|1\rangle$ .
2. Apply Hadamard gates:  $H|0\rangle H|1\rangle$ .
3. Apply the quantum oracle  $U_f$ .
4. Apply another Hadamard gate to the first qubit.
5. Measure the first qubit.

## 1.7 Qiskit Implementation for the Deutsch algorithm

The following libraries are required.

Listing 1: required libraries

```
from qiskit import QuantumCircuit, transpile
from qiskit_aer import Aer
from qiskit_aer import AerSimulator
```

Listing 2: Qiskit Implementation for the Deutsch algorithm

```
def deutsch_algorithm(oracle_function):
    """ Implements the Deutsch Algorithm for a 1-bit function. """
    circuit = QuantumCircuit(2, 1)  # 1 input qubit + 1 ancilla qubit

    # Step 1: Initialize ancilla to  $|\text{ket}\{1\}$  and apply Hadamard
    circuit.x(1)
    circuit.h([0, 1])

    # Step 2: Apply the oracle
    oracle_function(circuit)

    # Step 3: Apply Hadamard again
```

```

circuit.h(0)

# Step 4: Measure the first qubit
circuit.measure(0, 0)

return circuit

# Define Oracle Functions
def constant_zero_oracle(circuit):
    """ Oracle where f(x) = 0 for all inputs (does nothing). """
    pass # No operation needed

def constant_one_oracle(circuit):
    """ Oracle where f(x) = 1 for all inputs (flips the ancilla). """
    circuit.x(1)

def balanced_x_oracle(circuit):
    """ Oracle where f(x) = x (CNOT gate applies conditional flipping). """
    circuit.cx(0, 1) # Apply CNOT from input to ancilla

def run_deutsch_algorithm(oracle_function):
    """ Runs the Deutsch Algorithm and prints the result. """
    circuit = deutsch_algorithm(oracle_function)
    simulator = AerSimulator()
    compiled_circuit = transpile(circuit, simulator)
    job = simulator.run(compiled_circuit, shots=1000)
    result = job.result().get_counts()
    output = "Constant" if "0" in result else "Balanced"
    print(f"Oracle: {oracle_function.__name__}, Result: {result}, Function Type: {output}")

# Run the Algorithm
print("\nDeutsch Algorithm Results:\n")
run_deutsch_algorithm(constant_zero_oracle) # Expect:
Constant
run_deutsch_algorithm(constant_one_oracle) # Expect:
Constant
run_deutsch_algorithm(balanced_x_oracle) # Expect:
Balanced

```

The output of the above code is

Deutsch Algorithm Results:

```

Oracle: constant_zero_oracle, Result: {'0': 1000}, Function Type: Constant
Oracle: constant_one_oracle, Result: {'0': 1000}, Function Type: Constant
Oracle: balanced_x_oracle, Result: {'1': 1000}, Function Type: Balanced

```

## 2 Mathematical Formulation of the Deutsch-Jozsa Problem

Consider a Boolean function  $f$  defined as below.

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

where  $\{0, 1\}^n$  represents the set of all  $n$  binary strings and  $0, 1$  represents the set of Boolean outputs (0 or 1) and the function  $f$  is guaranteed to be either:

- **Constant:** The function  $f$  yields the same output for all possible  $n$ -bit inputs. Formally, this can be expressed as:

$$f(x) = c, \quad \forall x \in \{0, 1\}^n, \quad \text{where } c \in \{0, 1\}$$

This means that either  $f(x) = 0$  for all  $x$ , or  $f(x) = 1$  for all  $x$ .

- **Balanced:** The function  $f$  produces an equal number of 0s and 1s across all possible  $n$ -bit inputs. This can be mathematically stated as:

$$\sum_{x \in \{0, 1\}^n} f(x) = 2^{n-1}$$

This implies that exactly  $2^{n-1}$  inputs result in  $f(x) = 0$ , and  $2^{n-1}$  inputs result in  $f(x) = 1$ .

**The Deutsch-Jozsa problem asks: Given the function  $f$  as above, determine whether it is constant or balanced.**

**Example Scenario:**

Consider a Boolean function  $f$  defined as below.

$$f : \{0, 1\}^3 \rightarrow 0, 1$$

where  $\{0, 1\}^3$  represents the set of all 3 binary strings  $\{000, 001, 010, 011, 100, 101, 110, 111\}$  and  $0, 1$  represents the set of Boolean outputs (0 or 1) and the function  $f$  is guaranteed to be either:

- **Constant:** The function  $f$  yields the same output for all possible 3-bit inputs.

$$f(x) = c, \quad \forall x \in \{0, 1\}^3, \quad \text{where } c \in \{0, 1\}$$

This means that either  $f(x) = 0$  for all  $x$ , or  $f(x) = 1$  for all  $x$ .

- **Balanced:** The function  $f$  produces an equal number of 0s and 1s across all possible 3-bit inputs. This can be mathematically stated as:

$$\sum_{x \in \{0, 1\}^3} f(x) = 2^2 = 4$$

This implies that exactly 4 inputs result in  $f(x) = 0$ , and 4 inputs result in  $f(x) = 1$ .

**The Challenge: How Do We Figure Out the Type of Function?**

With a classical computer, the worst case requires checking more than half the inputs up to  $2^{n-1} + 1$  evaluations.

- If we only get 0s or only 1s, the function is constant.
- If we get both 0 and 1 as outputs, the function is balanced.

This means that for large values of  $n$ , checking enough inputs to be sure takes a long time.

### Quantum Computation Advantage

Quantum parallelism uses superposition to evaluate a function  $f(x)$  for all possible inputs at the same time.

The Deutsch-Jozsa algorithm solves this problem in a single function evaluation using quantum parallelism.

By encoding all inputs into a superposition state and applying interference, we extract the answer in one quantum measurement.

## 2.1 Superposition and Hadamard Transformation

The Hadamard transformation generates an equal superposition of all basis states.

The Hadamard gate  $H$  acts on a single qubit as:

$$H|x\rangle = \frac{1}{\sqrt{2}} \sum_{z=0}^1 (-1)^{x \cdot z} |z\rangle$$

where  $x, z \in \{0, 1\}$ .

The Hadamard gate  $H$  acts on  $n$ -qubits as:

$$H^{\otimes n}|x\rangle = \frac{1}{\sqrt{2^n}} \sum_{z=0}^{2^n-1} (-1)^{x \cdot z} |z\rangle$$

where

- $x, z$  are now  $n$ -bit binary numbers.
- $x \cdot z$  is the dot product (binary dot product) (bitwise AND followed by XOR sum).

The Hadamard transformation prepares our system in a uniform superposition, but on its own, it does not provide any information about the function  $f(x)$ . To introduce function-dependent behavior, we use a quantum oracle, which encodes  $f(x)$  into the quantum state via a unitary transformation.

## 2.2 Quantum Oracle

Before introducing the Deutsch-Jozsa Algorithm, let's first understand quantum oracles.

1. **black-box-** In quantum computing, a black-box refers to a system where we can query a function without knowing its internal workings. We only observe its input-output behavior but not how it is implemented. For example,
  - A vending machine functions as follows: an input (button press) results in an output (drink dispensed), while the internal operational mechanisms remain unknown.
  - A password verification system processes an input password, producing an output of either access granted or access denied, without revealing its internal verification process.
2. **Quantum Oracle or Quantum Black-Box-** A quantum oracle is the quantum equivalent of a classical black-box function. Instead of directly computing  $f(x)$ , it encodes it into a unitary transformation  $U_f$ , allowing us to process multiple inputs in superposition simultaneously.  
Mathematically, the quantum oracle acts as:

$$U_f |x\rangle |y\rangle = |x\rangle |y \oplus f(x)\rangle$$

where:

- $|x\rangle$  is the input register (unchanged).
- $|y\rangle$  is an auxiliary qubit that stores  $f(x)$ .
- $\oplus$  is modulo-2 addition (XOR), which ensures reversibility.

When the quantum oracle  $U_f$  acts on a superposition, it effectively applies the function  $f(x)$  to all inputs  $x$  in the superposition simultaneously.

This is the Quantum parallelism.

**Unlike a classical function call, a quantum oracle can evaluate multiple inputs at once due to superposition, enabling quantum parallelism.**

## 2.3 Step-by-Step Explanation of the Deutsch-Jozsa Algorithm

The Deutsch-Jozsa algorithm is a quantum algorithm that determines whether a given function  $f(x)$  is constant or balanced in just one function evaluation.

### 1. Initialize the Quantum Register

We start with  $n + 1$ - qubit system.

- The first  $n$  qubits are initialized in the  $|0\rangle$  state. The first  $n$  qubits are used for input  $x$ .
- The last qubit (ancilla- this is auxiliary ) is initialized to  $|1\rangle$ . The last qubit is used for function evaluation.

At this stage our system is  $|\psi_{system}\rangle = |0\rangle^n \otimes |1\rangle$ .

### 2. Superposition and Hadamard transformation

We apply a Hadamard gate  $H^{\otimes(n+1)}$  to create a superposition of all possible inputs:

- $H^{\otimes(n)}|0\rangle^{\otimes(n)} = \frac{1}{\sqrt{2^n}} \sum_{x=0}^{2^n-1} |x\rangle$ .
- $H|1\rangle = \frac{1}{\sqrt{2}} (|0\rangle - |1\rangle)$ .

So total state after applying Hadamard gate is,

$$H(|\psi_{system}\rangle) = \left( \frac{1}{\sqrt{2^n}} \sum_{x=0}^{2^n-1} |x\rangle \right) \otimes \left( \frac{1}{\sqrt{2}} (|0\rangle - |1\rangle) \right) = \frac{1}{\sqrt{2^{n+1}}} \sum_{x=0}^{2^n-1} |x\rangle (|0\rangle - |1\rangle)$$

### 3. Apply the Function $f(x)$ using an Oracle

The (Boolean) function  $f(x)$  is implemented using a quantum oracle  $U_f$  which acts as

$$U_f |x\rangle |y\rangle = |x\rangle |y \oplus f(x)\rangle$$

where:

- $|x\rangle$  is the input register (unchanged).
- $|y\rangle = |0\rangle - |1\rangle$  is an auxiliary qubit that stores  $f(x)$ .

- $\oplus$  is modulo-2 addition (XOR), ensuring reversibility.

[Note: This is equivalent to the more general oracle action when the ancilla qubit is initialized to  $|-\rangle = \frac{1}{\sqrt{2}}(|0\rangle - |1\rangle)$ :

$$U_f |x\rangle |-\rangle = |x\rangle |-\oplus f(x)\rangle = (-1)^{f(x)} |x\rangle |-\rangle$$

where:

- $|x\rangle$  is the input register (unchanged).
- $f(x)$  is the Boolean output of the function.

] If  $f(x) = 0$  then  $|y \oplus f(x)\rangle = |y\rangle$  and if  $f(x) = 1$  then  $|y \oplus f(x)\rangle = -|y\rangle$  and so

$$|y \oplus f(x)\rangle = (-1)^{f(x)} |y\rangle.$$

After applying a quantum oracle  $U_f$  to  $H(|\psi_{system}\rangle) = \frac{1}{\sqrt{2^{n+1}}} \sum_{x=0}^{2^n-1} |x\rangle (|0\rangle - |1\rangle)$ , we get  $\frac{1}{\sqrt{2^{n+1}}} \sum_{x=0}^{2^n-1} (-1)^{f(x)} |x\rangle (|0\rangle - |1\rangle)$ .

Since the second qubit ( $|0\rangle - |1\rangle$ ) is now independent, we can ignore it and focus on the first register:

$$\frac{1}{\sqrt{2^n}} \sum_{x=0}^{2^n-1} (-1)^{f(x)} |x\rangle$$

This **adds a phase factor**  $(-1)^{f(x)}$  to each basis state  $|x\rangle$ .

#### 4. Apply Hadamard Transformation Again

Now, we apply the Hadamard gate to the first  $n$  qubits again:

$$\frac{1}{\sqrt{2^n}} \frac{1}{\sqrt{2^n}} \sum_{z=0}^{2^n-1} \sum_{x=0}^{2^n-1} (-1)^{x \cdot z + f(x)} |z\rangle$$

where

- $x, z$  are now  $n$ -bit binary numbers.
- $x \cdot z$  is the dot product (bitwise AND followed by XOR sum).

#### 5. Analysis of the result

If  $f(x)$  is a constant:

- The phase factor  $(-1)^{f(x)}$  is the same for all  $x$ .
- The final state is  $|0\rangle^{\otimes n}$ .

If  $f(x)$  is a balanced:

- The phase factor  $(-1)^{f(x)}$  varies, leading to interference.
- The final measurement will never yield  $|0\rangle^{\otimes n}$  because of destructive interference.

#### 6. Measure the First $n$ Qubits

- If we measure all 0s (000...0), the function is constant.
- If we measure any state other than  $|0\rangle^{\otimes n}$ , the function is balanced.

Thus, the Deutsch-Jozsa algorithm determines whether  $f(x)$  is constant or balanced in just one function call!

### 3 Qiskit Implementation of the Deutsch-Jozsa algorithm

We implement the above algorithm in Qiskit.

#### 3.1 Initialize the Quantum Register, Hadamard Gates and Oracle functions

For  $n$ -qubit system, we need first  $n$ -qubits initialized to  $|0\rangle$  state and  $n + 1$ -th qubit is ancilla qubit which we need to be initialized to state  $|1\rangle$ .

Listing 3: Initialize the Quantum Register first Hadamard Gates Oracle functions

```
def create_deutsch_jozsa_circuit(oracle_function, n):
    """ Creates Deutsch-Jozsa circuit with n input qubits """
    qr = list(range(n + 1)) # Create a list of qubit indices
    circuit = QuantumCircuit(n+1, n) # n input qubits + 1
        ancilla

    # Initialize the ancilla qubit in  $|\text{ket}\{1\}$  state
    circuit.x(qr[n])

    # Apply Hadamard to all qubits
    for i in range(n + 1):
        circuit.h(qr[i])

    # Apply the oracle
    oracle_function(circuit, qr, n)

    # Apply Hadamard again to the input qubits
    for i in range(n):
        circuit.h(qr[i])

    # Measure all input qubits
    circuit.measure(qr[:n], range(n))

    return circuit
```

So here we have covered steps 1 to 4 of the algorithm.

#### 3.2 Oracle functions

In Deutsch-Jozsa Algorithm, we need to determine whether the function is constant (always 0 or 1) or balanced (equal number of 0s and 1s).

1. constant\_zero\_oracle  $f(x) = 0$  for all  $x$

This oracle function returns 0 for all inputs. The oracle does nothing to the circuit.

Listing 4: constant\_zero\_oracle

```
def constant_zero_oracle(circuit, qr, n):
    """ Oracle for  $f(x) = 0$  (constant function returning
        always 0) """
    pass # Do nothing
```



2. `constant_one_oracle`  $f(x) = 1$  for all  $x$   
 Note that Quantum circuits do not "return" values like classical functions. In Deutsch-Jozsa setup, the last qubit or ancilla qubit is initialized to  $|1\rangle$ .  $X$ -gate on ancilla qubit flips  $|0\rangle$  to  $|1\rangle$ .

Listing 5: `constant_one_oracle`

```
def constant_one_oracle(circuit, qr, n):
    """ Oracle for  $f(x) = 1$  (constant function returning
        always 1) """
```

3. `balanced_xor_oracle` (`balanced_identity_oracle`)

This oracle function

- Computes XOR of all input bits and stores the result in the last qubit (ancilla).
- Uses CNOT gates from each input qubit to the ancilla.

Listing 6: `balanced_xor_oracle`

```
def balanced_identity_oracle(circuit, qr, n):
    """ Balanced oracle where  $f(x) = x_1 \text{ XOR } x_2 \text{ XOR } \dots
        \text{ XOR } x_n$  """
    for i in range(n):
        circuit.cx(qr[i], qr[n]) # Apply CNOT from
                                # input qubits to ancilla
```

4. `balanced_random_oracle` (`random_balanced_oracle`)

This oracle function

- Randomly flips half of the possible inputs.
- Uses  $X$  gates on half of the input qubits before and after applying CNOT gates.

Listing 7: `balanced_random_oracle`

```
def balanced_random_oracle(circuit, qr, n):
    """ Balanced oracle that flips half of the possible
        inputs """
    for i in range(n // 2): # Flip half of the
        qubits randomly
        circuit.x(qr[i])
    for i in range(n):
        circuit.cx(qr[i], qr[n])
    for i in range(n // 2):
        circuit.x(qr[i])
```

### 3.3 Simulation

The following code demonstrates the required simulation for algorithm.

Listing 8: Simulation

```
def run_deutsch_jozsa_algorithm(oracle_function, n, shots
    =3000):
    """ Runs the Deutsch-Jozsa algorithm for an n-qubit system """
    "
```

```

circuit = create_deutsch_jozsa_circuit(oracle_function, n
)
simulator = AerSimulator()
compiled_circuit = transpile(circuit, simulator)
job = simulator.run(compiled_circuit, shots=shots)
result = job.result()
counts = result.get_counts(circuit)
return counts

```

### 3.4 test the algorithm

The following test the algorithm.

Listing 9: Testing

```

def test_deutsch_jozsa(n=3):
    """ Test Deutsch-Jozsa algorithm for n input qubits """

    print("\nTesting Deutsch-Jozsa Algorithm with", n, "input
        qubits...\n")

    # Constant Function f(x) = 0
    result = run_deutsch_jozsa_algorithm(constant_zero_oracle
        , n)
    print("Constant Zero Oracle:", result)

    # Constant Function f(x) = 1
    result = run_deutsch_jozsa_algorithm(constant_one_oracle,
        n)
    print("Constant One Oracle:", result)

    # Balanced Function f(x) = x1 XOR x2 XOR ... XOR xn
    result = run_deutsch_jozsa_algorithm(
        balanced_identity_oracle, n)
    print("Balanced Identity Oracle:", result)

    # Balanced Function (Random Flip of Half Inputs)
    result = run_deutsch_jozsa_algorithm(
        balanced_random_oracle, n)
    print("Balanced Random Oracle:", result)

# Run test for n = 5 qubits
test_deutsch_jozsa(n=5)

```

The complete code for Deutsch-Jozsa Algorithm is as below.

Listing 10: Deutsch-Jozsa Algorithm

```

def constant_zero_oracle(circuit, qr, n):
    """ Oracle for f(x) = 0 (constant function returning always
        0) """
    pass # Do nothing

def constant_one_oracle(circuit, qr, n):

```

```

""" Oracle for  $f(x) = 1$  (constant function returning always
1) """
    circuit.x(qr[n]) # Flip the last qubit (ancilla)

def balanced_identity_oracle(circuit, qr, n):
    """ Balanced oracle where  $f(x) = x_1 \text{ XOR } x_2 \text{ XOR } \dots \text{ XOR } x_n$  """
    for i in range(n):
        circuit.cx(qr[i], qr[n]) # Apply CNOT from input
            qubits to ancilla

def balanced_random_oracle(circuit, qr, n):
    """ Balanced oracle that flips half of the possible inputs """
    "
        for i in range(n // 2): # Flip half of the qubits
            randomly
                circuit.x(qr[i])
        for i in range(n):
            circuit.cx(qr[i], qr[n])
        for i in range(n // 2):
            circuit.x(qr[i])

def create_deutsch_jozsa_circuit(oracle_function, n):
    """ Creates Deutsch-Jozsa circuit with n input qubits """
    qr = list(range(n + 1)) # Create a list of qubit indices
    circuit = QuantumCircuit(n+1, n) # n input qubits + 1
        ancilla

# Initialize the ancilla qubit in  $|\text{ket}\{1\}\rangle$  state
circuit.x(qr[n])

# Apply Hadamard to all qubits
for i in range(n + 1):
    circuit.h(qr[i])

# Apply the oracle
oracle_function(circuit, qr, n)

# Apply Hadamard again to the input qubits
for i in range(n):
    circuit.h(qr[i])

# Measure all input qubits
circuit.measure(qr[:n], range(n))

return circuit

def run_deutsch_jozsa_algorithm(oracle_function, n, shots
=3000):
    """ Runs the Deutsch-Jozsa algorithm for an n-qubit system """
    "
        circuit = create_deutsch_jozsa_circuit(oracle_function, n
            )
        simulator = AerSimulator()
        compiled_circuit = transpile(circuit, simulator)

```

```

    job = simulator.run(compiled_circuit, shots=shots)
    result = job.result()
    counts = result.get_counts(circuit)
    return counts

def test_deutsch_jozsa(n=3):
    """ Test Deutsch-Jozsa algorithm for n input qubits """

    print("\nTesting Deutsch-Jozsa Algorithm with", n, "input
          qubits...\n")

    # Constant Function f(x) = 0
    result = run_deutsch_jozsa_algorithm(constant_zero_oracle, n)
    print("Constant Zero Oracle:", result)

    # Constant Function f(x) = 1
    result = run_deutsch_jozsa_algorithm(constant_one_oracle, n)
    print("Constant One Oracle:", result)

    # Balanced Function f(x) = x1 XOR x2 XOR ... XOR xn
    result = run_deutsch_jozsa_algorithm(balanced_identity_oracle
    , n)
    print("Balanced Identity Oracle:", result)

    # Balanced Function (Random Flip of Half Inputs)
    result = run_deutsch_jozsa_algorithm(balanced_random_oracle,
    n)
    print("Balanced Random Oracle:", result)

    # Run test for n = 5 qubits
    test_deutsch_jozsa(n=5)

```

The output of the above code is

```
Testing Deutsch-Jozsa Algorithm with 5 input qubits...
```

```

Constant Zero Oracle: {'00000': 3000}
Constant One Oracle: {'00000': 3000}
Balanced Identity Oracle: {'11111': 3000}
Balanced Random Oracle: {'11111': 3000}

```

## 4 Next Steps: Bernstein-Vazirani Algorithm

The Deutsch-Jozsa algorithm demonstrated how quantum computation can distinguish between constant and balanced functions in a single query. However, what if instead of determining a function's type, we needed to extract hidden information encoded within the function? This leads us to the Bernstein-Vazirani Algorithm, which extends Deutsch-Jozsa by efficiently solving a hidden binary string problem.

In the next article, we will explore how Bernstein-Vazirani builds on superposition and Hadamard gates to discover a hidden bitstring in a single quantum query, while classical algorithms require multiple evaluations.

## References

- [1] M. A. Nielsen and I. L. Chuang, *Quantum Computation and Quantum Information*, Cambridge University Press, 2010. See Sections 1.4.3, 1.4.4, and 2.2.3 for a detailed discussion of the Deutsch and Deutsch-Jozsa algorithms.