# Introduction to: Quantum Computing For Computer Scientists

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## Information about the talk

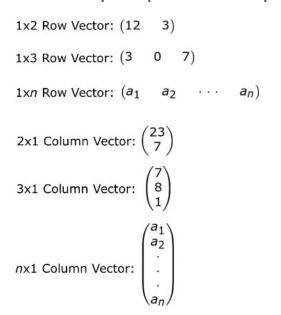
- The talk was originally given at Microsoft in 2018 by Andrew Helwer
- The talk covers a basic introduction to quantum computing and the problems it can solve.
  - Representing computation with basic linear algebra (matrices and vectors)
  - The computational workings of qbits, superposition, and quantum logic gates
  - Solving the Deutsch oracle problem: the simplest problem where a quantum computer outperforms classical methods
  - quantum entanglement and teleportation
  - A live demonstration of quantum entanglement on a real-world quantum computer
  - A demo of the Deutsch oracle problem implemented in Q# with the Microsoft Quantum Development Kit

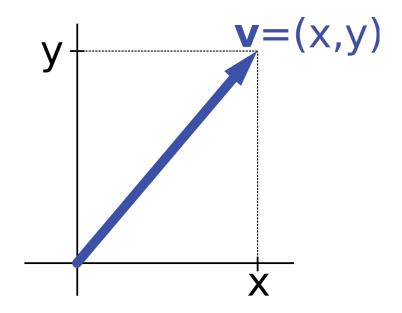
# Linear Algebra Recap

It'll be quick I promise.

## Linear Algebra Review: Vectors

 Vectors are mathematical objects representing lists of numbers, but can also be thought of from a physics perspective as a quantity with direction and magnitude. Both perspectives represent the same abstract idea.





# What we'll be using vectors for today?

Representing bits, qbits, individual state representations of multiple qbits (It'll
make sense later I promise). In a quantum computer we represent our
classical bits as vectors of the form:

One bit with the value 0, also written as |0) (Dirac vector notation) One bit with the value 1, also written as |1)

\*We will build to representing multiple bits as tensor products of single bits that are represented by 2<sup>n</sup> tall column vectors.

$$|4\rangle = |100\rangle = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \otimes \begin{pmatrix} 1 \\ 0 \end{pmatrix} \otimes \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \end{pmatrix}$$

## Linear Algebra Review : Matrices

- Matrices 2d "grids" of numbers that can be used as "transformations" when applied to vectors using matrix multiplication.
- Simple examples of matrices transforming 2d vectors.

$$\begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix} \cdot \begin{bmatrix} 2 \\ 2 \end{bmatrix} \qquad \begin{bmatrix} 2 \\ -2 \end{bmatrix}$$

#### 1. The Identity Matrix

When multiplying by this matrix, the start matrix is unaffected and the new matrix is exactly the same.

	1	0
	0	1
_		20

#### 2. Reflection in the x-axis

When multiplying by this matrix, the x co-ordinate remains unchanged, but the y co-ordinate changes sign.



#### 3. Reflection in the y-axis

When multiplying by this matrix, the y co-ordinate remains unchanged, but the x co-ordinate changes sign.

#### 4. Rotate 180°

When multiplying by this matrix, the point matrix is rotated 180 degrees around (0,0). This changes the sign of both the x and y co-ordinates.

$$\begin{pmatrix} 1 & 1 & 2 \\ 2 & 1 & 3 \\ 1 & 4 & 2 \end{pmatrix} \begin{pmatrix} 3 \\ 1 \\ 2 \end{pmatrix} = \begin{pmatrix} 1 \cdot 3 + 1 \cdot 1 + 1 \cdot 2 \\ 2 \end{pmatrix}$$
 First row,

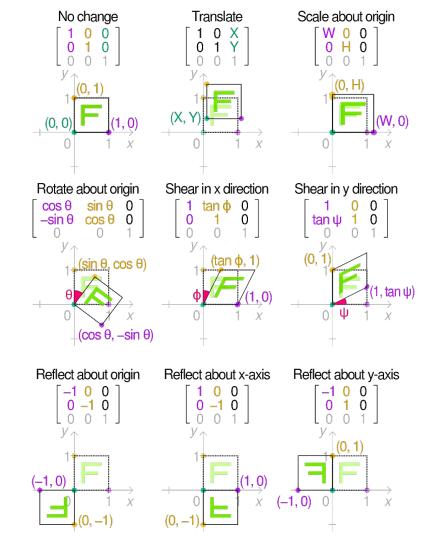
$$\begin{pmatrix} 1 & 1 & 2 \\ 2 & 1 & 3 \\ 1 & 4 & 2 \end{pmatrix} \begin{pmatrix} 3 \\ 1 \\ 2 \end{pmatrix} = \begin{pmatrix} 1 \cdot 3 + 1 \cdot 1 + 1 \cdot 2 \\ 2 \cdot 3 + 1 \cdot 1 + 3 \cdot 2 \end{pmatrix}$$
 next row,

$$\begin{pmatrix} 1 & 1 & 2 \\ 2 & 1 & 3 \\ 1 & 4 & 2 \end{pmatrix} = \begin{pmatrix} 1 \cdot 3 + 1 \cdot 1 + 1 \cdot 2 \\ 2 \cdot 3 + 1 \cdot 1 + 3 \cdot 2 \\ 1 \cdot 3 + 4 \cdot 1 + 2 \cdot 2 \end{pmatrix} = \begin{pmatrix} 6 \\ 13 \\ 11 \end{pmatrix}$$
 last row, then do the additional equation of the equati

# More Complex Matrix Transformations

Why we need 3x3 Matrices for more advanced 2d transformations?

<a href="https://stackoverflow.com/questions/10698962/why-do-2d-transformations-need-3x3-matrices">https://stackoverflow.com/questions/10698962/why-do-2d-transformations-need-3x3-matrices</a>



# What we'll be using matrices for today?

We're going to see that we can use matrices to represent functions on bits, using our previous definitions of bits as 2d vectors, as well as more complex multi-bit transformations.

Identity 
$$f(x) = x$$
  $0 \longrightarrow 0 \ 1 \longrightarrow 1$   $(1 \ 0) \ 0 = (1) \ 0 \longrightarrow 1$   $(1 \ 0) \ 0 = (0) \ 1 \longrightarrow 1$  Negation  $f(x) = \neg x$   $0 \longrightarrow 0 \ 1 \longrightarrow 1$   $(0 \ 1) \ 0 = (0) \ 1 \longrightarrow 1$   $(0 \ 1) \ 0 = (0) \ 1 \longrightarrow 1$   $(0 \ 1) \ 0 = (0) \ 1 \longrightarrow 1$   $(0 \ 1) \ 0 = (0) \ 1 \longrightarrow 1$   $(0 \ 1) \ 0 = (0) \ 1 \longrightarrow 1$   $(0 \ 1) \ 0 = (0) \ 1 \longrightarrow 1$   $(0 \ 1) \ 0 = (0) \ 0 \longrightarrow 1$   $(0 \ 1) \ 0 = (0) \ 1 \longrightarrow 1$   $(0 \ 0) \ 0 = (0) \ 1 \longrightarrow 1$   $(0 \ 0) \ 0 = (0) \ 1 \longrightarrow 1$   $(0 \ 0) \ 0 = (0) \ 1 \longrightarrow 1$   $(0 \ 0) \ 0 = (0) \ 1 \longrightarrow 1$   $(0 \ 0) \ 0 = (0) \ 1 \longrightarrow 1$ 

## **Tensor Product**

- Fancy looking new operator X in circle: <sup>⊗</sup>
- Don't worry about what it is or why it works like this, just realize what it does and how it can be used.
- We use the tensor product for representing multiple bits as a single vector called the <u>individual state representation</u> as shown on side 5

#### Tensor Product Formula

$$\begin{pmatrix} x_0 \\ x_1 \end{pmatrix} \otimes \begin{pmatrix} y_0 \\ y_1 \end{pmatrix} = \begin{pmatrix} x_0 \begin{pmatrix} y_0 \\ y_1 \end{pmatrix} \\ x_1 \begin{pmatrix} y_0 \\ y_1 \end{pmatrix} \end{pmatrix} = \begin{pmatrix} x_0 y_0 \\ x_0 y_1 \\ x_1 y_0 \\ x_1 y_1 \end{pmatrix}$$

$$\begin{pmatrix} x_0 \\ x_1 \end{pmatrix} \otimes \begin{pmatrix} y_0 \\ y_1 \end{pmatrix} \otimes \begin{pmatrix} z_0 \\ z_1 \end{pmatrix} = \begin{pmatrix} x_0 y_0 z_1 \\ x_0 y_1 z_0 \\ x_0 y_1 z_1 \\ x_1 y_0 z_0 \\ x_1 y_0 z_1 \\ x_1 y_1 z_0 \\ x_1 y_1 z_1 \end{pmatrix}$$

#### Individual state representation of 4

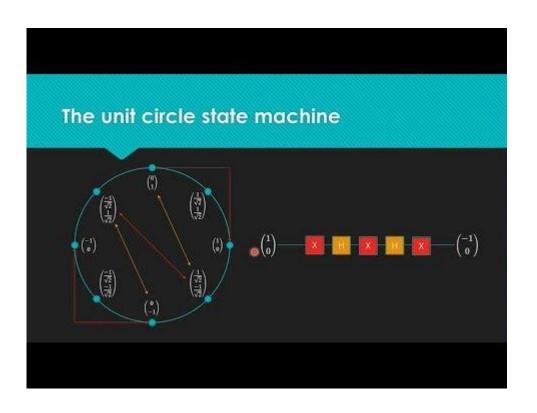
$$|4\rangle = |100\rangle = \begin{pmatrix} 0 \\ 1 \end{pmatrix} \otimes \begin{pmatrix} 1 \\ 0 \end{pmatrix} \otimes \begin{pmatrix} 1 \\ 0 \end{pmatrix} = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 1 \\ 0 \\ 0 \end{pmatrix}$$

# The Talk

## Some things that will be mentioned

- Shor's Algorithm: Famous quantum algorithm developed by Peter Shor to factor really large numbers, breaks current encryption that relies on really large numbers being really hard to factor.
- <u>Deutsch–Jozsa Algorithm:</u> Solves the more complex case of the Deutch oracle problem, in 1 operation on a quantum computer making it an O(1) constant time algorithm, while a classical computer requires 2<sup>(n-1)</sup> + 1 operations making it an O(2<sup>n</sup>) exponential time algorithm.
- Google being expected to achieve <u>Quantum Supremacy</u> by the end of the year? Didn't happen, they achieved it one year later and even this claim has been hotly contested by competitors.

## Let's Watch the talk!



## Let's Discuss!

- What did we learn? What questions do we have?
  - Mathematics behind quantum computing
  - Quantum gates & quantum circuits
  - The Deutch oracle problem
  - Entanglement & Teleportation
  - A brief look at Q#
- Did this talk help get you interested in Quantum Computing or scare you away?
- Can you think of any interesting applications?
- Would you play around with IBMs Free Quantum Computation in your free time?

#### Resources

- Run Quantum Algorithms on a real quantum computer through IBM: <a href="https://quantum-computing.ibm.com/docs/">https://quantum-computing.ibm.com/docs/</a>
- Quantum Computing For Computer Scientists Talk Page:
   <u>https://www.microsoft.com/en-us/research/video/quantum-computing-computer-scientists/</u>

Thanks For Coming!