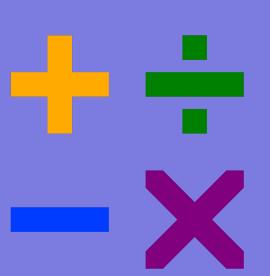
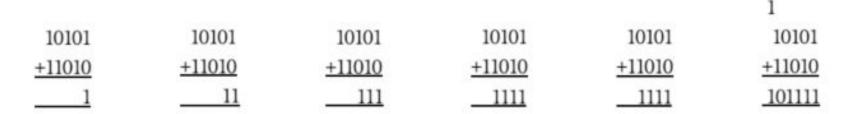
QUANTUM ML RECAP + FUTURE QUANTUM ML RESEARCH

RECAP OF PREVIOUS WEEK -QUANTUM ADDER



ADDING NORMALLY WITH BITS - CONCEPT OF CARRY BIT

- 2 operations during addition
 - Carry operation & sum operation
- Algorithm
 - Start with rightmost bit
 - o Perform sum operation and pass carry bits to column to left
 - Repeat process till you reach left most column and no more carry bits



Adding 21 and 26 in binary - demonstrating the sum and carry operation

ADDING ON QUANTUM COMP

- ➤ Objective Convert diff rules of adding in quantum ⇒ quantum gates
 - NOT, CX, CCX gates
- ➤ Algorithm
 - \circ Flip original states of $|0\rangle$ to $|1\rangle$ of qubits in beginning to start process of adding numbers (Use X gate)
 - Need to compute "carry bit" in adding
 - Rule = if at least 2 input qubits are in 1 state ⇒ output = 1
 - 3 input qubits = input carry from previous iteration and 2 addends
 - Use CX and CCX gates for carry bit

ADDING ON QUANTUM COMP

Input Carry	Bit from A	Bit from B	Output
0	0	0	0
0	0	1	0
0	1	0	0
0	1	1	1
1	0	0	0
1	0	1	1
1	1	0	1
1	1	1	1

Carry Gate Operation Truth Table

ADDITION ON QUANT COMP, FASTER METHOD

- Use Quantum Fourier Transformation algo on 2 addends
 - Quantum fourier transformation function which finds the "ingredients" of a particular "dish"
 - Obtaining the simpler waves from the complex wave
- > QFT Mathematically = rotating the qubit in complex plane
 - \circ Quantum gate equivalent = Keep on repeating controlled-U gates \Rightarrow applies the rotations around the circle \Rightarrow rotations rep as wave
 - Obtain simpler waves (rotations about plane)

$$y_k = \frac{1}{\sqrt{N}} \sum_{j=0}^{N-1} x_j e^{\frac{2\pi i k j}{N}}$$
 $F(\Psi) = \frac{1}{\sqrt{N}} \sum_{j=0}^{N-1} \Psi_j e^{\frac{2\pi i k j}{N}}$

BUT HOW IS QFT HELPFUL FOR ADDITION?

Represent 2 original equations y = 7 and y = 3 as diff circles (precalc knowledge)

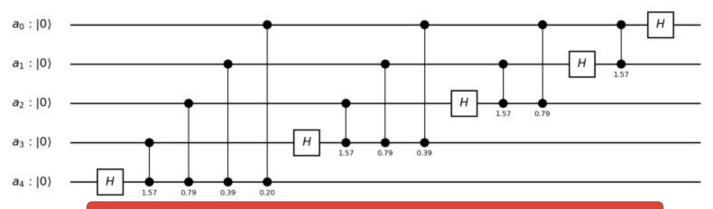
$$f(t) = 3e^{2\pi i0t}$$
$$g(t) = 7e^{2\pi i0t}$$

- ➤ Once obtaining circles → can compute waves from QFT algo
- ➤ Add the 2 Fourier transformations together (the 2 waves)

$$\hat{g}(f+g) = \hat{g}(f) + \hat{g}(g)$$

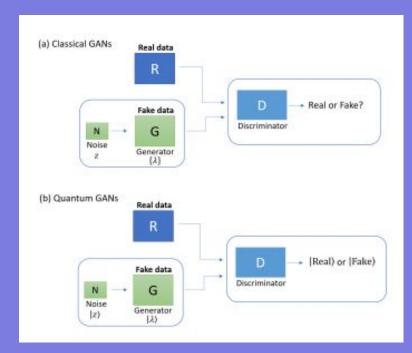
ADDITION ON QUANT COMP, FASTER METHOD (CONT.)

- > Algorithm
 - Take input
 - Apply Hadamard gate on qubit
 - Apply controlled-U1 gate
 - Rotates the qubit



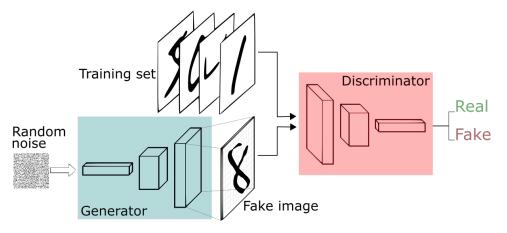
Numbers = angle by which qubit is rotated in complex plane

RECAP OF QUANTUM ML ALGOS - QGAN



RECAP OF ML - GAN

- Unsupervised model using supervised loss function
- 2 competing neural networks
 - Generative and Discriminator
- ➤ Great model which solves issue of not having enough data in real world cases → generates own data and gets trained from that
 - Still several challenges of GAN



RECAP OF ML-GAN (CONT.)

- Generator neural network creates fake images from input vector
- Discriminator network learns to distinguish between the real and false images (real images taken from dataset)
 - Discriminator returns diff prob of each image being real (1 = real and 0 = fake)
- Cop-counterfeiter analogy
 - Counterfeiter learning to pass fake notes and cop is also in training as is able to start distinguishing

ARCHITECTURE QGAN (HOW IT WORKS HIGH LEVEL)

- ➤ 2 quantum circuits
 - 1 models the generator & the other circuit models the discriminator
 - o G is generative quantum circuit & D is discriminator quantum circuit
 - Gates of G parameterized by theta_G, similar with circuit D
- Input state + noise inputted into circuit G and G feeds D various types of fake data
 - o D's job to get trained and distinguish real vs. fake
- Compute gradient of quantum optimization with quantum circuit (Quant circuit computes gradient)
 - Classical gradient descent is used to optimize (minimize) GAN cost function
 - Cost function ⇒ measures how well algo matches estimate

QGAN

https://arxiv.org/p df/1804.08641.pdf

QUANTUM COMPUTING/ML RESEARCH + FUTURE DIRECTIONS

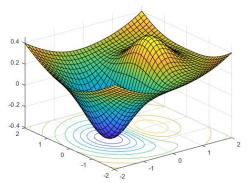


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QUANT COMPUTING/QUANT ML APPLICATION AREAS

- Cryptography
- Molecular Modeling
- > Communication
- Forecasting Weather
- > Experiments Undertaken in Particle Physics
- > Genomics
- > Optimization



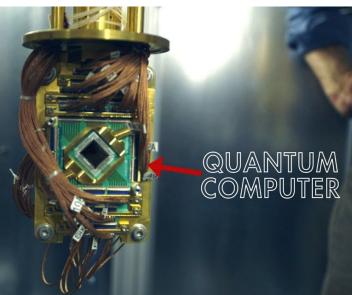


SOME COMPANIES+AGENCIES IN QUANTUM COMPUTING

- Some companies+agencies involved in quantum computing
- ➤ D-wave
- ➤ Google
- > IBM
- > Xanadu
- > Microsoft
- > NASA







QUANTUM COMPUTINGCRYPTOGRAPHY

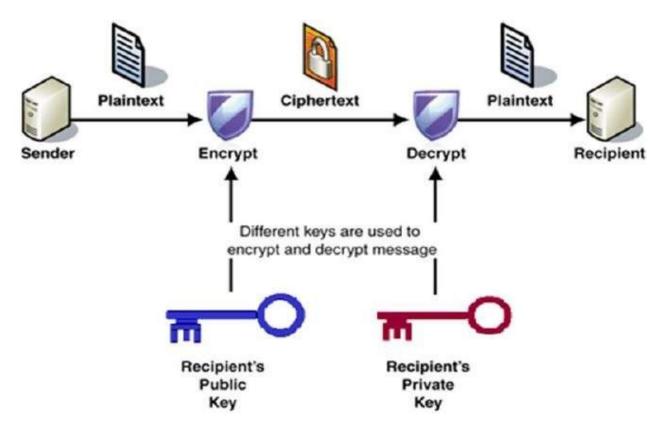
CLASSICAL CRYPTOGRAPHY CONCEPTS + VOCAB

- Algorithms developed to
 - Maintain secrecy and privacy of data being transmitted
 - Authenticate data being received
- > Plaintext vs. ciphertext
 - Plaintext = original message
 - Ciphertext = message altered through cipher
- ➤ Cipher
 - Algo itself which transforms plaintext so that it can't be easily read
 - Transformation, substitution, etc.
- ➤ Encipher/Encode
 - Plaintext ⇒ ciphertext using cipher + key
- Decipher/Decode
 - Ciphertext ⇒ plaintext using cipher + key

CLASSICAL CRYPT - PUBLIC KEY

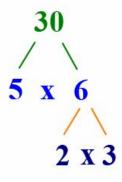
- Public key (Asymmetric cryptography)
 - Uses both public and private keys
 - Public key → can be shared with anyone
 - Private key → kept secret
 - Either key → used for encryption & opposite key → used for decryption
- Example scenario
 - Sender (Alice) & Receiver (Bob)
 - Alice get's Bob's public key
 - Plaintext encrypted with public key algo → converted to ciphertext
 - Ciphertext sent to Bob → Bob decrypts ciphertext with his private key to read message

CLASSICAL CRYPT - PUBLIC KEY



CLASSICAL CRYPT - PRIME FACTORIZATION

- What are the algorithms used for classical crypt?
- ➤ Prime factorization → public key encryption
 - Multiply 2 large primes together → encrypts message
 - To decrypt → need to break the big number into its prime factors
- No shortcut → trial and error
 - Large numbers initially → hard to break the message because hard to decrypt back into its original prime factors



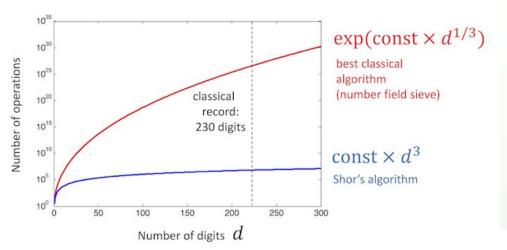
RSA - A QUANTUM ALGORITHM

- Standard cryptographic algo
- > Used for Internet
- Public key encryption, uses prime factorization technique



SHOR'S QUANTUM ALGORITHM

- Solves problem of trying to find prime factors for very large number efficiently
- Can be run on both classical and quantum computers
 - Runs very slow on classical comp
 - Very fast on quantum comp



Shor's algorithm $|0\rangle - H$ $|0\rangle - H$

THE POWER OF SHOR'S ALGO IN QUANTUM WORLD

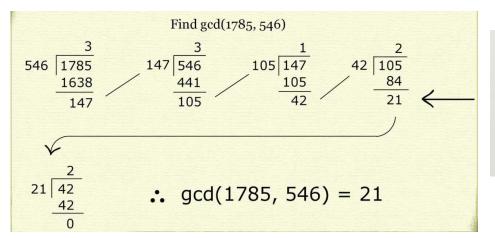
- Can decrypt message for very large prime numbers using Shor's algo
 - Thus, can afford to use big numbers for quantum cryptography in future
 - Hackers will need to break laws of quantum physics to compute the prime numbers

SHOR'S QUANTUM ALGORITHM (CONT.)

- ➤ The Algorithm -
 - Original number (N)
 - Random guess (g) which could be a factor
 - Transform (g) into a better guess which will likely be a factor of (N)
- Doesn't actually use quantum mechanics inside algo BUT works more efficient on quantum comp

NEED TO CHECK IF "G" SHARES COMMON FACTOR WITH "N"

- ➤ When picking g
 - Don't need g to be an actual factor of N
 - INSTEAD → g just needs to share a factor with N (>=1 factors shared)
- ➤ Euclid's algorithm → Ex divide N by g to get a → build
 factors of N → part of Number Theory



NEED TO CHECK IF "X" SHARES COMMON FACTOR WITH "N"

- Euclid's algo is still inefficient for large numbers
- Transform g to a number which will share a factor with N

Shares a factor with N?

SHOR'S ALGORITHM (CONT.)

$$\frac{N, g}{g^{p} = m \cdot N + 1}$$

$$g^{p} = m \cdot N$$

$$g^{p} - 1 = m \cdot N$$

$$(g^{p/2} + 1) \cdot (g^{p/2} - 1) = m \cdot N$$
Something something

$$(9^{8/2}+1)(9^{8/2}-1) = m \cdot N$$

 $a \cdot factor b \cdot factor$
 $7^{4/2}+1 = 50 \times 15$
 $7^{4/2}-1 = 48 \times 3$

Can find factors of $g^(p/2)+1 \& g^(p/2)-1$ with EUCLID'S ALGORITHM!!!

Even though 50 and 48 aren't factors of $15 \rightarrow$ share common factors with 15 which can be determined by applying Euclid's with gcd(50, 15) AND gcd(48, 15)

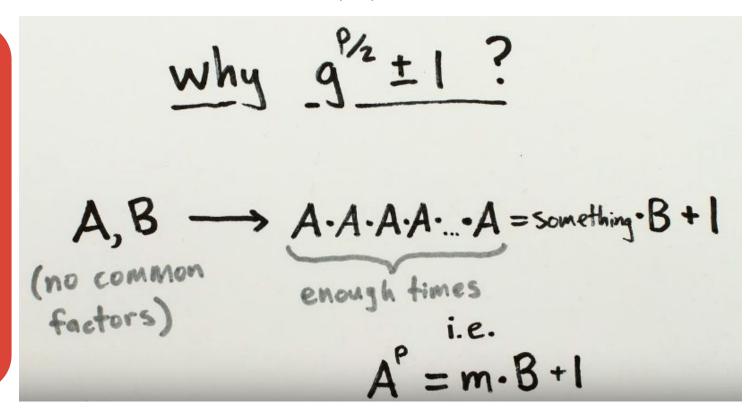
EXAMPLES

7, 15	42,13
$7^{2} = 3.15 + 4$ $7^{3} = 22.15 + 13$ $7^{4} = 160.15 + 1$	$42^2 = 135 \cdot 13 + 9$ $42^3 = 5699 \cdot 13 + 1$

NEED TO CHECK IF "G" SHARES COMMON FACTOR WITH "N"

Reasoning for $g^(p/2) + 1$

- 2 #s A, B with no common factors
- If A is multiplied with itself enough times = m * B + 1



SOME PROBLEMS OF SHOR'S ALGORITHM

➤ 3 central problems ⇒ hard to implement on quantum computer

$$\left(\frac{g^{p_2}}{g^{p_2}+1}\right)\left(\frac{g^{p_2}-1}{g^{-1}}\right)=m\cdot N$$
a. N factor of m

Problem #1 - neither #s produced by algo will be useful for finding factors through Euclidean algo

SOME PROBLEMS OF SHOR'S ALGORITHM

$$9^{3/2}$$
 = 272.191...
 $19^{3/2}$ = 15.588...
 $19^{3/2}$ = 29897.688...

Problem #2 - number p is odd ⇒ fraction will not be an integer ⇒ number will not be an integer, only want int

SOME PROBLEMS OF SHOR'S ALGORITHM

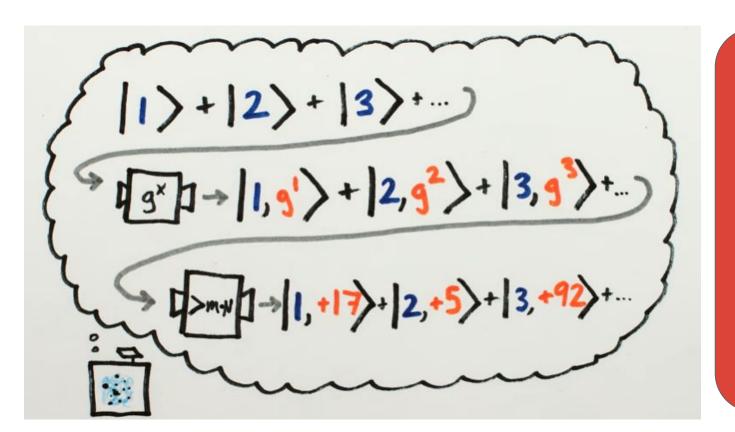
Problem #3 - Finding "p" \Rightarrow classical computer \rightarrow take a lot of time vs. quantum computer will harness quantum mechanics

Solution - quantum superposition → calculate all poss answers simultaneously

Output \rightarrow all wrong answers are destructively interfere & are removed singling out 1 correct answer

Essence of Shor's Algo = translates problem into quantum form so output has wrong answers destructively interfering and 1 right answer through superposition ⇒ find "p"

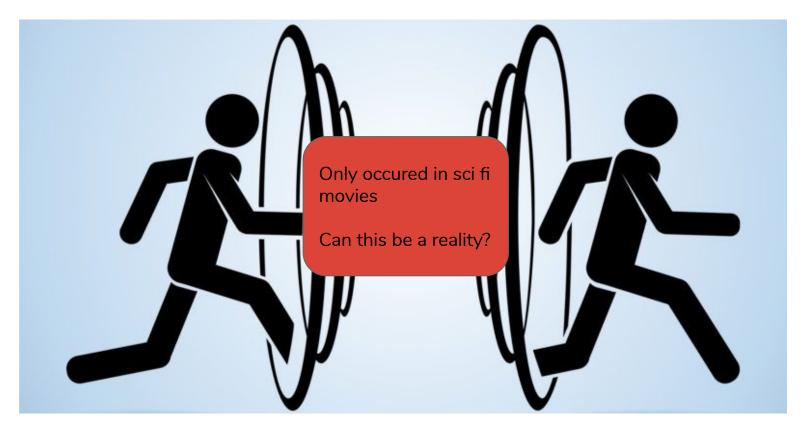
SOLUTION OF PROBLEM #3 DIAGRAMMATICALLY



Send in a superposition of #s → superposition of all possible powers guess can be raised to → superposition of how much bigger each of the guesses raised to a power are

QUANTUM COMPUTINGCOMMUNICATION

TELEPORTATION



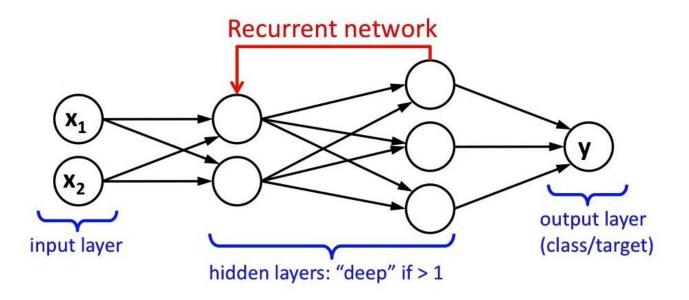
ENTANGLEMENT - TELEPORTATION

- "Spooky action a distance"
- Can be harnessed for teleportation
- Chinese scientists teleported packet of info to space
 - FARTHEST distance achieved long distance comm
 - Eavesdropper can't use this type of long distance system without alerting the other end
 - https://www.space.com/37506-quantum-teleportation-record-shattered.ht
 ml
- > Far from teleporting humans
 - Going from one place to the other
 - Same constituents transferred → exactly same person

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QUANTUM ML-QUANTUM BOLTZMANN MACHINE

RECURRENT NEURAL NETWORK

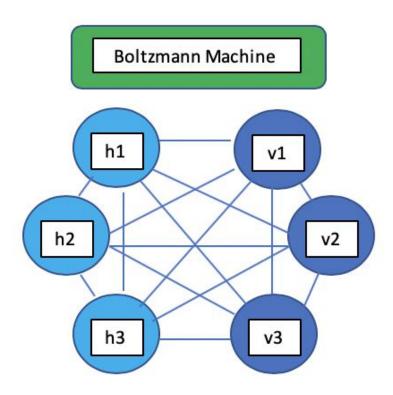


Looks at previous events in network to make conclusions for future layers

BOLTZMANN MACHINE

- Type of ML Recurrent neural network
 - Type of neural network which looks at previous outputs to determine current output
 - Neural network use inputs and adjust weights to get output, where error approaches 0 after several iterations
- > Stochastic vs Deterministic neural net
 - Deterministic/normal neural network
 - Output is unique/deterministic for fixed input
 - Introduce random variations into neural net
 - Can give stochastic weights or stochastic transfer function
 - Ex of stochastic transfer function ⇒ In Boltzmann machine, each neuron has "binary value" (fired or not fired) ⇒ chance of neuron firing depends on state of other neurons
 - Useful for optimization probs ⇒
 - Stochastic neural network
 - Output will be different/stochastic/random for fixed input

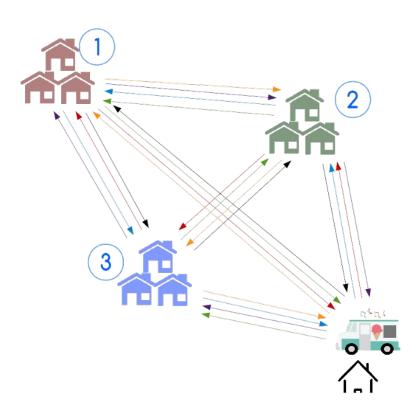
BOLTZMANN MACHINE



ANALOGY FOR QUANTUM ANNEALING

Problem Statement -

lce cream business & need to sell ice cream by going to different neighborhoods → optimal route going through all neighborhoods (Traveling Salesman Problem)



Solution-

Find shortest distance by comparing all poss paths → BUT inefficient if large # neighborhoods

ANALOGY FOR QUANTUM ANNEALING (CONT.)

Actual Solution -

- 1. Encode solution into physical system
 - a. Convert info of ice cream problem into physical system so **ground state = solution**
 - i. Why ground state?
 - Easier to go down the mountain vs. climbing

 → systems tend to ground state
 - b. Physical systems → described by
 Hamiltonian function
 - i. Can get energy of system, other quantum phys prop of system, etc.

Going back to Ice Cream Analogy -

- 1. Include constraints to problem (Ex amt of max distance ice cream truck can travel before gas runs out, which roads truck is not allowed to use, etc.)
 - a. If constraint is NOT satisfied \rightarrow add terms to Hamiltonian (inc energy of system \rightarrow farther from ground state = solution)
 - 2. Ex spin interaction changes based on distance traveled
 - a. More distance traveled =
 more spin (more energy),
 less distance = less spin
 (less energy) = closer to
 qround state
 - 3. Lowest energy state, ground state = solution through SIMULATION

QUANTUM ANNEALING

- > Find minimum of a function
- ➤ Type of optimization problem → go through all possible solutions and find min = solution
- D-wave machine = quantum annealer which runs quantum comp algos
- > Application = Predicting financial crashes
 - Great for predicting financial issues for Wall Street & equilibrium needed to avoid such crashes
 - Ex minimize cost of trading and minimize market risk ⇒ optimizing the global min value through quantum annealing
 - Able to identify right equilibrium and in turn predict financial crashes

QUANTUM BOLTZMANN MACHINE

- Quantum annealing with boltzmann machine
- > Boltzmann machine
 - Use binary variables
- Run boltzmann machine on quantum annealing processor to get solution to optimization problems
 - Training the Boltzmann machine on quantum annealing processor will be more efficient
 - Applications → make better predictions of financial problems (financial forecasting), etc.

RECAP



ANALOGY

- ➤ Objective
 - Search for a briefcase lost in building (100 levels)
- Classical Computer
 - One person checks each floor of the building repetitively
- > Supercomputer
 - A team of people search for briefcase at same time
- ➤ Quantum computer
 - 1 person is placed in every room of each level
 - Find briefcase INSTANTLY!



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HOPE YOU HAD FUN LEARNING QUANTUM

COMPUTING + QUANTUM ML!