

# **Quantum Finance:**

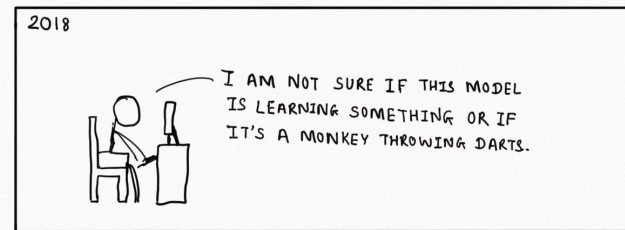
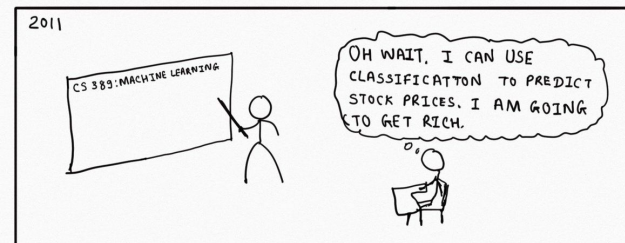
## **Using Quantum Methods to Speed-up Monte Carlo Simulations**

**Zach Battenwieser**

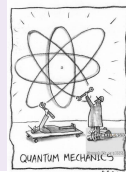
**Dillon Dunteman**

# Background

## Why did we choose this project?



2019



We can potentially predict prices faster and more accurately using quantum methods!

# Finance Introduction

What is a European call option?

$$f(S_T) = \max\{0, S_T - K\}$$



IBM stock has been trading  
at ~\$140 on the NYSE

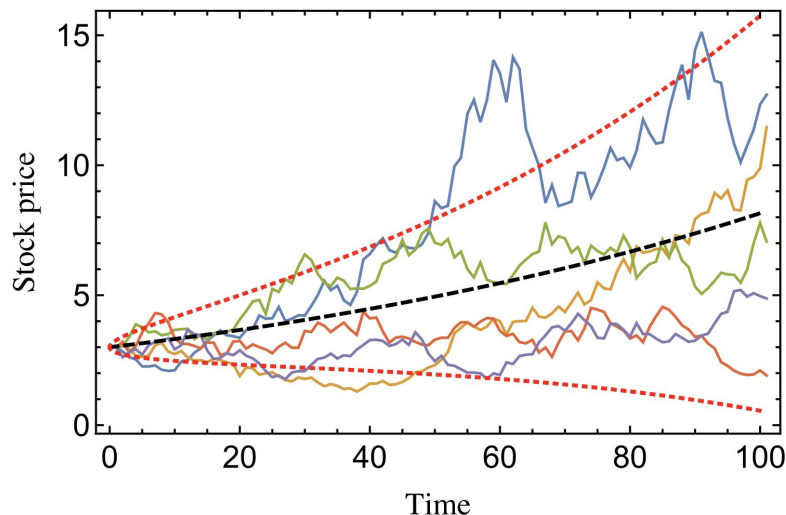
Assume that one month ago, Zach  
bought a European call option at a strike  
price of **\$130** with an expiration date of  
today.

He can then sell the stock at its market  
value of **\$140** for a profit of **\$10**.

# Finance Introduction

## Analytical Pricing: The Black Scholes Model

By taking the limit of a binomial tree, underlying assets are modeled by a stochastic process known as Brownian Motion



$$S_t = S_0 e^{\sigma W_t + (\alpha - \sigma^2/2)t}$$

$$f(S_T) = \max\{0, S_T - K\}$$

$$\Pi = e^{-rT} \mathbb{E}_{\mathbb{Q}}[f(S_T)]$$

# **Improving this Pricing Method**

**Can a quantum algorithm be utilized in order to improve efficiency of Monte Carlo methods and decrease estimation error?**

# Quantum Monte Carlo Methods

## Loading Uncertainty Models

Given:

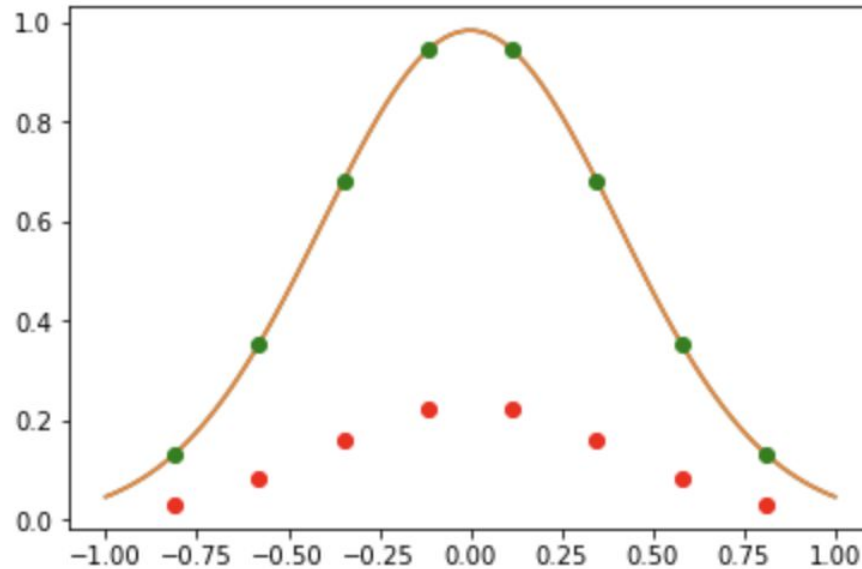
$$W_t - W_s \sim N(0, t - s)$$

We know:

$$p_T(x) = \frac{1}{\sqrt{2\pi T}} e^{-\frac{x^2}{2T}}.$$

# Quantum Monte Carlo Methods

**Qubits only hold discrete distributions**



# Quantum Monte Carlo Methods

## Create the G gate

$$\mathcal{G}|0^n\rangle = \sum_{j=0}^{2^n-1} \sqrt{p_j}|j\rangle.$$

Find G\_0:

```
[ [ 0.17747501+0.j]
  [ 0.28963622+0.j]
  [ 0.40147945+0.j]
  [ 0.47268143+0.j]
  [ 0.47268143+0.j]
  [ 0.40147945+0.j]
  [ 0.28963622+0.j]
  [ 0.17747501+0.j]]]
```



# Quantum Monte Carlo Methods

## Create the Chi State

The R gate encodes the payoff function  $v(x)$  in an ancillary qubit

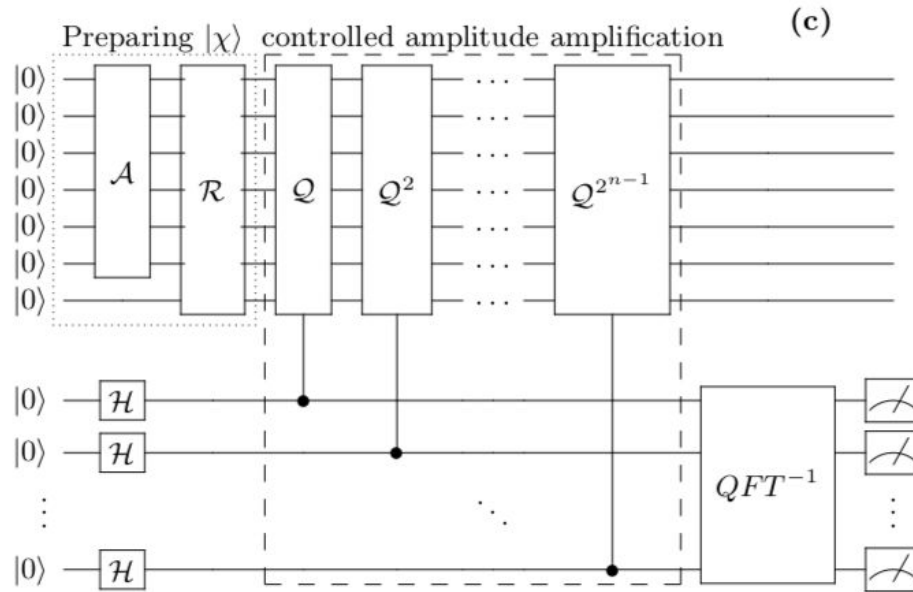
$$\mathcal{R}|x\rangle|0\rangle = |x\rangle(\sqrt{1-v(x)}|0\rangle + \sqrt{v(x)}|1\rangle).$$

$$\rightarrow \sum_{j=0}^{2^n-1} \sqrt{p(x_j)}|j\rangle \left( \sqrt{1-\tilde{v}(x_j)}|0\rangle + \sqrt{\tilde{v}(x_j)}|1\rangle \right) =: |\chi\rangle.$$

Measuring the final qubit gives the expectation

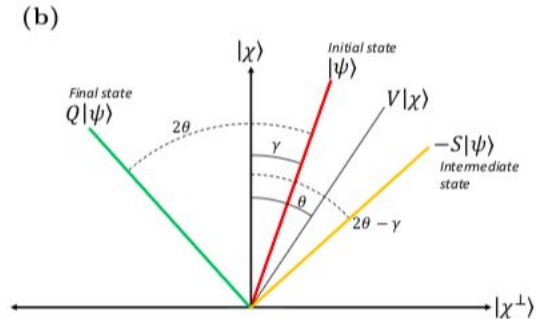
$$\mu = \langle \chi | (\mathcal{I}_{2^n} \otimes |1\rangle\langle 1|) | \chi \rangle = \sum_{j=0}^{2^n-1} p_T(x_j) \tilde{v}(x_j).$$

# Quantum Circuit Overview



## Q Gate and amplitude amplification

$$\mathcal{V}|\chi\rangle = \cos(\theta/2)|\chi\rangle + e^{i\phi} \sin(\theta/2)|\chi^\perp\rangle$$



Expectation value from multi-qubit state  $|\chi\rangle$  can be calculated if we find  $\theta$

$$1 - 2\mu = \cos(\theta/2).$$

# Our Simulation

In [834]: *# One Function to Price Option*

```
def Quantum_Price(S, K, T, r, sigma, qubits):
    def v_euro(x):
        return max(0, S * np.exp((sigma * x) + (r - 0.5 * sigma * sigma)*T)
    dp = disc_points(T, sigma, 2, qubits)
    norm_probs = norm_disc_probs(dp, T)
    G_0 = G_on_zero(norm_probs, qubits)
    chi = get_Chi(G_0, dp, norm_probs, qubits, v_euro)
    eq = mu(chi, qubits)
    return eq * np.e ** (-r * T)

print(Quantum_Price(100, 95, 60/365, 0.05, 0.3, 3))
print(Quantum_Price(100, 95, 60/365, 0.05, 0.3, 6))
print(bs_price(100, 95, 60/365, 0.05, 0.3))
```

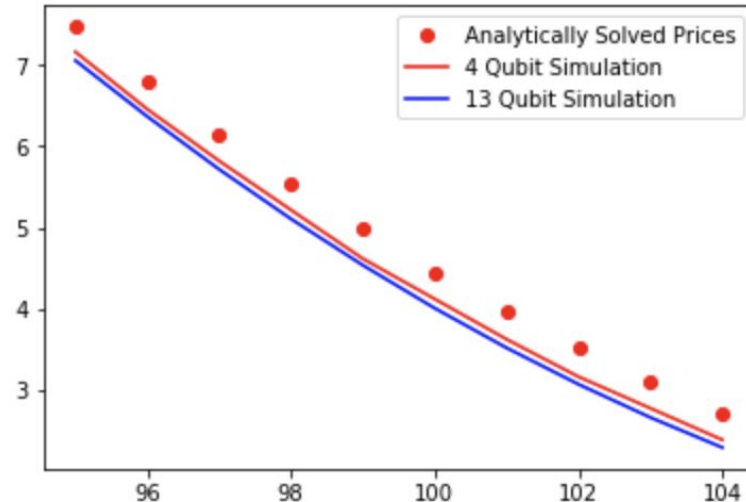
7.87276420958

7.65438447923

8.15000633124

# Our Simulation

Quantum Generated Prices (brown) vs.  
Analytically Calculated Prices (blue)



# Running on a Real Device

- Qiskit Aqua Finance Package

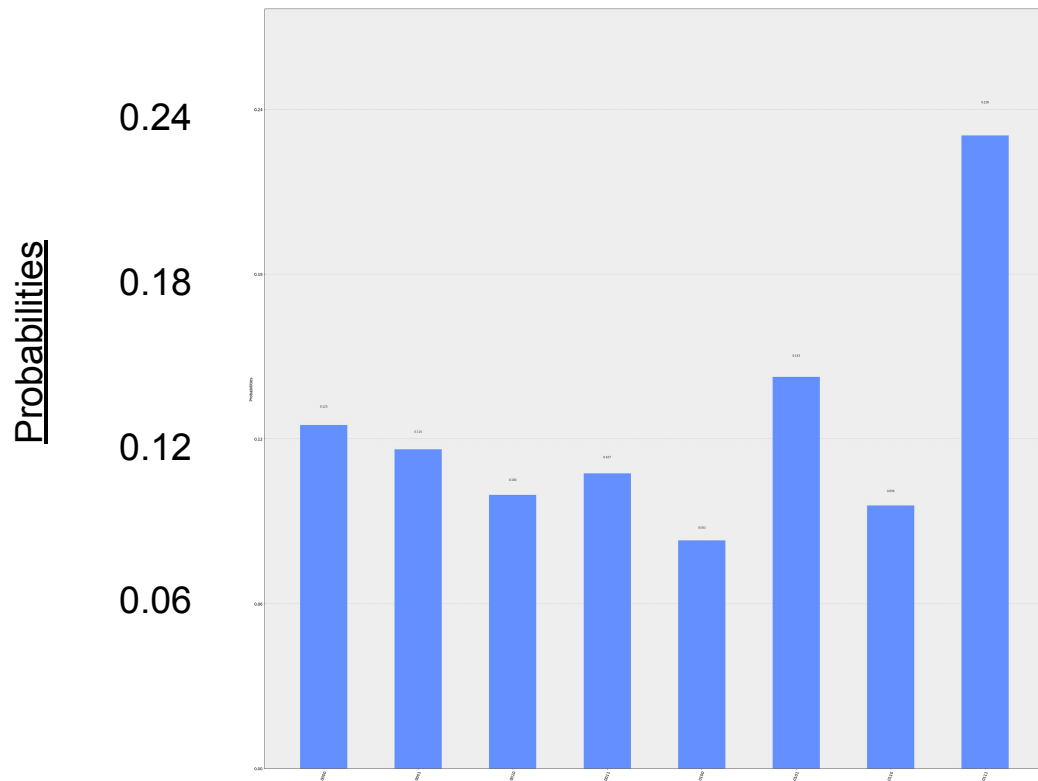
```
# construct circuit factory for payoff function
european_call = EuropeanCallExpectedValue(
    uncertainty_model,
    strike_price=strike_price,
    c_approx=c_approx
)
```

```
# number of ancillary qubits = number in uncertainty model+2
anc_qubits = num_uncertainty_qubits+2
N = num_uncertainty_qubits + anc_qubits

q = QuantumRegister(N)
c = ClassicalRegister(N)
qc = QuantumCircuit(q,c)
european_call.build(qc,q, q_ancillas=[q[i] for i in range(anc_qubits, N)])
qc.measure(q,c)
for i in range(0,7):
    qc.measure(q[i],c[i])
```

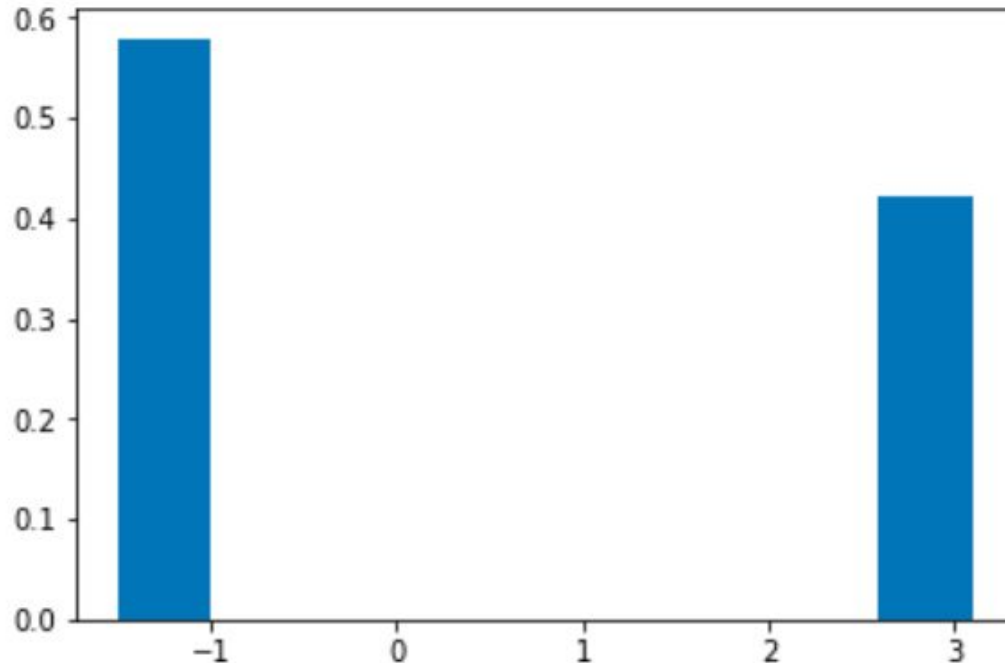
# Real Device Results

5 Qubit Device (3 ancillary qubits measured)



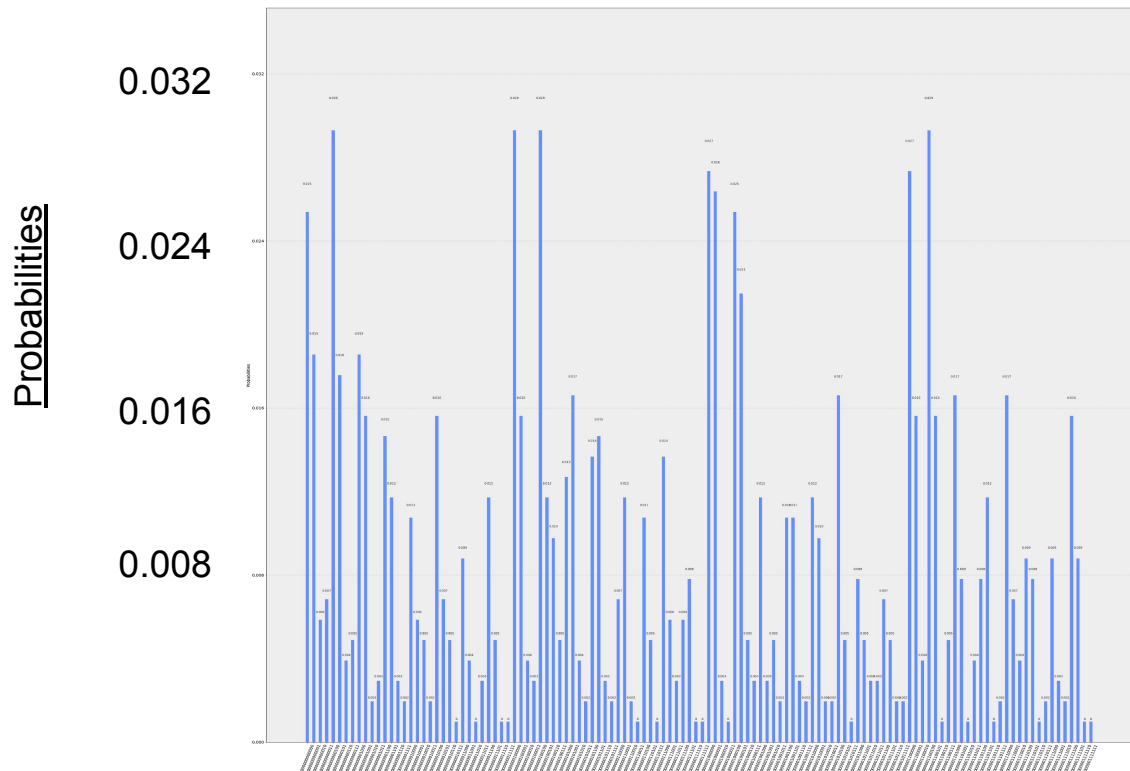
# Real Device Results - Amplitude Amplification

With a strike price of \$2 and payoff price estimated at ~\$3, we can see a calculated payoff of \$1



## Real Device Results

## 14 Qubit Device (7 ancillary qubits measured)





# Future Work

- **Future experimentation and analysis of results on the real device, along with successful amplitude estimation, could lead to meaningful conclusions about the speed-up**
- **Current theoretical speed-up is nearly quadratic, but robustness-to-noise is unknown**
- **Modifying “G” gate and changing payoff functions can allow for speed-up of pricing of more complex derivatives**