

European Options Pricing

Put - Call

CS 405: Quantum Computation

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Assumptions

- A payoff function $v(x)$ is available which provides the financial derivative to be priced.
- $v(x) : \{0, 1\}^n \rightarrow [0, 1]$
- A: produces a state vector with each component state $|x\rangle$ with a probability $|a_x|^2$, captures the probability distribution of the underlying asset i.e. stocks $A |0^{\otimes n}\rangle = \sum a_x |x\rangle$
- Goal: Estimate $E[v(x)] = \sum v(x) |a_x|^2$

Algorithm

- Assume we can implement a rotation R onto a ancilla qubit to get

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

- Perform R on $A|0^{\otimes n}\rangle|0\rangle$. Call this operation F

$$|\chi\rangle := \sum_{x=0}^{2^n-1} a_x |x\rangle (\sqrt{1-v(x)}|0\rangle + \sqrt{v(x)}|1\rangle)$$

$$F|0^{\otimes n+1}\rangle = R(A \otimes I_2)|0^{\otimes n+1}\rangle \equiv |\chi\rangle$$

Algorithm

- Consider observable $I_{2^n} \otimes |1\rangle\langle 1|$ on χ

$$\langle \chi | (I_{2^n} \otimes |1\rangle\langle 1|) | \chi \rangle = \sum |a_x|^2 v(x) =: \mu$$

- Now we will use phase estimation method to estimate the stock price which would lead us to the QMC estimation. Consider another observable

$$V = I_{2^{n+1}} - 2I_{2^n} \otimes |1\rangle\langle 1|$$

Algorithm

- Note that $\langle \chi | V | \chi \rangle = \langle \chi | I_{2^{n+1}} - 2I_{2^n} \otimes |1\rangle \langle 1| \chi \rangle = 1 - 2\mu$
- Any quantum state in the $(n + 1)$ -qubit Hilbert space can be expressed as a linear combination of $|\chi\rangle$ and an orthogonal complement $|\chi^\perp\rangle$.

$$V |x\rangle = \cos \frac{\theta}{2} |\chi\rangle + e^{i\phi} \sin \frac{\theta}{2} |\chi^\perp\rangle$$

$$1 - 2\mu = \cos \frac{\theta}{2}$$

Now our task is to find θ

Phase Estimation

- We will now define the transformation **Q** that encodes **Θ** in its eigenvalues. Before that, we will define two unitary operators.
- Unitary Reflection **U which acts as** $U|\chi\rangle = -|\chi\rangle$ **and** $U|\chi^\perp\rangle = |\chi^\perp\rangle$.
- **-U** reflects across $|\chi\rangle$ and leaves $|\chi\rangle$ itself unchanged.

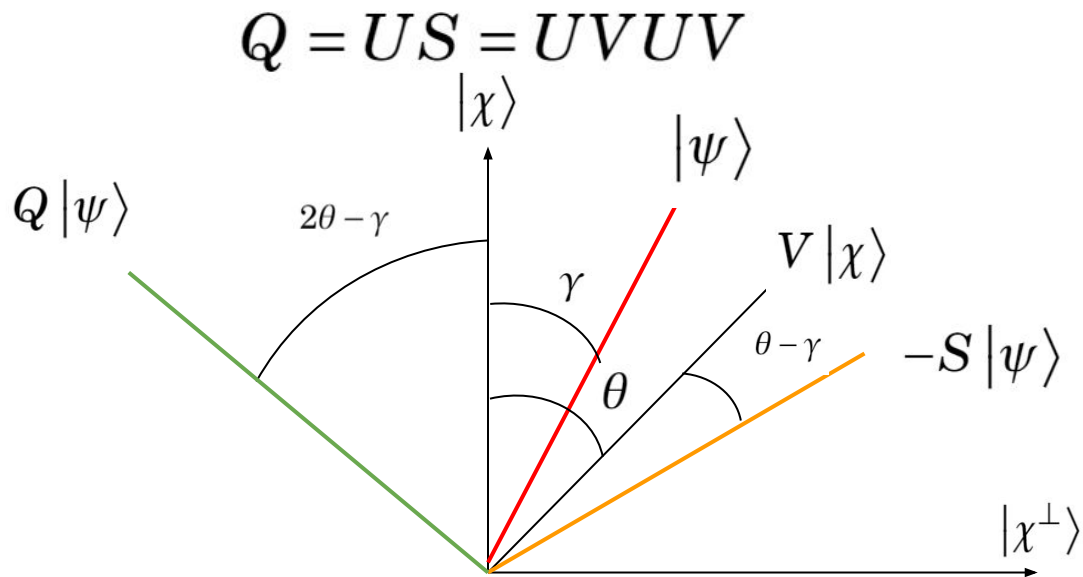
$$U = I_{2^{n+1}} - 2|\chi\rangle\langle\chi|$$

- **Similarly, we can define -S which reflects across** $V|x\rangle$

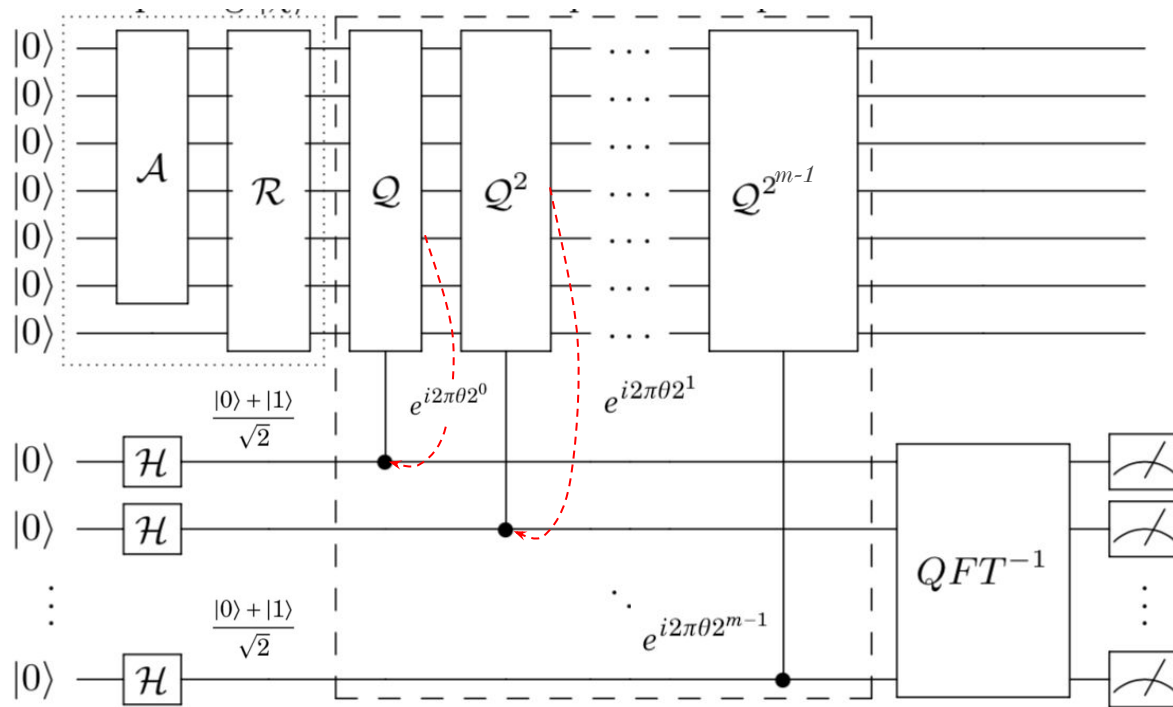
$$S = I_{2^{n+1}} - 2V|\chi\rangle\langle\chi|V \equiv VUV$$

Phase Estimation

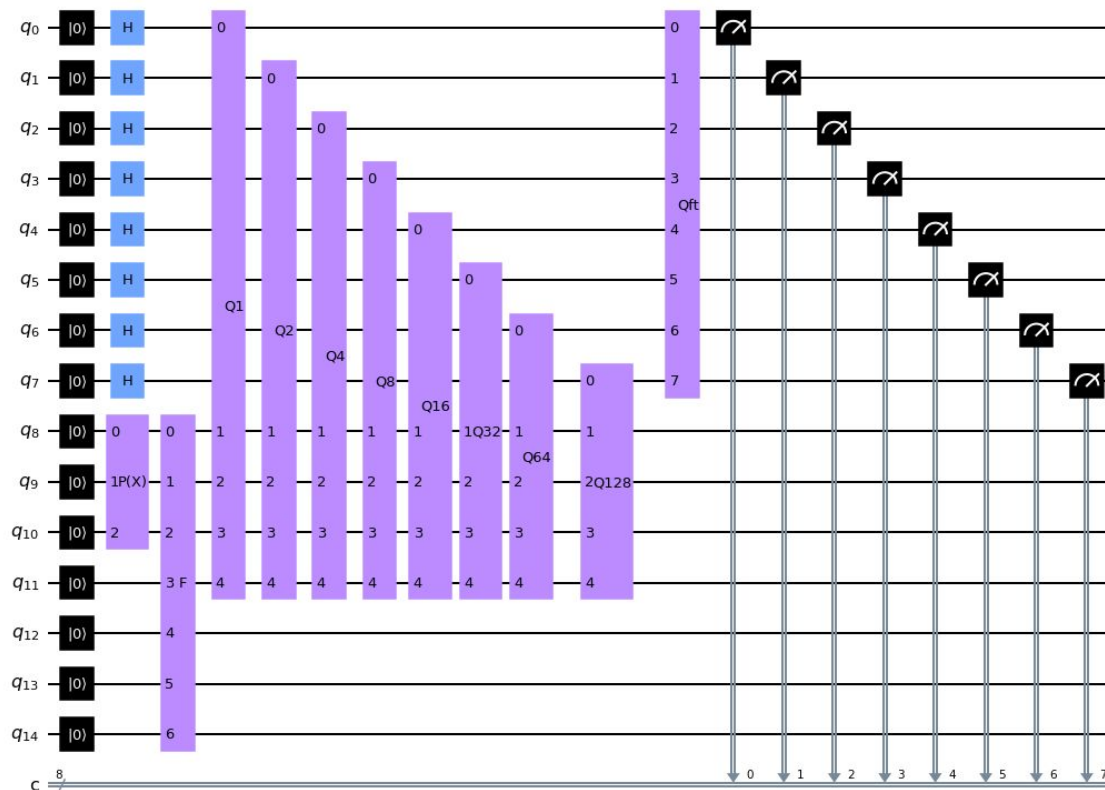
- Hence transformation \mathbf{Q} can be defined as



Phase Estimation Circuit



Custom Phase Estimation Circuit



Phase Estimation

- Phase estimation requires the conditional application of the operator Q .

$$Q^c : |j\rangle |\psi\rangle \rightarrow |j\rangle Q^j |\psi\rangle$$

- Applying hadamard gate on first m qubits, we got,

$$H^{\otimes m} |0^m\rangle |\chi\rangle = \frac{1}{\sqrt{2^m}} \sum_0^{2^m-1} |j\rangle |\chi\rangle$$

Notice that $0.\theta_1 \dots \theta_{m-1} \theta_m = \theta$

Phase Estimation

- Finally applying Q^c operator and inverse QFT,

$$\begin{aligned}
 \frac{1}{\sqrt{2^m}} \sum_{j=0}^{2^m-1} |j\rangle Q^j |\chi\rangle &= \left((|0\rangle + e^{i2\pi 2^{m-1}\theta} |1\rangle) \right) \left((|0\rangle + e^{i2\pi 2^{m-2}\theta} |1\rangle) \right) \times \\
 &\quad \times \dots \times \left((|0\rangle + e^{i2\pi\theta} |1\rangle) \right) \\
 &= \sum_{y=0}^{2^m-1} e^{i2\pi\theta y} |y\rangle.
 \end{aligned}$$

Diagrammatic annotations:

- An arrow points from $0.\theta_m$ to the first term $(|0\rangle + e^{i2\pi 2^{m-1}\theta} |1\rangle)$.
- An arrow points from $0.\theta_{m-1}\theta_m$ to the second term $(|0\rangle + e^{i2\pi 2^{m-2}\theta} |1\rangle)$.
- An arrow points from the last term $(|0\rangle + e^{i2\pi\theta} |1\rangle)$ to $0.\theta_1 \dots \theta_{m-1} \theta_m = \theta$.

Option Pricing

- $v(x)$ is the pay off function when considering some distribution
- $E[v(x)]$ is the price of the option at time T
- Then the price of the option would be the discounted value

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

- The probability density for this random variable is given by $P_T(x)$
- To prepare an approximate superposition of these probabilities, we take the support of this density form and discretize the interval $[-x_{\max}, x_{\max}]$.

- $$p_j = \frac{p_T(x_j)}{C} \qquad C = \sum_{j=0}^{2^n-1} p_T(x_j)$$

Option Pricing

- We now define the algorithm G which takes on the role of A described earlier, generates the superposition of n bits.

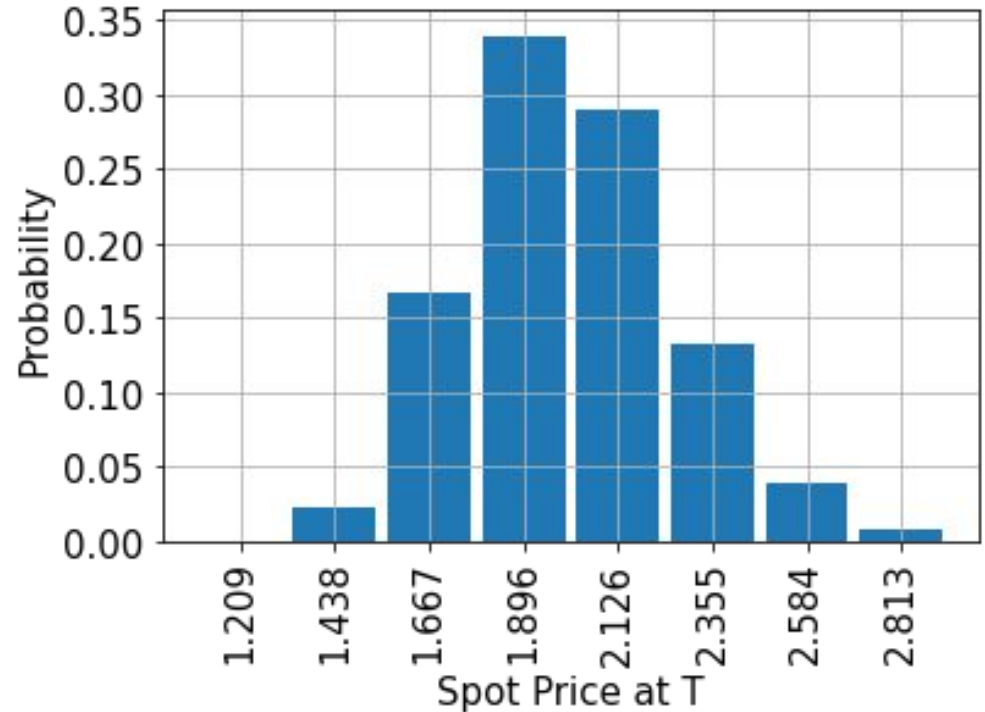
$$\mathcal{G}|0^n\rangle = \sum_{j=0}^{2^n-1} \sqrt{p_j}|j\rangle$$

- Similarly $v(x)$ function can be defined as

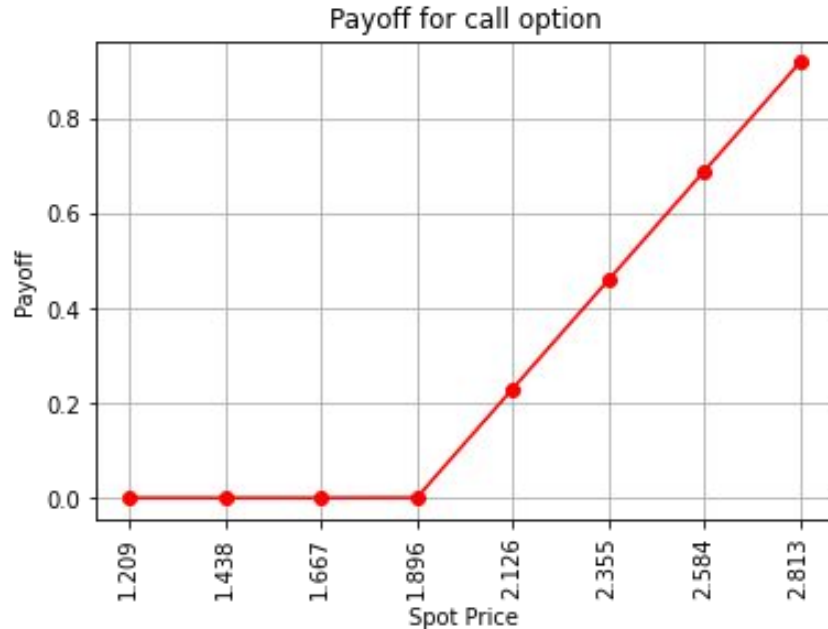
$$v_{\text{euro}}(x) = \max\{0, S_0 e^{\sigma x + (r - \frac{1}{2}\sigma^2)T} - K\}$$

Experiments - Lognormal

- Parameters
 - Spot price = 2
 - Volatility = 0.4
 - $r = 0.05$
 - $T = 40$ days
 - Strike price = 1.896



Experiments - Lognormal (Call)

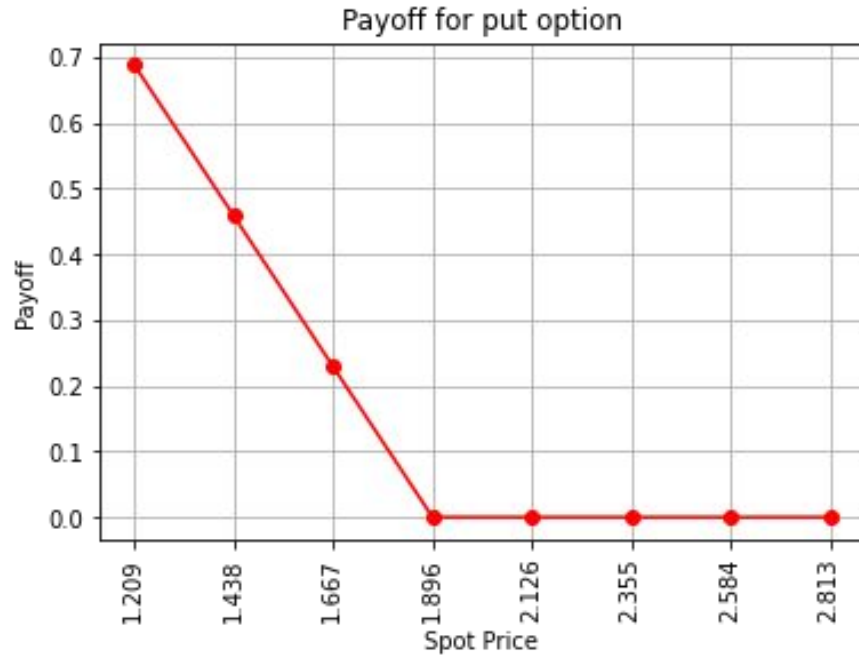


Exact expected value: 0.1623

Estimated expected value: 0.1681

Error percentage: 3.57%

Experiments - Lognormal (Put)



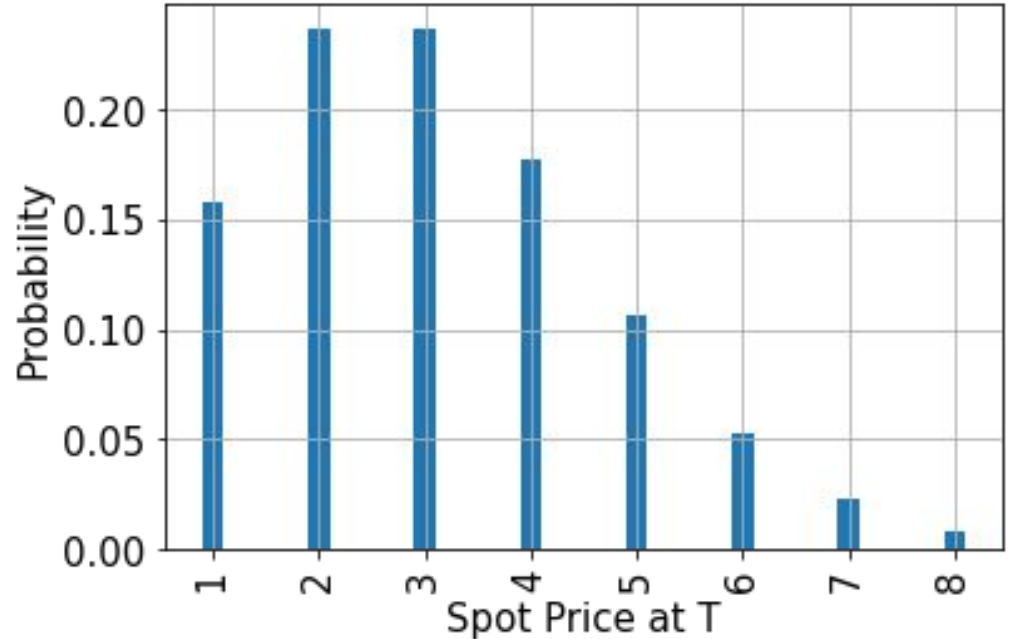
Exact expected value: 0.0490

Estimated expected value: 0.0565

Error percentage: 15.3%

Experiments - Poisson

- Binomial \rightarrow approximation
 \rightarrow Poisson
- Parameters
 - Mean = 3
 - Standard deviation = 2

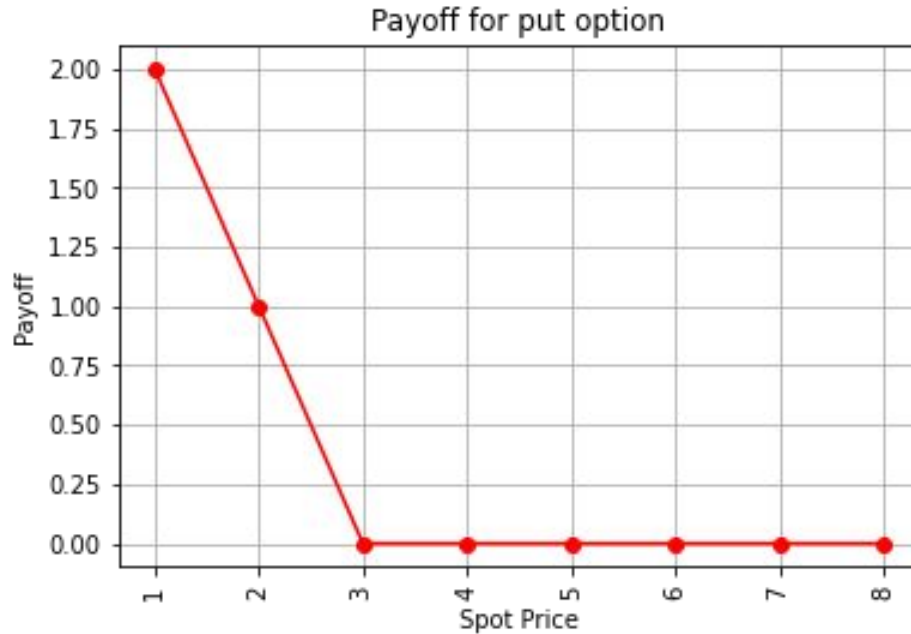


Experiments - Poisson (Call)



Exact expected value: 0.6845
Estimated expected value: 0.7198
Error percentage: 5.0%

Experiments - Poisson (Put)



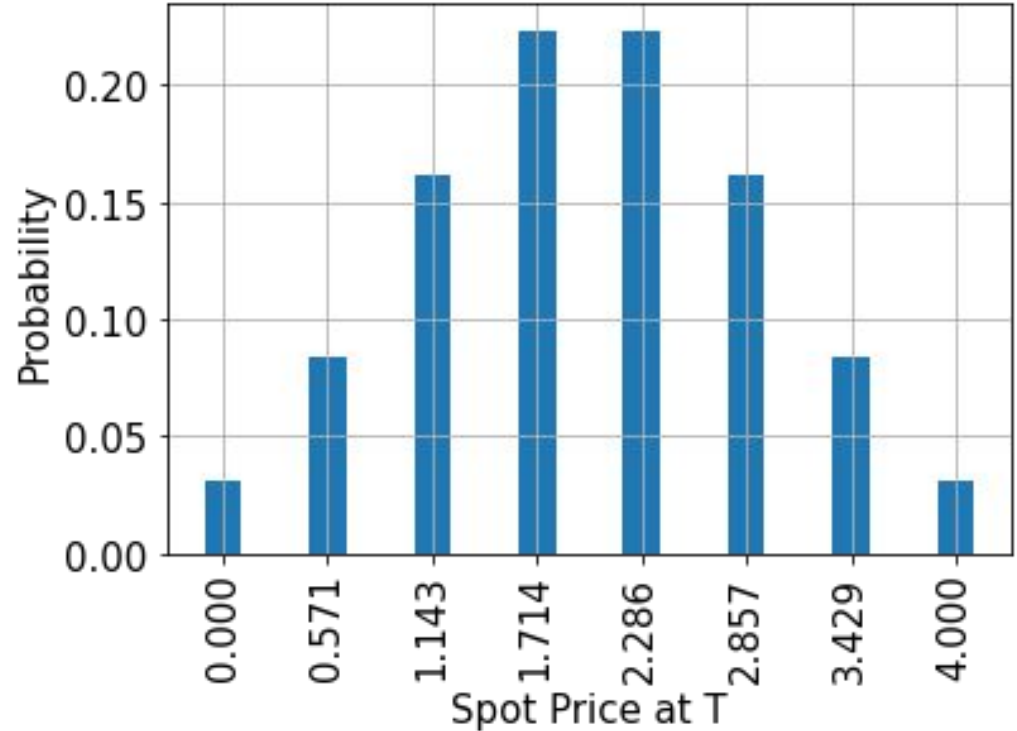
Exact expected value: 0.5524

Estimated expected value: 0.5565

Error percentage: 0.7%

Experiments - Gaussian

- Parameters
 - Mean = 2
 - Standard Deviation = 1



Experiments - Gaussian (Call)



Exact expected value: 0.3848

Estimated expected value: 0.3992

Error percentage: 3.7%

Experiments - Gaussian (Put)



Exact expected value: 0.3848

Estimated expected value: 0.3881

Error percentage: 0.8%

Conclusion

- For classical Monte Carlo:
 - Each sample is independent of other
 - By Central Limit Theorem, variance = $\sigma_{\hat{x}}^2 = \sigma_x^2 / t$
 - **For accuracy ϵ , $t=O(1/\epsilon^2)$ iterations.**
- For Quantum Monte Carlo:
 - Theorems (Amplitude estimation, Mean estimation for bounded functions, Mean estimation for bounded variance) show - **For accuracy ϵ , $t=O(1/\epsilon)$ iterations.**
- Hence, quadratic speedup!

Thank you!