

Q1:

For  $r$  under normal distribution with mean of 0 and assumes standard deviation with 0.1, and  $P_{t-1} = 100$ , the expected value of price and the standard deviation 1 period ahead under classical Brownian Motion ( $P_t = P_{t-1} + r$ ) are 100 and 0.1 respectively.

$$E[P_t] = P_{t-1} + E[r]$$

$$SD[P_t] = \sqrt{\text{Var}[P_t]} = \sqrt{\text{Var}[P_{t-1} + r]} = \sqrt{\text{Var}[P_{t-1}] + \text{Var}[r]}$$

For  $r$  under normal distribution with mean of 0 and assumes standard deviation with 0.1, and  $P_{t-1} = 100$ , the expected value of price and the standard deviation 1 period ahead under Arithmetic Return System ( $P_t = P_{t-1} \cdot (1+r)$ ) are 100 and 10 respectively.

$$E[P_t] = P_{t-1} \cdot (1 + E[r])$$

$$SD[P_t] = \sqrt{\text{Var}[P_t]} = \sqrt{(P_{t-1})^2 \cdot \text{Var}[r] + (\text{Var}[P_{t-1}] \cdot (1 + E[r])^2)}$$

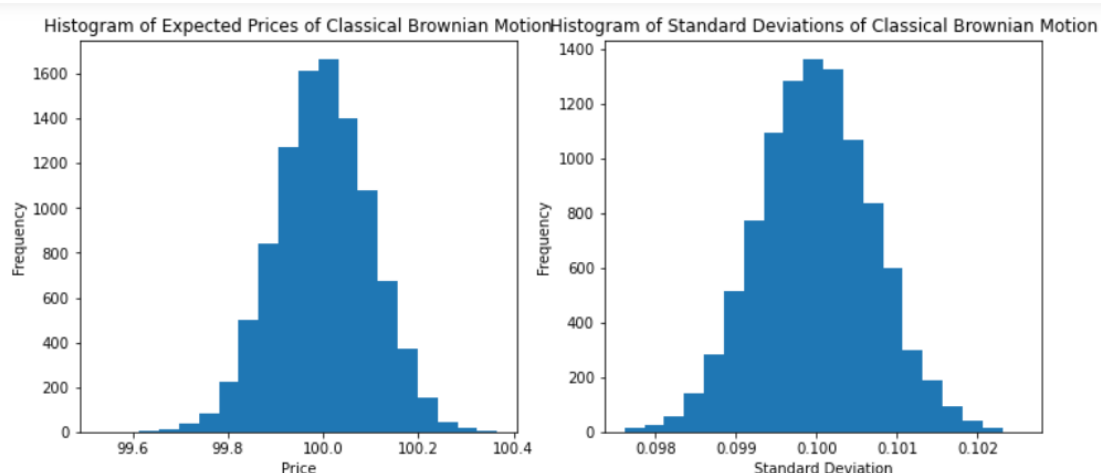
For  $r$  under normal distribution with mean of 0 and assumes standard deviation with 0.1, and  $P_{t-1} = 100$ , the expected value of price and the standard deviation 1 period ahead under Geometric Brownian Motion ( $P_t = P_{t-1} \cdot (e^{r \cdot t})$ ) are 100 and 10.08 respectively.

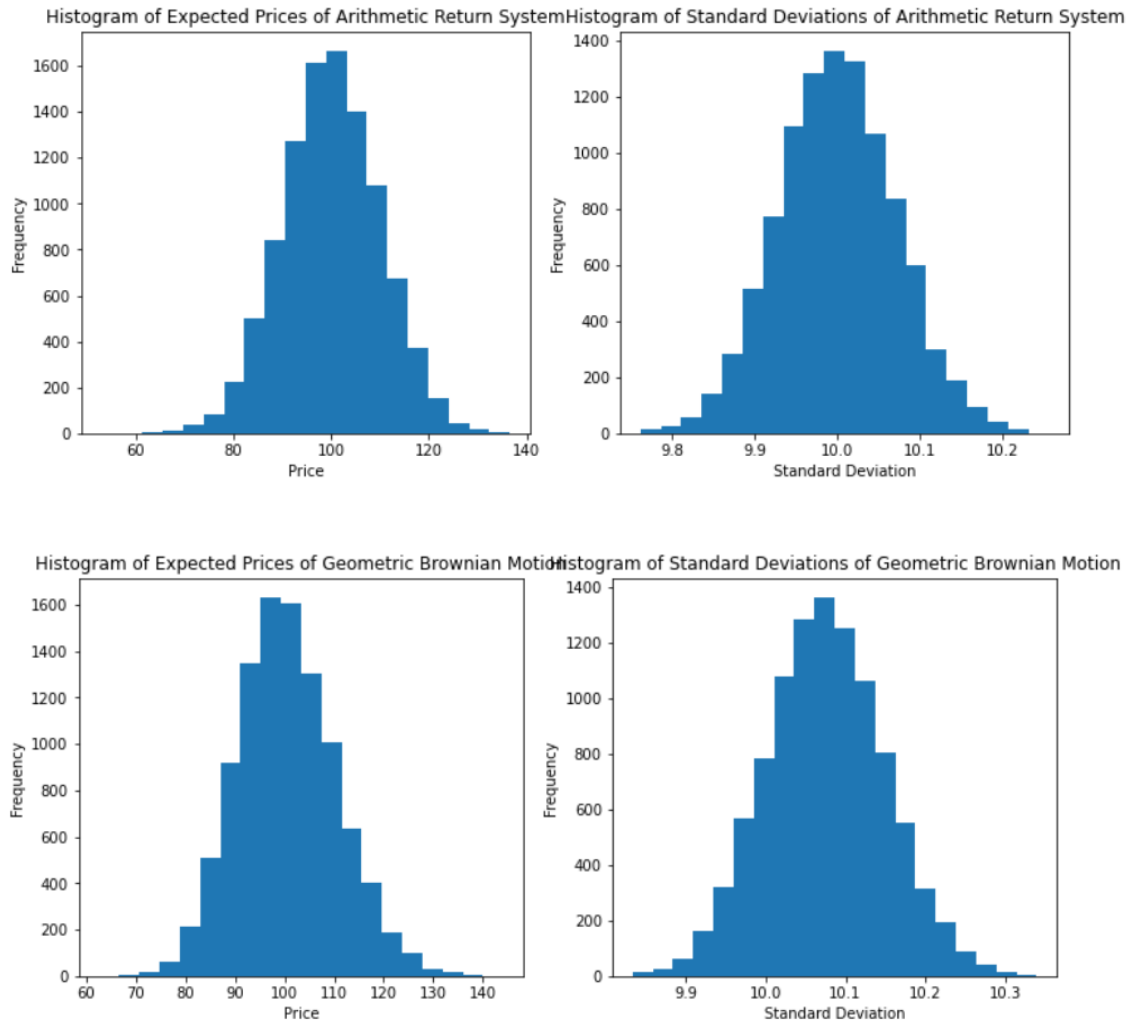
$$E[P_t] = P_{t-1} \cdot e^{E[r] \cdot t}$$

$$SD[P_t] = \sqrt{\text{Var}[P_t]} = \sqrt{(P_{t-1})^2 \cdot \text{Var}[r] \cdot (e^{2 \cdot E[r] \cdot t} - 1)}$$

For  $P_t = P_{t-1} + r$ , the expected value of price one period ahead is 100.00 and the standard deviation is 0.10  
 For  $P_t = P_{t-1} \cdot (1+r)$ , the expected value of price one period ahead is 100.00 and the standard deviation is 10.00  
 For  $P_t = P_{t-1} \cdot (e^{r \cdot t})$ , the expected value of price one period ahead is 100.00 and the standard deviation is 10.08

Here are graphs to show prices and standard deviation under different types of returns with a simulation of 10000 times:





It is clear that the mean and standard deviation match my expectation above.

I changed the standard deviation of  $r$  to 0.2 and do all steps again:

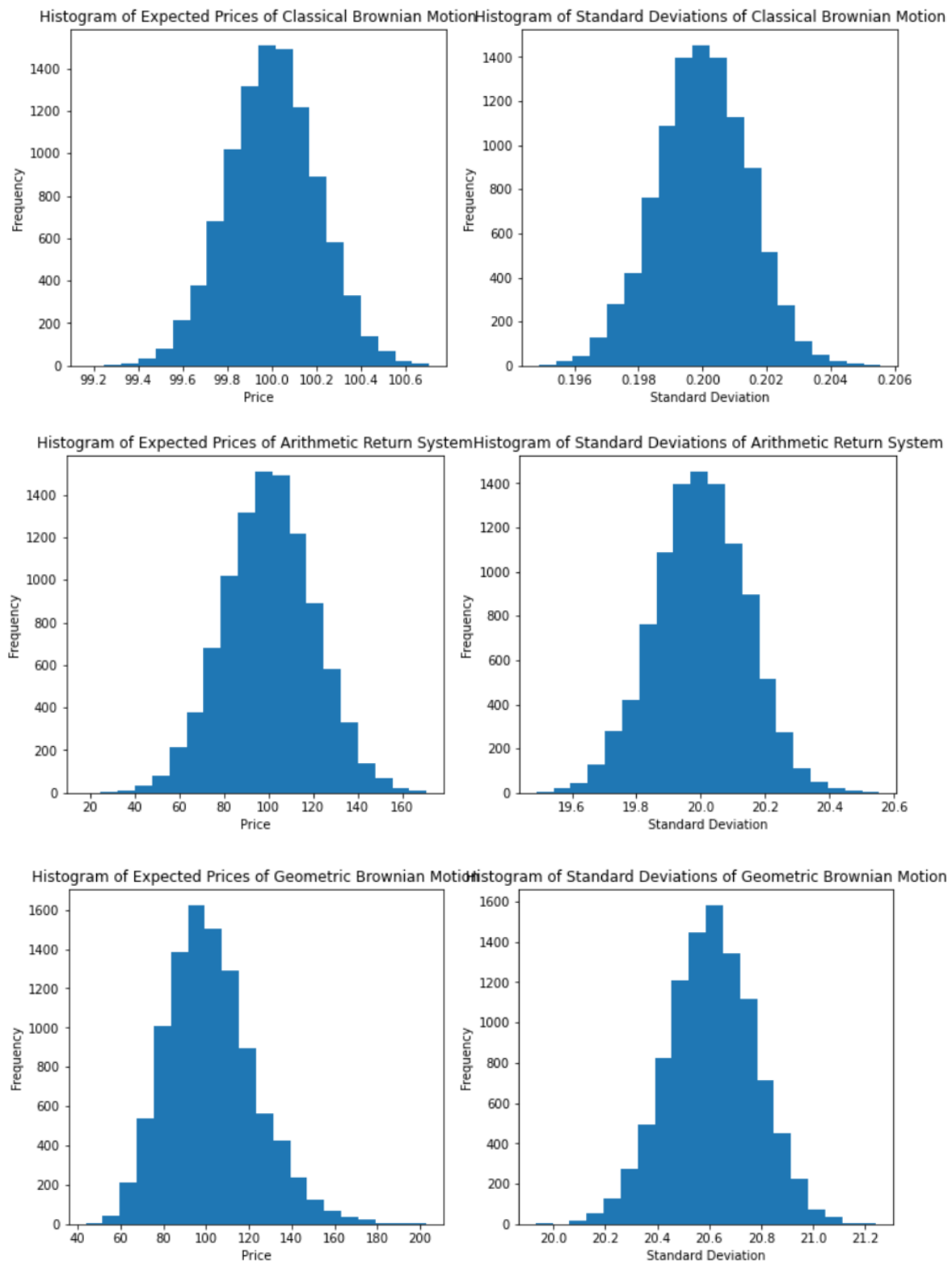
For  $r$  under normal distribution with mean of 0 and assumes standard deviation with 0.2, and  $P_{t-1} = 100$ , the expected value of price and the standard deviation 1 period ahead under classical Brownian Motion ( $P_t = P_{t-1} + r$ ) are 100 and 0.2 respectively.

For  $r$  under normal distribution with mean of 0 and assumes standard deviation with 0.2, and  $P_{t-1} = 100$ , the expected value of price and the standard deviation 1 period ahead under Arithmetic Return System ( $P_t = P_{t-1} \cdot (1+r)$ ) are 100 and 20 respectively.

For  $r$  under normal distribution with mean of 0 and assumes standard deviation with 0.2, and  $P_{t-1} = 100$ , the expected value of price and the standard deviation 1 period ahead under Geometric Brownian Motion ( $P_t = P_{t-1} \cdot (e^{rt})$ ) are 100 and 20.61 respectively.

For  $P_t = P_{t-1} + r$ , the expected value of price one period ahead is 100.00 and the standard deviation is 0.20  
 For  $P_t = P_{t-1} \cdot (1+r)$ , the expected value of price one period ahead is 100.00 and the standard deviation is 20.00  
 For  $P_t = P_{t-1} \cdot (e^{rt})$ , the expected value of price one period ahead is 100.00 and the standard deviation is 20.61

Here are graphs to show prices and standard deviation under different types of returns with a simulation of 10000 times:



It is clear that the mean and standard deviation match my expectation above as well.

Q2:

Here are the arithmetic returns for all prices:

	Date	SPY	AAPL	MSFT	AMZN	TSLA	\
1	2/15/2022 0:00	0.016127	0.023152	0.018542	0.008658	0.053291	
2	2/16/2022 0:00	0.001121	-0.001389	-0.001167	0.010159	0.001041	
3	2/17/2022 0:00	-0.021361	-0.021269	-0.029282	-0.021809	-0.050943	
4	2/18/2022 0:00	-0.006475	-0.009356	-0.009631	-0.013262	-0.022103	
5	2/22/2022 0:00	-0.010732	-0.017812	-0.000729	-0.015753	-0.041366	
..	...	...	...	...	...	...	
244	2/3/2023 0:00	-0.010629	0.024400	-0.023621	-0.084315	0.009083	
245	2/6/2023 0:00	-0.006111	-0.017929	-0.006116	-0.011703	0.025161	
246	2/7/2023 0:00	0.013079	0.019245	0.042022	-0.000685	0.010526	
247	2/8/2023 0:00	-0.010935	-0.017653	-0.003102	-0.020174	0.022763	
248	2/9/2023 0:00	-0.008669	-0.006912	-0.011660	-0.018091	0.029957	

	GOOGL	GOOG	META	NVDA	...	PNC	MDLZ	\
1	0.007987	0.008319	0.015158	0.091812	...	0.012807	-0.004082	
2	0.008268	0.007784	-0.020181	0.000604	...	0.006757	-0.002429	
3	-0.037746	-0.037669	-0.040778	-0.075591	...	-0.034949	0.005326	
4	-0.016116	-0.013914	-0.007462	-0.035296	...	-0.000646	-0.000908	
5	-0.004521	-0.008163	-0.019790	-0.010659	...	0.009494	0.007121	
..	...	...	...	...	...	...	...	
244	-0.027474	-0.032904	-0.011866	-0.028053	...	-0.004694	-0.011251	
245	-0.017942	-0.016632	-0.002520	-0.000521	...	-0.014451	0.003945	
246	0.046064	0.044167	0.029883	0.051401	...	-0.000368	-0.016473	
247	-0.076830	-0.074417	-0.042741	0.001443	...	-0.008469	-0.004456	
248	-0.043876	-0.045400	-0.030039	0.005945	...	-0.016588	-0.007717	

	MO	ADI	GILD	LMT	SYK	GM	TFC
1	0.004592	0.052344	0.003600	-0.012275	0.033021	0.026240	0.028572
2	0.005763	0.038879	0.009294	0.012244	0.003363	0.015301	-0.001389
3	0.015017	-0.046988	-0.009855	0.004833	-0.030857	-0.031925	-0.033380
4	0.007203	-0.000436	-0.003916	-0.005942	-0.013674	-0.004506	-0.003677
5	-0.008891	0.003243	-0.001147	-0.000673	0.008342	-0.037654	-0.002246
..	...	...	...	...	...	...	...
244	-0.001277	-0.002677	0.038211	0.004134	0.002336	-0.008916	-0.005954
245	0.001066	-0.007102	0.022012	0.021826	-0.041181	0.005106	-0.009782
246	-0.008518	0.019544	-0.003590	-0.001641	0.003573	0.001451	0.008669
247	-0.001289	-0.018009	-0.004416	0.002819	-0.015526	0.004106	-0.015391
248	-0.003656	0.004275	-0.001634	0.000937	-0.014391	0.001443	-0.016619

	TJX
1	0.013237
2	-0.025984
3	-0.028763
4	0.015038
5	-0.013605
..	...
244	0.001617
245	-0.004595
246	-0.003618
247	0.009363
248	0.005603

Here are the returns for META after removing the mean from the series:

```
1      0.015175
2     -0.020165
3     -0.040761
4     -0.007446
5     -0.019774
...
244   -0.011850
245   -0.002503
246    0.029899
247   -0.042725
248   -0.030022
Name: META, Length: 248, dtype: float64
```

Here is the VaR using a normal distribution:

```
0.06560156967533282
```

Here is the VaR using a normal distribution with an Exponentially Weighted variance ( $\lambda = 0.94$ ):

```
0.05253206870346172
```

Here is the VaR using a MLE fitted T distribution:

```
0.0022964447067896376
```

Here is the VaR using a fitted AR(1) model:

```
0.06592848204123766
```

Here is the VaR using a Historic Simulation:

```
0.05223042912767907
```

Comparing VaR values from different simulations, I found that the VaR values using an MLE-fitted T distribution are relatively smaller than the other four simulations. I think the reason behind it is that the T distribution has fatter tails than the normal distribution, which means that the probability of enormous loss and gain is larger so that under the same confidence level (95%), the value on the x-axis which is the return will be smaller, which is a smaller VaR.

Q3:

Here are VaR values under exponentially weighted covariance and normal distribution:

I chose normal distribution to calculate VaR because it is an easy way and I found that portfolio A, B, C and all have a smaller VaR compare to VaR using Exponentially weighted covariance.

```
portfolio:A
VaR Using Exponentially weighted variance: 24302.5420929740
portfolio:B
VaR Using Exponentially weighted variance: 21551.2756024795
portfolio:C
VaR Using Exponentially weighted variance: 21453.6784730695
portfolio:ALL
VaR Using Exponentially weighted variance: 22131.3735012223
```

```
VAR Using normal distribution:
12571.119582860485
21857.25085508876
30760.215902723063
67206.27308362359
```

