Week07 project

Wengi Cai

Problem1

1.1 The result of comparing the values between the two methods for both a call and a put:

```
delta: (call, put)
GBSM: 0. 08301107089626869, -0. 9169889291037313
FDD: 0. 08297130374668171, -0. 9165496329472944
gamma: (call, put)
GBSM: 0. 016830979206204362, 0. 016830979206204362
FDD: 0.016822911064195978, 0.016822951920403284
vega: (call, put)
GBSM: 6.942036604441163, 6.942036604441163
FDD: 6. 938653056250743, 6. 93865305626673
theta: (call, put)
GBSM:-8.126522359668838, -1.9409914783019566
FDD:-8.126308803761084, -1.9407779203106656
rho: (call, put)
GBSM:1.1025939156368187, -13.758003122735788
FDD:-0.030359909416688424, -1.2427313238703164
carry rho: (call, put)
GBSM:1.132953825011723, -12.515271800549371
FDD:1.1329550097096686, -12.515270634423814
```

The reasons for discrepancy between the two methods for calculating Rho (GBSM vs. FDD) may be: 1) The FDD method is highly sensitive to numerical errors (step size Δr); 2) The underlying pricing model in FDD may differ in assumptions compared to the GBSM closed-form solution.

Carry Rho (sensitivity of option value to cost of carry, b) is typically not used for American options with dividends. This is because the cost of carry (b) does not directly appear as a parameter in pricing models for such options.

1.2 The binomial tree valuation for American options with and without discrete dividends:

```
Binomial tree value without dividend for call: 0.341603945513609
Binomial tree value without dividend for put: 14.020014650639181
Binomial tree value with dividend for call: 0.30041508863955924
Binomial tree value with dividend for put: 14.560397204371986
```

1.3 The Greeks of the call and the put are:

delta: (call, put)

 $0.\ 07564529288392463\ -0.\ 9304253539674789$

gamma: (call, put)

 $0.\ 01624307464913788\ \ 0.\ 014779533669457834$

vega: (call, put)

 $6.\ 42956107974818\ \ 6.\ 0562654358991$

theta: (call, put)

-7.5594852571106985 -0.8555281205477883

rho: (call, put)

 $0.\ 9528888093222332\ -12.\ 430363528616262$

1.4 Sensitivity to dividend amount:

Call: -0.034; Put: 0.934

Problem2

2.1 Calculate Mean, VaR and ES:

	Mean	VaR(\$)	ES (\$)
Call	3.097361	6.799017	6.799953
CallSpread	2.659462	4.589017	4.589953
CoveredCall	149.117511	7.260838	10.602005
ProtectedPut	156.756094	3.003057	4.272518
Put	9.631896	-0.268636	0.799042
PutSpread	5.011379	-0.219114	0.373946
Stock	150.074736	11.310838	14.652005
Straddle	12.729257	0.973351	0.984207
SynLong	-6.534535	17.095658	19.424308

2.2 Calculate VaR and ES using Delta-Normal:

	Mean	VaR	ES
Portfolio			
Call	0	9.922584	12.443321
CallSpread	0	3.36575	4.220786
CoveredCall	0	3.717425	4.661801
ProtectedPut	0	9.887767	12.399659
Put	0	2.265449	2.840964
PutSpread	0	1.102982	1.383185
Stock	0	12.108229	15.184207
Straddle	0	7.657136	9.602357
SynLong	0	12.188033	15.284285

2.3 Compare these results to last week's results (rerun the code from last week using these inputs):

```
Portfolio Current Value ($) Mean Simulated Value ($) \
                  68000.0
          Call
1 CallSpread 2 CoveredCall 1469800.0 3 ProtectedPut 1540400.0 48500.0
                                                         NaN
                                                         NaN
                                                  83.861954
                        48500.0
                                                   4. 371051
          Put
    PutSpread 30100.0
Stock 1510300.0
Straddle 116500.0
5
                                                    1.394586
6
                                                  165. 063209
                                                         NaN
8
                          19500.0
                                                         NaN
       SynLong
  VaR 95% ($ Loss) ES 95% ($ Loss)
                NaN
1
                NaN
                                 NaN
2
               NaN
                                 NaN
                      1.540321e+06
3
     1. 540225e+06
                     4.849590e+04
     4.849190e+04
                     3. 009898e+04
5
      3.009294e+04
6
      1. 510122e+06
                       1.510136e+06
                                 NaN
                NaN
```

Compared to European options, the values of American options, as calculated in this week's problem, are generally higher. This is consistent with observations from last week's assignment, though there are exceptions—such as puts that are unlikely to be exercised, which may have lower values. Typically, higher values or returns are associated with greater risks, which explains why the VaR and ES for American options are often higher.

Problem3

3.1 The result of expected returns for each stock (annualized) in Python:

```
Expected Returns for Each Stock (Annualized):
AAPL: 0.000705
META: 0.001032
UNH: 0.000235
MA: 0.000591
MSFT: 0.000779
NVDA: 0.001424
HD: 0.000597
PFE: 0.000334
AMZN: 0.000963
BRK-B: 0.000483
PG: 0.000342
XOM: 0.000311
TSLA: 0.001256
JPM: 0.000525
V: 0.000568
DIS: 0.000513
GOOGL: 0.000861
JNJ: 0.000295
BAC: 0.000620
CSCO: 0.000535
```

3.2 The result of the annual covariance matrix for the given stocks is:

Annua1	ized Cova	riance Mat	rix:					
	AAPL	META	UNH	MA	MSFT	NVDA	HD	\
AAPL	0.050501	0.020870	-0.001577	0.010380	0.021561	0.033939	0.010867	
META	0.020870	0. 136272	-0.011660	0. 015539	0.041722	0.077580	0.010079	
UNH	-0.001577	-0.011660	0.049432	0.005066	-0.001927	-0.016474	0.006551	
MA	0.010380	0. 015539	0.005066	0. 027007	0.012275	0.021657	0.011621	
MSFT	0.021561	0.041722	-0.001927	0.012275	0. 039604	0.045418	0.011498	
NVDA	0. 033939	0.077580	-0.016474	0.021657	0.045418	0. 253770	0.023843	
HD	0.010867	0.010079	0.006551	0.011621	0.011498	0.023843	0.042762	
PFE	0.002864	0.000547	0.010636	0.003087	0.005221	-0.016984	0.005594	
AMZN	0.024202	0.064354	-0.006390	0.015380	0. 035189	0.067779	0.016908	
BRK-B	0.006499	0.008382	0.006006	0.010321	0.006653	0.003171	0.010077	
		-0.000426	0.007212	0.004023	0.002121	-0.009521	0.003260	
XOM		-0.005812	0.001574		-0. 005784		0.005945	
TSLA	0. 046972	0. 033953	0. 001457	0. 021594		0. 071756	0. 028368	
JPM	0.001670	0.006064	0. 004938	0.008724	0. 004695	0.005717	0.013261	
V	0.010306	0. 014285	0.003748	0. 019691	0. 012743	0.018001	0.010632	
DIS	0.003784	0. 010023	0.000456	0.007901	0. 008303	0.015492	0. 012559	
GOOGL	0. 025260		-0.004703	0. 011537	0. 029757	0. 051591	0.007758	
JNJ		-0. 005947	0.008443		-0. 000898		0.006870	
BAC	0.005001	0.005183	0. 005151	0.006302		-0.000161	0. 022778	
CSC0	0. 008293	0.007240	0. 005028	0.009910	0. 008667	0. 010232	0. 013129	
	PFE	AMZN	BRK-B	PG	XOM	TSLA	JPM	\
AAPL	0.002864	0. 024202		-0. 000199		0. 046972	0.001670	\
META	0.002804	0. 024202		-0. 000199		0. 040972	0.001070	
UNH		-0. 006390	0. 006362	0.000420	0.003812	0. 033933	0.004938	
MA	0. 010030	0. 000390	0. 010321	0. 001212	0.001374	0. 001437	0.004938	
MSFT	0. 005087	0. 015380	0. 010321		-0. 005784	0. 021394	0. 003724	
NVDA	-0. 016984	0. 055169		-0. 009521		0. 029251	0.004033	
HD	0.005594	0. 016908	0. 010077	0. 003321	0. 017233	0. 028368	0. 003717	
PFE	0.062464	0. 010308	0. 010077	0.003200	0.003940	0. 020300	0. 013201	
AMZN	0.002464	0.002400		-0.001785		0. 002340	0.010240	
BRK-B	0.002400	0.001074	0. 016801	0.001700	0. 008345	0. 009049	0.003200	
PG		-0. 001785	0. 010501	0. 022672		-0. 003049	0. 000597	
XOM		-0. 005728	0. 004301	0. 002867		-0. 001924	0. 011345	
TSLA	0. 002940	0. 046419		-0. 003999		0. 293876	0. 020562	
JPM	0. 010246	0. 008206	0. 013116	0. 000597	0. 011345	0. 020562	0. 033249	
V	0. 004419	0. 013531	0.009401	0. 004643	0. 002715	0. 021799	0. 008395	
DIS	0. 008145	0. 013054		-0. 000892	0. 009776	0. 026769	0. 011302	
GOOGL	0.003924	0. 045175		-0. 000272		0. 031925	0. 006162	
JNJ		-0.002814	0. 007518	0. 005866		-0.001402	0. 007898	
BAC	0.010263	0.008630	0. 014265	0.002463	0. 016620	0.028866	0.028629	
CSC0	0. 010953	0. 018381	0.007840	0.002357	0.008121	0.017523	0.009708	
	V	DIS	GOOGL	JNJ	BAC	CSC0		
AAPL	0. 010306	0.003784	0.025260			0.008293		
META	0.014285	0.010023		-0.005947		0.007240		
UNH	0.003748		-0.004703	0.008443		0.005028		
MA	0. 019691	0.007901	0. 011537	0.004104		0.009910		
MSFT	0.012743	0.008303		-0. 000898		0.008667		
NVDA	0. 018001	0. 015492		-0. 020411		0. 010232		
HD	0. 010632	0. 012559	0. 007758		0. 022778	0. 013129		
PFE	0.004419	0.008145	0.003924	0. 016889				
AMZN	0. 013531	0. 013054		-0.002814		0. 018381		
BRK-B	0.009401	0.008012	0.008250	0.007518	0. 014265	0.007840		
PG		-0.000892		0.005866		0. 002357		
XOM	0.002715		-0.004480	0.005893		0.008121		
TSLA	0. 021799			-0.001402		0. 017523		
JPM	0.008395	0. 011302	0.006162	0.007898	0. 028629			
V DIC	0.022884		0. 011023	0.004906		0.009791		
DIS	0. 005352 0. 011023	0.071514	0.009315	0. 001750 -0. 002217	0. 015920 0. 006355	0. 010569		
GOOGL	0. 011023	0.009315	-0. 002217	0. 023668	0. 006355	0. 013982		
JNJ BAC	0.004906	0. 001750	0. 006355	0. 023008	0. 010143	0. 006289 0. 011909		
CSC0	0.000003	0. 010569	0. 013982	0. 010143		0. 011909		
COCO	0.009191	0. 010009	0.013902	0.000209	0. 011909	0. 040733		

3.3 The super-efficient portfolio is:

Optimal	Portfolio	Weights
Stock	Weight (%)	
AAPL	0.0	
META	30.0	
UNH	0.0	
MA	0.0	
MSFT	0.0	
NVDA	30.0	
HD	0.0	
PFE	0.0	
AMZN	10.0	
BRK-B	0.0	
PG	0.0	
XOM	0.0	
TSLA	30.0	
JPM	0.0	
V	0.0	
DIS	0.0	
G00GL	0.0	
JNJ	0.0	
BAC	0.0	
CSC0	0.0	

I set up a diversification constraint: restrict the maximum allocation for each stock to avoid overly concentrated portfolios (Limit each stock to a maximum of 30% of the portfolio).

From the debug outputs, the portfolio weights are highly concentrated in a few stocks (META, NVDA, TSLA, and AMZN).