

MFE 409: Financial Risk Management

Problem Set 1

Jiaqi Li

April 6, 2019

1 VaR for an exponential distribution

1. Assume an asset value in 10 days follows an exponential distribution with mean W_0 . Derive a formula for the value-at-risk for confidence c and reference level W_0 for the asset. Apply with $W_0 = 200$ and $c = 99.9\%$.

$$\begin{aligned}\int_0^{W_0 - VaR} \lambda e^{-\lambda W} dW &= 1 - c \\ [-e^{-\lambda W}]_0^{W_0 - VaR} &= 1 - c \\ -e^{-\lambda(W_0 - VaR)} + e^{-\lambda \times 0} &= 1 - c \\ \ln(c) &= -\lambda(W_0 - VaR) \\ VaR &= \frac{\ln(c)}{\lambda} + W_0\end{aligned}$$

For $\mu = W_0 = 200$ and $c = 99.9\%$:

$$\begin{aligned}\frac{1}{\lambda} &= \mu = 200 \Rightarrow \lambda = \frac{1}{200} \\ VaR &= \frac{\ln(99.9\%)}{\frac{1}{200}} + 200 = 199.79\end{aligned}$$

2. Now assume you are short this asset. Compute your value-at-risk for confidence c and apply with the same numerical values.

For shorting this asset, we want $W \geq W_0 + VaR$, then we can derive formula for VaR :

$$\begin{aligned}\int_{W_0 + VaR}^{+\infty} \lambda e^{-\lambda W} dW &= 1 - c \\ \left[-W e^{-\lambda W} - \frac{1}{\lambda} e^{-\lambda W} \right]_{W_0 + VaR}^{+\infty} &= 1 - c \\ [-e^{-\lambda W}]_{W_0 + VaR}^{+\infty} &= 1 - c \\ e^{-\lambda(W_0 + VaR)} &= 1 - c \\ VaR &= -\frac{\ln(1 - c)}{\lambda} - W_0\end{aligned}$$

Then for $W_0 = 200$ and $c = 99.9\%$, $VaR = -\frac{\ln(1 - 99.9\%)}{\frac{1}{200}} - 200 = 1181.551$.

3. Comment on the similarity or difference between the results in the two questions.

The above 2 VaRs are both from an exponential distribution. Because the exponential distribution is not symmetric, tails on different sides of the distribution is also different, which gives us 2 different VaRs from the same distribution.

4. Repeat the previous questions with expected shortfall.

For longing asset:

$$\begin{aligned}
 ES &= W_0 - E(W|W \leq W_0 - VaR) \\
 &= W_0 - \frac{\int_0^{W_0 - VaR} W \lambda e^{-\lambda W} dW}{\int_0^{W_0 - VaR} \lambda e^{-\lambda W} dW} \\
 &= W_0 - \frac{-(W_0 - VaR)e^{-\lambda(W_0 - VaR)} - \frac{1}{\lambda}e^{-\lambda(W_0 - VaR)} + \frac{1}{\lambda}}{1 - e^{-\lambda(W_0 - VaR)}} \\
 &= \frac{(W_0 - VaR)e^{-\lambda(W_0 - VaR)}}{1 - e^{-\lambda(W_0 - VaR)}} \\
 ES_{long} &= \frac{(200 - 199.79)e^{-\frac{1}{200}(200 - 199.79)}}{1 - e^{-\frac{1}{200}(200 - 199.79)}} = 199.895 \text{ where } VaR = 199.79
 \end{aligned}$$

For shorting asset:

$$\begin{aligned}
 ES &= E(W|W \geq W_0 + VaR) - W_0 \\
 &= \frac{\int_{W_0 + VaR}^{+\infty} W \lambda e^{-\lambda W} dW}{\int_{W_0 + VaR}^{+\infty} \lambda e^{-\lambda W} dW} - W_0 \\
 &= \frac{(W_0 + VaR)e^{-\lambda(W_0 + VaR)} + \frac{1}{\lambda}e^{-\lambda(W_0 + VaR)}}{e^{-\lambda(W_0 + VaR)}} - W_0 \\
 &= \frac{1}{\lambda} + VaR \\
 ES_{short} &= 200 + 1181.551 = 1381.551 \text{ where } VaR = 1181.551
 \end{aligned}$$

2 VaR for mixtures

You have to allocate \$1bn in the stock market. You are discussing with your partner regarding the volatility of returns. She has a view that, in line with historical averages, the volatility of returns will be of $\sigma_1=12\%$ in the next year. However, you believe that volatility will be higher, in the orders of $\sigma_2=20\%$ for the next year. After discussing with your partner, you agree in the following way: stocks returns follow a normal distribution with mean μ and σ_1 with probability π and normal distribution with mean μ and σ_2 with probability $1 - \pi$. For now assume that $\mu = 8\%$ and $\pi = 0.7$.

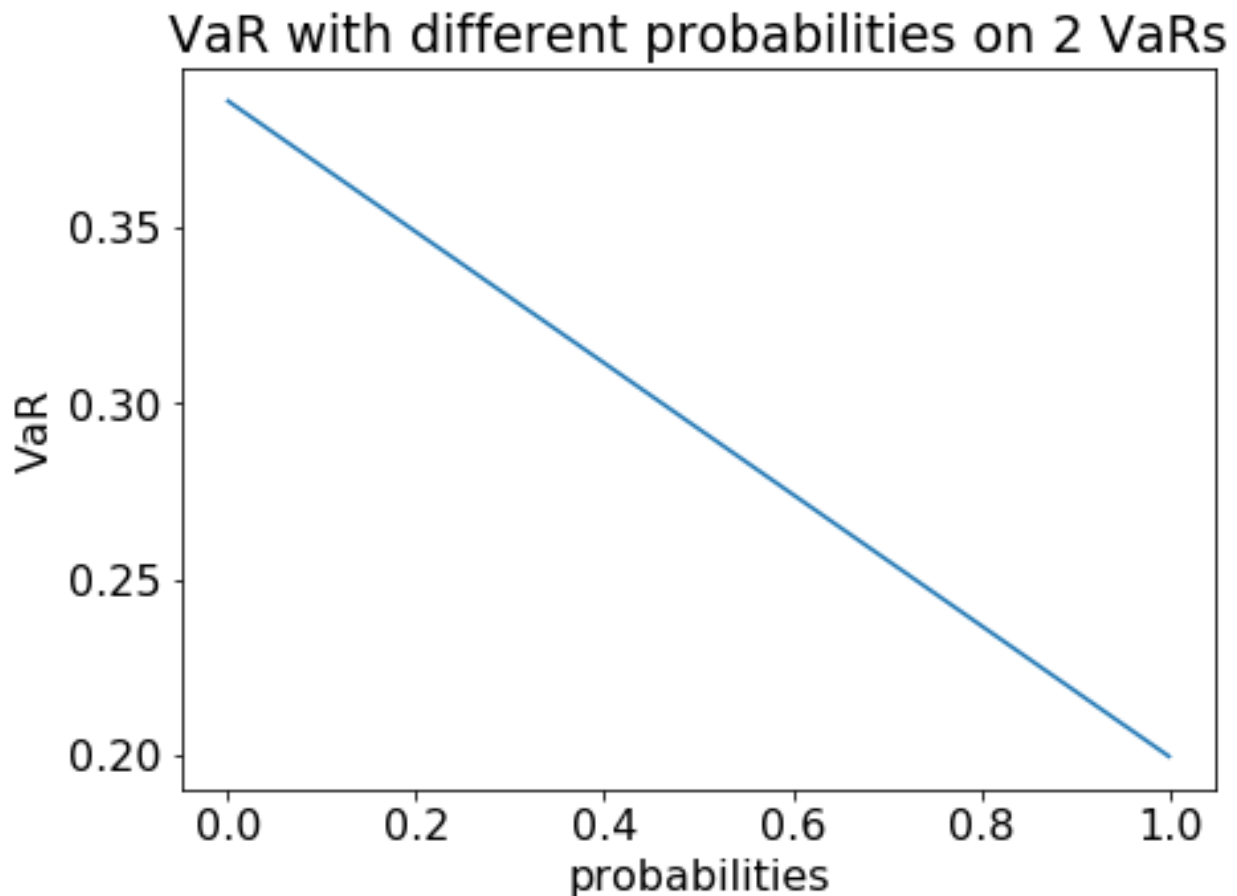
1. Compute the 1-year 99% VaR with your view, with your partner view, and with the common view. Compare results and provide a very brief explanation as if you were presenting to your manager.

$$VaR_{my} = -(0.08 - 2.33 \times 0.2) = 0.386$$

$$VaR_{partner} = -(0.08 - 2.33 \times 0.12) = 0.1996$$

$$VaR_{combined} = 0.386 \times 0.3 + 0.1996 \times 0.7 = 0.25552$$

2. To understand the role of π , plot a chart with π between 0 and 1 on the horizontal axis which and the corresponding VaR on the vertical axis. Comment on your results.



Based on the plot, π decides how volatility of the return will be. We can explain π as a bargaining power, which means who has the stronger power decides what the volatility should be.

3. More challenging. After presenting your common view to your manager, you are challenged with an alternative view about volatility: σ is timevarying. The volatility trader suggests that a sensible model for sigma is a gamma distribution. Explain in as many details as possible (either derive of formula or use a computer program) how to compute the VaR of your portfolio when returns have a normal distribution conditional on σ and σ is distributed according to a Gamma distribution.

Let $\sigma \sim \Gamma(\alpha, \beta)$ and assume $\alpha, \beta > 0$

$\Gamma(\alpha, \beta)$ has PDF of $f(x) = \frac{\beta^\alpha}{\Gamma(\alpha)} x^{1-\alpha} e^{-\beta x}$

$$E(\sigma) = \frac{\alpha}{\beta}$$

$$W - W_0 \sim N(\mu, \Gamma(\alpha, \beta))$$

$$W = W_0 + N(\mu, \Gamma(\alpha, \beta))$$

$$W \sim N(W_0 + \mu, \Gamma(\alpha, \beta))$$

$$\begin{aligned} VaR &= W_0 - W \\ &= W_0 - [W_0 + \mu + z(c)\Gamma(\alpha, \beta)] \\ &= -[\mu + z(c)\Gamma(\alpha, \beta)] \\ &= -\left[\mu + z(c) \frac{\beta^\alpha}{\Gamma(\alpha)} W^{1-\alpha} e^{-\beta W}\right] \end{aligned}$$

We can use this formula for to compute VaR with volatilities following a gamma distribution.

If we want to do the simulation, we can simulate volatility (σ) first, then plug σ into the formula to compute VaR for each σ , then take mean of all VaRs to estimate the expected value of VaR.

3 VaR for options

You will need to write code to solve this question. You are a hedge fund manager with \$100 million of capital, and you can take any long or short position as long as you keep the 99% 10-day VaR for your portfolio under your capital. You have limited liability. You have access to long and short positions on a stock, 3-month European calls and puts on the underlying, and risk free bonds. You choose your positions today and must hold them for the next 10 days. Assume the risk-free asset has annual rate of return $r = 2\%$, and the stock price S_t has dynamics:

$$\frac{dS_t}{S_t} = \mu dt + \sigma dW_t$$

with $\mu = 7\%$ and $\sigma = 16\%$ (again, in annual units), and the current value $S_0 = 50$. Assume there are 252 trading days in a year.

1. Find a formula for the 10-day VaR for one share of the stock as a function of μ and σ . Hint: if X is a random variable with quantile c equal to x_0 , what is the quantile c of $g(X)$ if g is a monotone function?

First derive the expression of the stock price as following:

$$dS_t = S_t \mu dt + S_t \sigma dW_t$$

Let $X = \log(S_t)$, then applies Ito's lemma on X we could get:

$$dX = 0 \times dt + \frac{1}{S_t} \times dS_t - \frac{1}{2} \frac{1}{S_t^2} \times dS_t^2$$

Insert dS_t into dX and we could get:

$$\begin{aligned} dX &= (\mu dt + \sigma dW_t) - \left(\frac{1}{2}\sigma^2 dt\right) \\ dX &= \left(\mu - \frac{1}{2}\sigma^2\right) dt + \sigma dW_t \\ d(\log(S_t)) &= \left(\mu - \frac{1}{2}\sigma^2\right) dt + \sigma dW_t \\ \int_t^T d(\log(S_t)) &= \int_t^T \left(\mu - \frac{1}{2}\sigma^2\right) dt + \int_t^T \sigma dW_t \\ \log(S_T) - \log(S_t) &= \left(\mu - \frac{1}{2}\sigma^2\right)(T - t) + \sigma(W_T - W_t) \\ \log(S_T) &= \log(S_t) + \left(\mu - \frac{1}{2}\sigma^2\right)(T - t) + \sigma(W_T - W_t) \\ S_T &= S_t e^{\left(\mu - \frac{1}{2}\sigma^2\right)(T-t) + \sigma(W_T - W_t)} \end{aligned}$$

Thus, S_T is lognormal distribution with mean = $\left(\mu - \frac{1}{2}\sigma^2\right)(T - t)$ and $sd = \sigma\sqrt{T - t}$

In this case, let $Z = \left(\mu - \frac{1}{2}\sigma^2\right)(T - t) + \sigma(W_T - W_t)$, then $Z \sim N\left(\left(\mu - \frac{1}{2}\sigma^2\right)(T - t), \sigma\sqrt{T - t}\right)$

Since Z is a random variable and $g(Z) = e^Z$ is a monotone function, the quantile c of $g(Z)$ is:

$$e^{\left(\mu - \frac{1}{2}\sigma^2\right)(T-t) + z(c)\sigma\sqrt{T-t}} \text{ and we call this } Z^*$$

Since the stock price follows a lognormal distribution and we want to get the VaR of the stock price

at date 10 where $T = \frac{10}{252}$, $\mu = 0.07$, $\sigma = 0.16$, $S_t = 50$ where $t = 0$, $c = 0.99$, $r = 0.02$, $W_0 = 100$,

Then,

$$\begin{aligned} VaR &= S_t - S_T^* \\ VaR &= S_t - S_t \times e^{Z^*} \\ VaR &= S_t - S_t e^{\left(\mu - \frac{1}{2}\sigma^2\right)(T-t) + z(c)\sigma\sqrt{T-t}} \\ VaR &\approx 3.47 \end{aligned}$$

2. If you can only invest in stocks and bonds and want to maximize the average return on equity (subject to the VaR constraint), which portfolio do you choose?

Let w be the weight of the stock and $(1 - w)$ be the weight of the bond, then the return is:

$$\mu_{portfolio} = R_{portfolio} = wR_{stock} + (1 - w)R_{rf}$$

$$\sigma_{portfolio} = w\sigma_{stock}\sqrt{T}$$

with $R_{stock} = \left(\mu - \frac{1}{2}\sigma^2\right)(T - t) = 0.0022698$, $R_{rf} = 0.02 \times (T - t)$, $\sigma_{stock} = 0.16$,

$$\begin{aligned}
VaR &= -(\mu_{portfolio} + z(c)\sigma_{portfolio}) \times 100 \text{ millions} \\
100 \text{ millions} &= -(wR_{stock} + (1-w)R_{rf} + z(c)w\sigma_{stock}\sqrt{T}) \times 100 \text{ millions} \\
w &= -\frac{1 + R_{rf}}{R_{stock} - R_{rf} + z(c)\sigma_{stock}\sqrt{T}} \\
w &\approx 13.75
\end{aligned}$$

Then we should long $13.75 \times 100 = 1375$ billions stocks and short $-(1 - 13.75) \times 100 = 1275$ billions bonds.

3. If you can only invest in ATM calls and bonds, which portfolio do you choose?

In that case, we should long the call options as many as possible, we can compute the option weight by simulations as following:

```

51 #####
52 #                                     3.Question 3                                     #
53 #####
54 S0 = 50
55 r = 0.02
56 sd = 0.16
57 T = 1/4 #3 months maturity
58 paths = 1000 #1000 simulations
59 steps = 21*3 #91 days in 3 months
60 K = 50 #ATM call option stick price
61
62 stock = StockPrices(S0,r,sd,T,paths,steps) #simulate 10000 stock prices
63 S10 = stock[:,10] #take stock prices at day 10
64 C0 = f_BS(S0,K,T,sd,r,"call")
65 #compute Call option price today using black-scholes
66 C10 = f_BS(S10,K,T-10/252,sd,r,"call")
67 #compute call option prices at day 10 for all simulations
68 Cc = np.quantile(C10,0.01)
69 #take the 0.01th option price at day 10 from the distribution
70 VaR_C = C0 - Cc #compute VaR
71
72 N_Option = 100/VaR_C #number of options to buy
73 Cost_C = C0*N_Option #total cost of all the options bought
74 Bond_C = 100-Cost_C #total bond to buy/short
75 w_option_C = Cost_C/(Cost_C+Bond_C) #weight on call option
76 w_bond_C = 1 - w_option_C #weight on bond
77

```

By doing so, we can get:

```

In [212]: w_option_C
Out[212]: 1.2466926925192754

In [213]: w_bond_C
Out[213]: -0.2466926925192754

```

This mean: long $1.2467 \times 100 = 124.67$ millions of call options

short $0.2467 \times 100 = 24.67$ millions of bonds at risk free rate

4. If you can only invest in ATM puts and bonds, which portfolio do you choose?

In that case, we should short the put options as many as possible, we can compute the option weight by simulations as following:

```
79 #####
80 #                                     3.Question 4                                     #
81 #####
82 S0 = 50
83 r = 0.02
84 sd = 0.16
85 T = 1/4 #3 months maturity
86 paths = 1000 #1000 simulations
87 steps = 21*3 #91 days in 3 months
88 K = 50 #ATM put option stick price
89
90 stock = StockPrices(S0,r,sd,T,paths,steps) #simulate 10000 stock prices
91 S10 = stock[:,10] #take stock prices at day 10
92 P0 = f_BS(S0,K,T,sd,r,"put")
93 #compute Put option price today using black-scholes
94 P10 = f_BS(S10,K,T-10/252,sd,r,"put")
95 #compute Put option prices at day 10 for all simulations
96 Pc = np.quantile(P10,0.99)
97 #take the 0.99th option price at day 10 from the distribution
98 VaR_P = P0 - Pc #compute VaR
99
100 N_Poption = 100/VaR_P #number of options to buy
101 Cost_P = P0*N_Poption #total cost of all the options bought
102 Bond_P = 100-Cost_P #total bond to buy/short
103 w_option_P = Cost_P/(Cost_P+Bond_P) #weight on put option
104 w_bond_P = 1 - w_option_P #weight on bond
105
```

By doing so, we can get:

```
In [215]: w_option_P
Out[215]: -0.687336088490109

In [216]: w_bond_P
Out[216]: 1.687336088490109
```

This mean: short $0.6873 \times 100 = 68.73$ millions of put options

long $1.6873 \times 100 = 168.73$ millions of bonds at risk free rate

5. Now you can choose one of the stock, call, or put (with arbitrary strike) to combine with bonds. Plot the optimal portfolio position as well as your expected return for each strike. Which strike and portfolio do you choose? Explain the intuition behind this result.

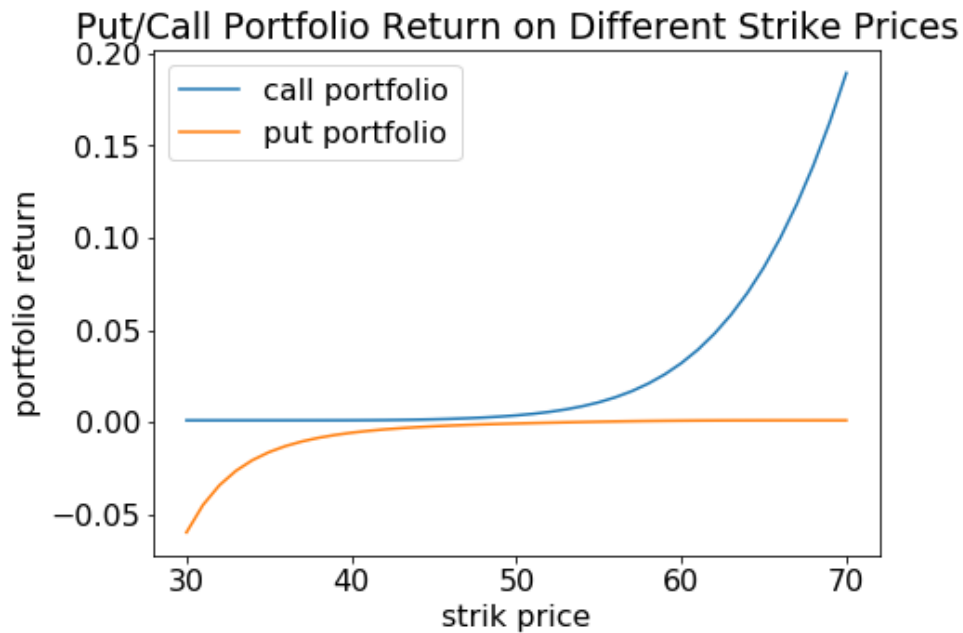
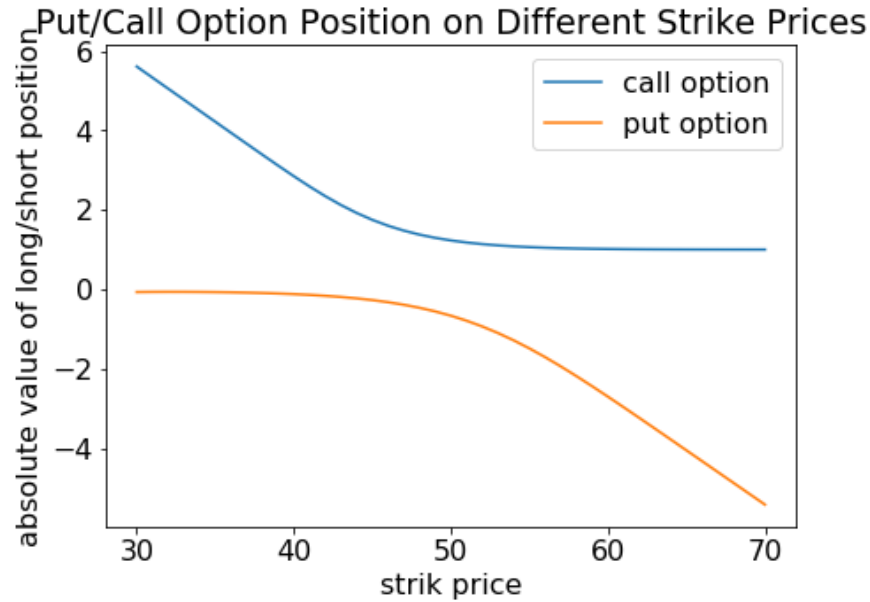
```
107 #####
108 #                                     3.Question 5                                #
109 #####
110 S0 = 50
111 r = 0.02
112 sd = 0.16
113 T = 1/4 #3 months maturity
114 paths = 1000 #1000 simulations
115 steps = 21*3 #91 days in 3 months
116 K = np.arange(30,71) #put option stick price range
117 n = len(K) #length of all put option stick prices
118
119 #call option-----
120 C0 = np.zeros(n)
121 for i in range(n):
122     C0[i] = f_BS(S0,K[i],T,sd,r,"call")
123 #compute Put option price today using black-scholes
124 stock = StockPrices(S0,r,sd,T,paths,steps) #simulate 10000 stock prices
125 S10 = stock[:,10] #take stock prices at day 10
126 C10 = np.zeros((len(S10),n))
127 Cc = np.zeros(n)
128 for i in range(n):
129     C10[:,i] = f_BS(S10,K[i],T-10/252,sd,r,"call")
130     #compute Put option price at day 10 for each simulation
131     Cc[i] = np.quantile(C10[:,i],0.01)
132     #take the 0.99th option price at day 10 from the distribution
133 VaR_C = C0 - Cc #compute VaR
134
135 N_Option = 100/VaR_C #number of options to buy
136 Cost_C = C0*N_Option #total cost of all the options bought
137 Bond_C = 100-Cost_C #total bond to buy/short
138 w_option_C = Cost_C/(Cost_C+Bond_C) #weight on put option
139 w_bond_C = 1 - w_option_C #weight on bond
```



```

142 #put option-----
143 P0 = np.zeros(n)
144 for i in range(n):
145     P0[i] = f_BS(S0,K[i],T,sd,r,"put")
146 #compute Put option price today using black-scholes
147 P10 = np.zeros((len(S10),n))
148 Pc = np.zeros(n)
149 for i in range(n):
150     P10[:,i] = f_BS(S10,K[i],T-10/252,sd,r,"put")
151     #compute Put option price at day 10 for each simulation
152     Pc[i] = np.quantile(P10[:,i],0.99)
153     #take the 0.99th option price at day 10 from the distribution
154 VaR_P = P0 - Pc #compute VaR
155
156 N_Poption = 100/VaR_P #number of options to buy
157 Cost_P = P0*N_Poption #total cost of all the options bought
158 Bond_P = 100-Cost_P #total bond to buy/short
159 w_option_P = Cost_P/(Cost_P+Bond_P) #weight on put option
160 w_bond_P = 1 - w_option_P #weight on bond
161
162 #plot the weights-----
163 plt.figure(2,figsize=(7,5))
164 plt.rcParams.update({'font.size': 16})
165 ax1 = plt.plot(K,w_option_C)
166 ax2 = plt.plot(K,w_option_P)
167 plt.legend(["call option","put option"])
168 plt.xlabel("striking price")
169 plt.ylabel("absolute value of long/short position")
170 plt.title("Put/Call Option Position on Different Strike Prices")
171
172 #return-----
173 C_R = np.zeros(n);P_R = np.zeros(n)
174 for i in range(n):
175     C_R[i] = w_option_C[i]*(np.mean(C10[:,i])-C0[i])/C0[i]+w_bond_C[i]*(r)*10/252
176     P_R[i] = w_option_P[i]*(np.mean(P10[:,i])-P0[i])/P0[i]+w_bond_P[i]*(r)*10/252
177 plt.figure(3,figsize=(7,5))
178 plt.rcParams.update({'font.size': 16})
179 ax1 = plt.plot(K,C_R)
180 ax2 = plt.plot(K,P_R)
181 plt.legend(["call portfolio","put portfolio"])
182 plt.xlabel("striking price")
183 plt.ylabel("portfolio return")
184 plt.title("Put/Call Portfolio Return on Different Strike Prices")
185

```



Based on the plots, we should construct portfolios that short the put options if the put option is ITM, and long the call options if the call option is ITM. If both the options are at the money, we can either short put options or long call options. Thus, if the strike is larger than 50, we should choose the portfolio that is shorting put; if the strike is smaller than 50, we should choose the portfolio that is longing the call.

6. Extra question. What happens if you have to respect a constraint on expected shortfall instead of value-at-risk?