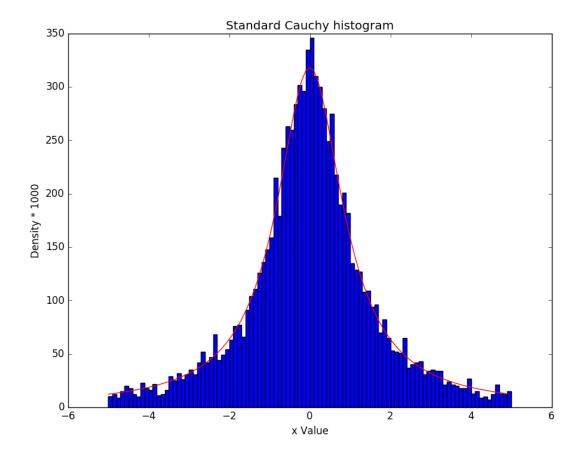
Question 1)b)

```
1. # importing the necessary packages
2. import numpy as np
3. from numpy import pi, tan
4. import matplotlib.pyplot as plt
5.
7. # CAUCHY
8.
9. # density function
10. def f(x):
11.
       return (1/pi) * 1/(1 + x**2)
12.
13. # standard cauchy distribution
14. X = np.random.standard cauchy(10000)
15.
16. # generalized inverse
17. # there is no built in function for cotangent.
18. # so I used the identity: cot(x) = 1/tan(x)
19. Y = -(1./tan(pi*X))
20.
21. # setting up variables to plot original density
22. Z = np.arange(-5, 5, 0.1)
23. W = f(Z)*1000
24.
25. # standard cauchy distribution plot
26. plt.plot(Z, W, color = 'red')
27. # histogram sample plot
28. plt.hist(Y, range = (-5, 5), bins = 100, color = 'blue')
29. plt.title ("Standard Cauchy histogram")
30. plt.xlabel ("x Value")
31. plt.ylabel ("Density * 1000")
32.
33. plt.show()
```

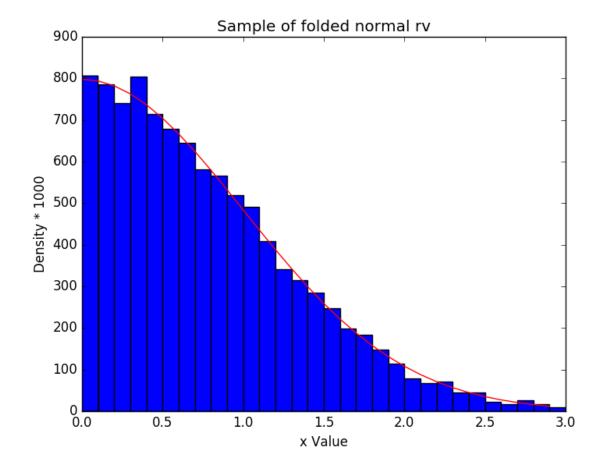
Figure:



Question 3)b)

```
1. # importing the necessary packages
2. import numpy as np
3. from numpy import pi, sqrt, exp
4. import matplotlib.pyplot as plt
7. # acceptance rejection method
8.
9. # original density function
10. def f(x):
       return 2./sqrt(2.*pi) * exp(-(x**2) / 2.)
12. # density function of dominant exponential distribution
13. def g(x):
14. return exp(-x)
15.
16. # array that will be filled with accepted variables
17. a = np.zeros(0)
18.
19. # constant c for method
20. c = 2./sqrt(2.*pi) * exp(0.5)
21.
22. n = 0
23.
24. plt.clf()
25.
26. # want 10000 samples
27. while n < 10000:
28. U1 = np.random.rand()
29.
       U2 = np.random.rand()
30. Y = -np.log(1 - U1)
31.
32. if c * g(Y) * U2 <= f(Y):
33.
          n += 1
34.
         # accept
35.
           a = np.append(a, Y)
36.
37.
38.
39. plt.hist(a, range = (0, 3), bins = 30, color = 'blue')
41. # setting up variables to plot original density
42. Z = np.arange(0, 3, 0.1)
43. W = f(Z)*1000
44.
45. # original distribution plot
46. plt.plot(Z, W, color = 'red')
47. plt.title ("Sample of folded normal rv")
48. plt.xlabel ("x Value")
49. plt.ylabel ("Density * 1000")
50.
51. plt.show()
```

Figure:



Question 5) Box Muller .py

```
1. # importing the necessary packages
import numpy as np
3. from numpy import pi, sin, cos, sqrt, exp, log
4. import matplotlib.pyplot as plt
5. import timeit
6.
8. # Box Muller method
9.
10. # starting timer
11. start = timeit.default timer()
12.
13. plt.clf()
14.
15. # exact normal density function
16. def f(x):
       return 1/(sqrt(2 * pi)) * exp(-(x)**2 / 2)
17.
18.
19. # initialize array of X rv's
20. X1 = np.zeros(0)
21. X2 = np.zeros(0)
22.
23. # defining Box Muller function producting std normal variables
24. def BoxMull(X1, X2):
25.
       n = 0
      while n < 50000:
26.
27.
28.
           U1 = np.random.rand()
29.
           U2 = np.random.rand()
30.
           X1 = np.append(X1, cos(2*pi*U1) * sqrt(-2*log(U2)))
31.
           X2 = np.append(X2, sin(2*pi*U1) * sqrt(-2*log(U2)))
32.
33.
34.
          n += 1
35.
36. return X1, X2
37.
38. # get function values
39. X1, X2 = BoxMull(X1, X2)
40.
41. # setting up variables to plot original density
42. Z = np.arange(-3, 3, 0.1)
43. W = f(Z)*5000
44.
45. # standard normal distribution plot
46. plt.plot(Z, W, color = 'red')
47.
48. # histogram sample plot
49. plt.hist([X1, X2], range = (-3, 3), bins = 60, alpha = 0.5)
50. plt.title ("Box-Muller method sampling")
51. plt.xlabel ("x Value")
52. plt.ylabel ("Density * 5000")
53.
54. plt.show()
55.
```

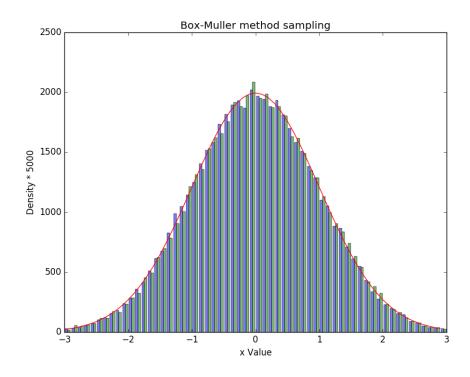
```
56. # stopping timer
57. stop = timeit.default_timer()
58.
59. print "Box Muller method runtime = " , stop - start, " seconds"
```

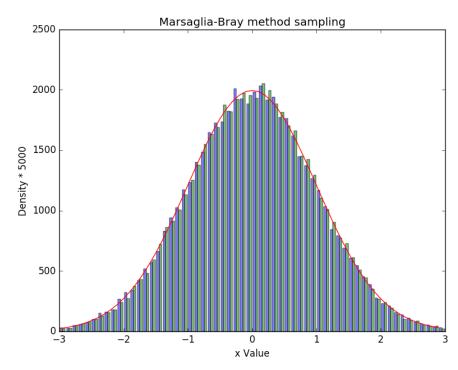
Marsaglia Bray .py

```
1. # importing the necessary packages
2. import numpy as np
3. from numpy import pi, sqrt, exp, log
4. import matplotlib.pyplot as plt
5. import timeit
9.
10. # starting timer
11. start = timeit.default_timer()
12.
13. plt.clf()
14.
15. # exact normal density function
16. def f(x):
       return 1/(sqrt(2 * pi)) * exp(-(x)**2 / 2)
17.
18.
19. # initialize array of accepted rv's
20. X1 = np.zeros(0)
21. X2 = np.zeros(0)
22.
23. # defining Marsaglia Bray function producing std normal variables
24. def MarsBray(X1, X2):
       n = 0
25.
       while n < 50000:
26.
27.
            # setting dummy initial value
28.
29.
30.
            # this while loop makes sure that we are using an accepted S
            while S > 1:
31.
32.
               V1 = 2 * np.random.rand() - 1
33.
               V2 = 2 * np.random.rand() - 1
34.
               S = (V1 ** 2) + (V2 ** 2)
35.
36.
            X1 = np.append(X1, V1 * sqrt(-2 * log(S) / S))
37.
            X2 = np.append(X2, V2 * sqrt(-2 * log(S) / S))
38.
39.
            n += 1
40.
41.
       return X1, X2
42.
43. # get function values
44. X1, X2 = MarsBray(X1, X2)
46. # setting up variables to plot original density
47. Z = np.arange(-3, 3, 0.1)
48. W = f(Z)*5000
49.
50. # standard normal distribution plot
51. plt.plot(Z, W, color = 'red')
52.
```

```
53. # histogram sample plot
54. plt.hist([X1, X2], range = (-3, 3), bins = 60, alpha = 0.5)
55. plt.title ("Marsaglia-Bray method sampling")
56. plt.xlabel ("x Value")
57. plt.ylabel ("Density * 5000")
58.
59. plt.show()
60.
61. # stopping timer
62. stop = timeit.default_timer()
63.
64. print "Marsaglia-Bray method runtime = " , stop - start, " seconds"
```

Figures:





Output of code:

Box Muller method runtime = 2.83222103119 seconds

Marsaglia-Bray method runtime = 2.75477385521 seconds

Question 6)

```
1. # importing the necessary packages
import numpy as np
3. from numpy import sqrt, exp, log
4. from scipy.stats import norm
5. import matplotlib.pyplot as plt
8. # question 6)a)
9.
10. # variables given
11. S 0 = 100.
12. B0 = 1
13. mu = 0.05
14. sigma = 0.2
15. r = 0.03
16.
17. # European Put, using Black Scholes formula
18. K = 110.
19. T = 0.5
21. d1 = (log(S 0/K) + (r + (sigma**2)/2)*T) / (sigma * sqrt(T))
22. d2 = d1 - sigma*sqrt(T)
23.
24. Price1 = K*exp(-r*T)*norm.cdf(-d2) - S 0*norm.cdf(-d1)
25.
26. print("European Put option price (using Black Scholes formula) is:")
27. print(round(Price1,2))
28.
30. # question 6)b)
31.
32. # defining the stock generation function
33. def generateStock(S_0, r, sigma, T):
      return S_0 * exp((r - (sigma**2)/2)*T + sigma * sqrt(T) * np.random.standard_normal
34.
 ())
35.
36. # defining payoff function, for this case it's regular European put
37. def payoff1(S, K):
38. return max(0, K - S)
39.
40. # setting number of simulation runs
41. runs = 10000
42. # initialize array that will have payoffs of option
43. payoffs = np.zeros(0)
44.
45. # looping through
46. for i in xrange(runs):
47.
      # generate future stock
48.
      S_T = generateStock(S_0, r, sigma, T)
49.
50.
      # append to the payoffs list whatever the payoff is
51.
       payoffs = np.append(payoffs,
52.
          payoff1(S T, K))
53.
54. Price2 = exp(-r*T) * sum(payoffs)/runs
55.
56. print("European Put option price (using Monte Carlo integration) is: ")
57. print(round(Price2,2))
```

```
60. # question 6)c)
61.
62. # defining monte carlo integration formula, to make graphing easier
63. def MonteCarlo1(runs):
       # initialize array that will have payoffs of option
       payoffs = np.zeros(0)
66.
67.
       # looping through
68.
       for i in xrange(runs):
69.
           # generate future stock
70.
           S_T = generateStock(S_0, r, sigma, T)
71.
72.
           # append to the payoffs list whatever the payoff is
73.
           payoffs = np.append(payoffs,
74.
                             payoff1(S_T, K))
75.
76.
       return exp(-r*T) * sum(payoffs)/runs
78. ##################################
79. plt.subplot(2, 2, 1)
80.
81. n = 0
82.
83. while n < 50:
84. y = MonteCarlo1(n)
       plt.plot(n, y, color = 'red', marker = '.')
86.
87.
88.
       n += 1
89.
90. plt.plot(n, y, color = 'red', marker = '.',
            label = "samples n = 50");
91.
92.
93. plt.title ("Approximation of European Put")
94. plt.xlabel ("Trials")
95. plt.ylabel ("Value")
96. plt.legend()
99. plt.subplot(2, 2, 2)
100.
101.
          n = 0
102.
103.
          while n < 100:
             y = MonteCarlo1(n)
104.
105.
              plt.plot(n, y, color = 'red', marker = '.')
106.
107.
108.
              n += 1
109.
          plt.plot(n, y, color = 'red', marker = '.',
110.
                  label = "samples n = 100");
111.
112.
          plt.title ("Approximation of European Put")
113.
114.
          plt.xlabel ("Trials")
          plt.ylabel ("Value")
115.
116.
          plt.legend()
117.
```

```
118.
119.
          plt.subplot(2, 2, 3)
120.
          n = 0
121.
122.
123.
          while n < 500:
124.
             y = MonteCarlo1(n)
125.
126.
              plt.plot(n, y, color = 'red', marker = '.')
127.
128.
              n += 1
129.
130.
          plt.plot(n, y, color = 'red', marker = '.',
131.
                  label = "samples n = 500");
132.
133.
          plt.title ("Approximation of European Put")
          plt.xlabel ("Trials")
134.
135.
          plt.ylabel ("Value")
136.
          plt.legend()
137.
138.
          139.
          plt.subplot(2, 2, 4)
140.
141.
          n = 0
142.
143.
          while n < 1000:
144.
              y = MonteCarlo1(n)
145.
              plt.plot(n, y, color = 'red', marker = '.')
146.
147.
148.
              n += 1
149.
150.
          plt.plot(n, y, color = 'red', marker = '.',
151.
                  label = "samples n = 1000");
152.
          plt.title ("Approximation of European Put")
153.
          plt.xlabel ("Trials")
154.
          plt.ylabel ("Value")
155.
156.
157.
          plt.legend()
158.
159.
          plt.show()
160.
          161.
162.
          # question 6)d)
163.
          # European asset-or-nothing digital call option
164.
165.
          # same maturity T = 0.5, same strike K = 110
166.
          # defining payoff function
167.
168.
          def payoff2(S, K):
              if S > K:
169.
170.
                  return S
171.
              else:
                 return 0
172.
173.
174.
          # setting number of simulation runs
175.
          runs = 10000
          # initialize array that will have payoffs of option
176.
177.
          payoffs = np.zeros(0)
```

```
178.
179.
          # looping through
180.
          for i in xrange(runs):
              # generate future stock
181.
182.
              S_T = generateStock(S_0, r, sigma, T)
183.
184.
              # append to the payoffs list whatever the payoff is
              payoffs = np.append(payoffs,
185.
186.
                                payoff2(S T, K))
187.
188.
          Price3 = exp(-r*T) * sum(payoffs)/runs
189.
190.
          print("European asset-or-
   nothing digital call option price (using Monte Carlo integration) is: ")
191.
          print(round(Price3,2))
192.
193.
          194.
          # question 6)e)
195.
196.
          # European cubic put option
197.
          # same maturity T = 0.5, same strike K = 110
198.
199.
          # defining payoff function
200.
          def payoff3(S, K):
201.
              return (max(0, K - S))**3
202.
203.
          # setting number of simulation runs
204.
          runs = 10000
205.
          # initialize array that will have payoffs of option
206.
          payoffs = np.zeros(0)
207.
          # looping through
208.
209.
          for i in xrange(runs):
210.
              # generate future stock
211.
              S T = generateStock(S 0, r, sigma, T)
212.
213.
              # append to the payoffs list whatever the payoff is
214.
              payoffs = np.append(payoffs,
215.
                                 payoff3(S T, K))
216.
          Price4 = exp(-r*T) * sum(payoffs)/runs
217.
218.
219.
          print("European cubic put option price (using Monte Carlo integration) is: ")
220.
          print(round(Price4,2))
221.
          222.
          # question 6)f)
223.
224.
          # European gap call option
225.
226.
          # same maturity T = 0.5, same strike K = 110
227.
          # L exercise level
228.
229.
          L = 105
230.
          # defining payoff function
231.
232.
          def payoff4(S, K):
233.
              if S > K:
234.
                 return max(0, S - L)
              else:
235.
236.
                 return 0
```

```
237.
238.
          # setting number of simulation runs
239.
          runs = 10000
          # initialize array that will have payoffs of option
240.
241.
          payoffs = np.zeros(0)
242.
          # looping through
243.
          for i in xrange(runs):
244.
245.
              # generate future stock
246.
              S_T = generateStock(S_0, r, sigma, T)
247.
248.
              # append to the payoffs list whatever the payoff is
              payoffs = np.append(payoffs,
249.
                                  payoff4(S T, K))
250.
251.
252.
          Price5 = exp(-r*T) * sum(payoffs)/runs
253.
254.
          print("European gap call option price (using Monte Carlo integration) is: ")
255.
          print(round(Price5,2))
256.
257.
          258.
          # question 6)g)
259.
          # European exponential put option
260.
          # same maturity T = 0.5, same strike K = 110
261.
262.
263.
          # defining payoff function
264.
          def payoff5(S, K):
265.
              return exp(max(0, K - S))
266.
267.
          # setting number of simulation runs
268.
          runs = 10000
269.
          # initialize array that will have payoffs of option
270.
          payoffs = np.zeros(0)
271.
272.
          # looping through
273.
          for i in xrange(runs):
274.
              # generate future stock
              S T = generateStock(S 0, r, sigma, T)
275.
276.
              # append to the payoffs list whatever the payoff is
277.
278.
              payoffs = np.append(payoffs,
279.
                                  payoff5(S_T, K))
280.
          Price6 = exp(-r*T) * sum(payoffs)/runs
281.
282.
          print("European exponential put option price (using Monte Carlo integration) is:
283.
284.
          print(round(Price6,2))
```

Output of code:

European Put option price (using Black Scholes formula) is:

10.97

European Put option price (using Monte Carlo integration) is:

11.03

European asset-or-nothing digital call option price (using Monte Carlo integration) is:

31.27 European cubic put option price (using Monte Carlo integration) is: 5809.78

European gap call option price (using Monte Carlo integration) is:

European exponential put option price (using Monte Carlo integration) is: 4.1053809922e+19

Figure:

