

contvar__py

March 20, 2025

0.1 Control Variate Sampling

```
[1]: import numpy as np
```

```
[2]: def f(x):  
    value = 1 / (1+x)  
    return value
```

```
[3]: def g(x):  
    value = 1 + x  
    return value
```

```
[4]: truth = 3.0 / 2.0
```

```
[5]: truth
```

```
[5]: 1.5
```

0.1.1 The Naive Solution

```
[21]: n = 15_000  
u = np.random.uniform(size=n)  
x1 = f(u)  
x1.mean()  
x1.var()  
x1.std()  
se = x1.std() / np.sqrt(n)
```

```
[31]: np.round(np.mean(x1), 4)
```

```
[31]: np.float64(0.6934)
```

```
[23]: se
```

```
[23]: np.float64(0.0011406080903478744)
```

0.1.2 The Control Variate Solution

```
[24]: c = 0.4773
      y = g(u)
      x2 = f(u) + c * (g(u) - truth)

[30]: np.round(np.mean(x2), 4)

[30]: np.float64(0.6932)

[26]: se2 = x2.std() / np.sqrt(n)

[19]: se2

[19]: np.float64(0.0002006231759532942)

[ ]:
```

0.2 Naive Monte Carlo in a BS World

```
[ ]: import time
      t1 = time.time()

[ ]: def VanillaCallPayoff(spot, strike):
      return np.maximum(spot - strike, 0.0)

[ ]: # The same old same old parameters

      S = 41.0
      K = 40.0
      r = 0.08
      v = 0.30
      q = 0.0
      T = 1.0
      M = 10000 # number of MC replications
      N = 252 # number of MC steps in a particular path

[ ]: dt = T
      nudt = (r - q - 0.5 * v * v) * dt
      sigdt = v * np.sqrt(dt)

[ ]: spot_t = np.empty((N))
      call_t = np.empty(M)

      z = np.random.normal(size=(M,N))

      for i in range(M):
          spot_t[0] = S
```

```

for j in range(1,N):
    spot_t[j] = S * np.exp(nudt + sigdt * z[i,j])
    call_t[i] = VanillaCallPayoff(spot_t[-1], K)

```

```

[ ]: call_prc = np.exp(-r * T) * call_t.mean()
[ ]: t2 = time.time()

```

```

[ ]: call_prc

```

```

[ ]: 6.9288468728032111

```

```

[ ]: se = call_t.std() / np.sqrt(M)

```

```

[ ]: se

```

```

[ ]: 0.10848490988044222

```

```

[ ]: print("The Naive Monte Carlo Price is: {0:.3f}".format(call_prc))
[ ]: print("The Naive Monte Carlo StdErr is: {0:.6f}".format(se))
[ ]: print("The total time take: {0}".format(t2-t1))

```

```

The Naive Monte Carlo Price is: 6.929
The Naive Monte Carlo StdErr is: 0.108485
The total time take: 4.806628227233887

```

```

[ ]:

```

0.2.1 The Control Variate Approach in a BS World

We will use the BS-Delta formula for our control variate. We can write the BS Delta function as follows:

```

[ ]: from scipy.stats import norm

```

```

[ ]: def BlackScholesDelta(spot, t, strike, expiry, volatility, rate, dividend):
[ ]:     tau = expiry - t
[ ]:     d1 = (np.log(spot/strike) + (rate - dividend + 0.5 * volatility *
[ ]:     ↪volatility) * tau) / (volatility * np.sqrt(tau))
[ ]:     delta = np.exp(-dividend * tau) * norm.cdf(d1)
[ ]:     return delta

```

```

[ ]: erddt = np.exp((r - q) * dt)
[ ]: beta = -1.0

[ ]: spot_t = np.empty((N))
[ ]: call_t = np.empty(M)
[ ]: #cash_flow_t = np.zeros((engine.replications, ))
[ ]: z = np.random.normal(size=(M,N))

```

```

for i in range(M):
    #spot_t = spot
    convar = 0.0
    #z = np.random.normal(size=int(engine.time_steps))
    spot_t[0] = S

    for j in range(1,N):
        t = i * dt
        delta = BlackScholesDelta(S, t, K, T, v, r, q)
        spot_t[j] = spot_t[j-1] * np.exp(nudt + sigdt * z[i,j])
        convar += delta * (spot_t[j] - spot_t[j-1]* erddt)
        #spot_t = spot_tn

    call_t[i] = VanillaCallPayoff(spot_t[-1], K) + beta * convar

```

```

/home/brough/anaconda3/lib/python3.5/site-packages/ipykernel/__main__.py:3:
RuntimeWarning: divide by zero encountered in double_scalars
    app.launch_new_instance()
/home/brough/anaconda3/lib/python3.5/site-packages/ipykernel/__main__.py:3:
RuntimeWarning: invalid value encountered in sqrt
    app.launch_new_instance()

```

```

[ ]: disc = np.exp(-r * T)
     call_prc = disc * call_t.mean()

```

```

[ ]: call_prc

```

```

[ ]: nan

```

```

[ ]:

```