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1 Problem 1

Following is the Value given by Black Scholes

```
In [4]: #Problem 1------
       S0 = 10
       sigma = 0.2
       r = 0.04
       dt = 0.002
       dX = sigma*np.array([np.sqrt(dt),np.sqrt(3*dt),np.sqrt(4*dt)])
       K = 10
       T = 0.5
       steps = int(T/dt)
       S = np.arange(4,17,1)
       def f_BS(S0,type):
          K=10.0
          T=0.5
          sigma=0.2
          r=0.04
          d_1=(np.log(S0/K)+(r+0.5*sigma**2)*T)/(sigma*np.sqrt(T))
          d_2=d_1-sigma*np.sqrt(T)
          if type=="call":
              option=(S0*si.norm.cdf(d_1,0.0,1.0)-K*np.exp(-r*T)*si.norm.cdf(d_2,0.0,1.0))
          if type=="put":
              option = (K*np.exp(-r*T)*si.norm.cdf(-d_2, 0.0, 1.0)-S0*si.norm.cdf(-d_1, 0.0, 1.0))
          return option
       BS\_EFD = np.zeros(13)
       for i in range(13):
```

BS_EFD[i] = f_BS(S[i],"put")

```
In [8]: print("Balck Scholes: ",BS_EFD,)
Balck Scholes: [5.80198673e+00 4.80198692e+00 3.80205782e+00 2.80535743e+00
 1.84426859e+00 1.02442812e+00 4.64694534e-01 1.71536863e-01
 5.24596231e-02 1.36511383e-02 3.10746681e-03 6.34561801e-04
 1.18785901e-041
def f_EFD(S0,dX,sigma,dt,r,K,steps):
           Pu = dt*(sigma**2/(2*dX**2)+(r-sigma**2/2)/(2*dX))
           Pm = 1-dt*sigma**2/(dX**2) - r*dt
           Pd = dt*(sigma**2/(2*dX**2)-(r-sigma**2/2)/(2*dX))
           TerP = np.arange(np.log(100)+dX,np.log(1)-dX,-dX)
           index = np.where(np.exp(TerP)<S0)[0][0]
           N = len(TerP)
           A = np.zeros(shape = (N,N))
           move = np.array([0,1,2])
           A[0,move] = [Pu,Pm,Pd]
           A[1,move] = [Pu,Pm,Pd]
           for i in range(1,(N-3)):
               A[i+1,move+i] = [Pu,Pm,Pd]
           A[N-2,move+N-3] = [Pu,Pm,Pd]
           A[N-1,move+N-3] = [Pu,Pm,Pd]
           F = np.maximum(K - np.exp(TerP),0)
           B = np.zeros(N)
           B[-1] = np.exp(TerP[-2])-np.exp(TerP[-1])
           for i in range(steps):
               F = np.dot(A,F)+B
           option = np.mean([F[index-1],F[index]])
           return option
       optionE = np.zeros((3,13))
       for i in range(3):
           for j in range(13):
               optionE[i,j] = f_EFD(S[j],dX[i],sigma,dt,r,K,steps)
       Err_EFD = BS_EFD - optionE
  Following is the option prices calculated by EXPLICIT METHOD with column represents
```

different S0 values and row represents different dX values:

```
In [12]: print(optionE)
[[5.78827438e+00 4.78252372e+00 3.79936205e+00 2.81673938e+00
  1.85171073e+00 1.04077220e+00 4.67558599e-01 1.66762471e-01
```

```
5.58697956e-02 1.45709312e-02 2.90395697e-03 6.80252673e-04
 1.04023936e-04]
[5.78462035e+00 4.81160099e+00 3.79211478e+00 2.78859638e+00
 1.89704928e+00 1.03172067e+00 4.57218418e-01 1.72958472e-01
 4.85085971e-02 1.30618534e-02 2.75245387e-03 6.58877253e-04
 1.33288592e-041
[5.77016098e+00 4.80476841e+00 3.82596355e+00 2.78566460e+00
 1.88464056e+00 1.01294878e+00 4.49636834e-01 1.76474418e-01
 5.25783055e-02 1.60492028e-02 2.72412706e-03 8.00402203e-04
 1.28975428e-04]]
-7.44214180e-03 -1.63440741e-02 -2.86406454e-03 4.77439259e-03
 -3.41017247e-03 -9.19792970e-04 2.03509834e-04 -4.56908719e-05
  1.47619650e-05]
[ 1.73663872e-02 -9.61406677e-03 9.94304568e-03 1.67610516e-02
 -5.27806852e-02 -7.29254735e-03 7.47611606e-03 -1.42160894e-03
  3.95102604e-03 5.89284820e-04 3.55012943e-04 -2.43154515e-05
 -1.45026905e-05]
[ 3.18257570e-02 -2.78148964e-03 -2.39057252e-02 1.96928364e-02
 -4.03719637e-02 1.14793477e-02 1.50577001e-02 -4.93755484e-03
 -1.18682373e-04 -2.39806455e-03 3.83339752e-04 -1.65840402e-04
 -1.01895264e-05]]
```

Following is the option prices ERRORS with EXPLICIT METHOD compared with Black Scholes with column represents different S0 values and row represents different dX values:

```
In [13]: print(Err_EFD)
-7.44214180e-03 -1.63440741e-02 -2.86406454e-03 4.77439259e-03
 -3.41017247e-03 -9.19792970e-04 2.03509834e-04 -4.56908719e-05
  1.47619650e-05]
 [ 1.73663872e-02 -9.61406677e-03 9.94304568e-03 1.67610516e-02
 -5.27806852e-02 -7.29254735e-03 7.47611606e-03 -1.42160894e-03
  3.95102604e-03 5.89284820e-04 3.55012943e-04 -2.43154515e-05
 -1.45026905e-05]
 [ 3.18257570e-02 -2.78148964e-03 -2.39057252e-02 1.96928364e-02
 -4.03719637e-02 1.14793477e-02 1.50577001e-02 -4.93755484e-03
 -1.18682373e-04 -2.39806455e-03 3.83339752e-04 -1.65840402e-04
 -1.01895264e-05]]
def f_IFD(S0,dX,sigma,dt,r,K,steps):
           Pu = -0.5*dt*(sigma**2/(dX**2)+(r-sigma**2/2)/(dX))
           Pm = 1+dt*sigma**2/(dX**2) + r*dt
           Pd = -0.5*dt*(sigma**2/(dX**2)-(r-sigma**2/2)/(dX))
```

```
TerP = np.arange(np.log(20)+dX,np.log(1)-dX,-dX)
    index = np.where(np.exp(TerP)<S0)[0][0]</pre>
    N = len(TerP)
    A = np.zeros(shape = (N,N))
    move = np.array([0,1,2])
    A[0,[0,1]] = [1,-1]
    A[1,move] = [Pu,Pm,Pd]
    for i in range(N-3):
        A[i+1,move+i] = [Pu,Pm,Pd]
    A[-2, [-3, -2, -1]] = [Pu, Pm, Pd]
    A[-1,[-2,-1]] = [1,-1]
    F = np.maximum(K - np.exp(TerP),0)
    B = np.zeros(N)
    B[1:-1] = F[1:-1]
    B[-1] = np.exp(TerP[-2])-np.exp(TerP[-1])
    for i in range(steps):
        F = np.dot(np.linalg.inv(A),B)
        B = np.zeros(N)
        B[1:-1] = F[1:-1]
        B[-1] = np.exp(TerP[-2])-np.exp(TerP[-1])
    option = np.mean([F[index-1],F[index]])
    return option
optionI = np.zeros((3,13))
for i in range(3):
    for j in range(13):
        optionI[i,j] = f_IFD(S[j],dX[i],sigma,dt,r,K,steps)
Err_IFD = BS_EFD - optionI
```

Following is the option prices calculated by IMPLICIT METHOD with column represents different S0 values and row represents different dX values:

```
5.21952606e-02 1.60484797e-02 2.79061044e-03 8.42000965e-04 1.43323902e-04]]
```

Following is the option prices ERRORS with IMPLICIT METHOD compared with Black Scholes with column represents different S0 values and row represents different dX values:

```
In [17]: print(Err_IFD)
[[ 1.58286438e-02 2.21135696e-02 5.85339343e-03 -7.88692764e-03
  -3.91320875e-03 -1.28824154e-02 -9.91545740e-05 6.14072864e-03
 -3.01478027e-03 -9.44354280e-04 1.31611936e-04 -8.64787335e-05
  1.46577856e-06]
 [ 2.42792216e-02 -1.02326877e-03 2.02737010e-02 2.85461830e-02
  -4.04847472e-02 3.82379965e-03 1.52227583e-02 2.53546515e-03
  5.19302609e-03 8.57195783e-04 -6.98778148e-04 -4.20321323e-05
 -2.45818214e-05]
 [ 3.39516641e-02 -1.42944369e-04 -2.07613364e-02 2.31960938e-02
 -3.68225278e-02 1.49839850e-02 1.78340819e-02 -3.45513839e-03
  2.64362551e-04 -2.39734147e-03 3.16856365e-04 -2.07439163e-04
 -2.45380008e-05]]
def f_CNFD(S0,dX,sigma,dt,r,K,steps):
            Pu = -1/4*dt*(sigma**2/(dX**2)+(r-sigma**2/2)/(dX))
            Pm = 1+dt*sigma**2/(2*dX**2) + r*dt/2
            Pd = -1/4*dt*(sigma**2/(dX**2)-(r-sigma**2/2)/(dX))
            TerP = np.arange(np.log(20)+dX,np.log(1)-dX,-dX)
            F = np.maximum(K - np.exp(TerP),0)
            index = np.where(np.exp(TerP)<S0)[0][0]
            N = len(TerP)
            move = np.array([0,1,2])
            X = np.zeros(shape = (N,N))
            X[0,[0,1]] = 0
            X[1,move] = [-Pu,-(Pm-2),-Pd]
            for i in range(1,(N-3)):
                X[i+1,move+i] = [-Pu,-(Pm-2),-Pd]
            X[-2,move+N-3] = [-Pu,-(Pm-2),-Pd]
            Y = np.zeros(N)
            Y[-1] = np.exp(TerP[-2])-np.exp(TerP[-1])
            A = np.zeros(shape = (N,N))
            A[0,[0,1]] = [1,-1]
            A[1,move] = [Pu,Pm,Pd]
            for i in range(1,(N-3)):
```

```
A[i+1,move+i] = [Pu,Pm,Pd]
A[-2,move+N-3] = [Pu,Pm,Pd]
A[-1,[-2,-1]] = [1,-1]

B = np.dot(np.linalg.inv(A),X)
D = np.dot(np.linalg.inv(A),Y)

for i in range(steps):
    F = np.dot(B,F)+D

    option = np.mean([F[index],F[index-1]])
    return option

optionCN = np.zeros((3,13))
for i in range(3):
    for j in range(13):
        optionCN[i,j] = f_CNFD(S[j],dX[i],sigma,dt,r,K,steps)

Err_CNFD = BS_EFD - optionCN
```

Following is the option prices calculated by CRANK-NICOLSON METHOD with column represents different S0 values and row represents different dX values:

```
In [20]: print(optionCN)

[[5.78615025e+00 4.77986543e+00 3.79618944e+00 2.81315496e+00 1.84800517e+00 1.03735518e+00 4.65072477e-01 1.65546083e-01 5.54490323e-02 1.45137172e-02 2.92332848e-03 6.96397803e-04 1.09897568e-04]

[5.77769967e+00 4.80300227e+00 3.78176861e+00 2.77671722e+00 1.88457606e+00 1.02065681e+00 4.49751356e-01 1.69154845e-01 4.72259555e-02 1.27133178e-02 3.74858670e-03 6.53347323e-04 1.35037185e-04]

[5.76802723e+00 4.80212194e+00 3.82280456e+00 2.78206810e+00 1.88091432e+00 1.00950193e+00 4.47139902e-01 1.75151977e-01 5.21634270e-02 1.59674806e-02 2.74066139e-03 8.15729714e-04 1.35100459e-04]]
```

Following is the option prices ERRORS with CRANK-NICOLSON METHOD compared with Black Scholes with column represents different S0 values and row represents different dX values:

```
In [22]: print(Err_CNFD)

[[ 1.58364851e-02     2.21214876e-02     5.86838001e-03 -7.79753206e-03
    -3.73657477e-03 -1.29270547e-02 -3.77942894e-04     5.99078022e-03
    -2.98940920e-03 -8.62578907e-04     1.84138327e-04 -6.18360017e-05
    8.88833323e-06]
```

2 Problem 2

```
In [23]: #Problem 2-----
S0 = 10
sigma = 0.2
r = 0.04
dt = 0.002
dS = 0.25
K = 10
T = 0.5
steps = int(T/dt)
```

Following is the function that calculates the option prices by using GENERALIZED FINITE-DIFFERENCE method

```
def f_A_PDE(a,dS,r,sigma,dt,S0,K,steps,name):
           TerP = np.arange(0,20+dS,dS)
           index = np.abs(TerP-S0).argmin()
           n = len(TerP)
           j = np.arange(1,n-1,1)
           alpha = a
           a1 = 0.5*((sigma**2)*(j**2)-r*j)*(1-alpha)
           a2 = -1/dt - ((sigma**2)*(j**2)+r)*(1-alpha)
           a3 = 0.5*((sigma**2)*(j**2)+r*j)*(1-alpha)
           b1 = 0.5*((sigma**2)*(j**2)-r*j)*alpha
           b2 = 1/dt - ((sigma**2)*(j**2)+r)*alpha
           b3 = 0.5*((sigma**2)*(j**2)+r*j)*alpha
           P = np.maximum(np.round(K - TerP,4),0)
           C = np.maximum(np.round(TerP - K,4),0)
           A = np.zeros(shape = (n,n))
           move = np.array([0,1,2])
           A[0,[0,1]] = [1,-1]
           for i in range(n-2):
               A[i+1,move+i] = [a1[i],a2[i],a3[i]]
```

```
A[-1,[-2,-1]] = [1,-1]
B = np.zeros(shape = (n,n))
for i in range(n-2):
    B[i+1,move+i] = [-b1[i],-b2[i],-b3[i]]
eC = np.zeros(n)
eC[0] = TerP[0] - TerP[1]
eP = np.zeros(n)
eP[0] = TerP[-2] - TerP[-1]
for i in range(steps):
    C = np.dot(np.linalg.inv(A),np.dot(B,C)+eC)
    C = np.maximum(C,np.maximum(np.round(TerP - K,4),0))
    P = np.dot(np.linalg.inv(A),np.dot(B,P)+eP)
    P = np.maximum(P,np.maximum(np.round(K - TerP,4),0))
if name == "C":
    return C[index]
else:
    return P[index]
```

The following generates the options prices for American Put/Call Option

```
In [25]: optionC = np.zeros((9,13))
         optionP = np.zeros((9,13))
         S = np.arange(4,17,1)
         dS = [0.25, 1, 1.25]
         a = [1,0,0.5]
         pS = [0,0,0,1,1,1,2,2,2]
         pa = [0,1,2,0,1,2,0,1,2]
         for i in range(9):
             for j in range(len(S)):
                 optionC[i,j] = f_A_PDE(a[pa[i]],dS[pS[i]],r,sigma,dt,S[j],K,steps,"C")
         for i in range(9):
             for j in range(len(S)):
                 optionP[i,j] = f_A_PDE(a[pa[i]],dS[pS[i]],r,sigma,dt,S[j],K,steps,"P")
In [26]: print(optionC)
[[3.23696901e-10 4.73061853e-07 8.95673120e-05 3.48932677e-03
  4.20739932e-02 2.21037817e-01 6.60797846e-01 1.36832742e+00
  2.25001819e+00 3.21149181e+00 4.20060288e+00 5.19630645e+00
  6.18995908e+00]
 [7.15081363e-10 6.94829322e-07 1.04309683e-04 3.65166541e-03
  4.24139964e-02 2.20923454e-01 6.60217199e-01 1.36799609e+00
  2.25004803e+00 3.21150804e+00 4.20007668e+00 5.19418020e+00
  6.18404856e+00]
```

```
[4.91185022e-10 5.77781039e-07 9.68547354e-05 3.57066850e-03
4.22441497e-02 2.20980311e-01 6.60507750e-01 1.36816144e+00
2.25003115e+00 3.21148329e+00 4.20024434e+00 5.19487489e+00
6.18595847e+00]
[7.92163922e-07 2.64510236e-05 5.06920716e-04 5.88940615e-03
4.29164858e-02 2.01044471e-01 6.23047544e-01 1.34812264e+00
2.24495750e+00 3.21142689e+00 4.20098941e+00 5.19616136e+00
6.18932658e+001
[8.98402156e-07 2.87121514e-05 5.31679021e-04 6.02573911e-03
4.32324386e-02 2.01054174e-01 6.22287807e-01 1.34785662e+00
2.24494162e+00 3.21103546e+00 4.19910596e+00 5.19019725e+00
6.17409466e+00]
[8.44309397e-07 2.75709614e-05 5.19263256e-04 5.95763041e-03
4.30747307e-02 2.01048932e-01 6.22667641e-01 1.34798610e+00
2.24492747e+00 3.21111016e+00 4.19952831e+00 5.19140032e+00
6.17673664e+001
[1.18170076e-06 5.71238451e-05 1.38619739e-03 1.83159179e-02
1.83159179e-02 1.37262478e-01 5.96307245e-01 1.55398273e+00
2.72281218e+00 2.72281218e+00 3.95284590e+00 5.19604680e+00
6.43617980e+001
[1.30190422e-06 6.06167536e-05 1.42978419e-03 1.85328723e-02
1.85328723e-02 1.37463642e-01 5.95522972e-01 1.55385564e+00
2.72254696e+00 2.72254696e+00 3.95088809e+00 5.18808734e+00
6.41198505e+001
[1.24099393e-06 5.88597004e-05 1.40796747e-03 1.84245399e-02
1.84245399e-02 1.37362970e-01 5.95913751e-01 1.55390480e+00
2.72256656e+00 2.72256656e+00 3.95121063e+00 5.18923776e+00
6.41480464e+00]]
```

dS = 1 S0=4 S0=5 S0=6 S0=7 S0=8 S0=9 S0=10 S0=11 S0=12 S0=13 S0=14 S0=15 S0=16 Crank- 8.4430**2**€7571**0**-0005**0920559570**£30**74**£30**74**£7010**49**62266**8**.347992.244933.211114.199535.1914 6.17674 Nicolson 07 05

In [27]: print(optionP)

[[6.0000000e+00 5.0000000e+00 4.0000000e+00 3.0000000e+00 2.00000000e+00 1.08128231e+00 4.80663846e-01 1.75062833e-01 5.31157712e-02 1.37847278e-02 3.14472884e-03 6.46276053e-04 1.22144542e-04] [6.00000000e+00 5.0000000e+00 4.0000000e+00 3.0000000e+00 2.00000000e+00 1.08057257e+00 4.79729532e-01 1.74591758e-01 5.31327553e-02 1.39374913e-02 3.25050207e-03 6.92775642e-04 1.38052364e-04] [6.00000000e+00 5.00000000e+00 4.00000000e+00 3.00000000e+00 2.00000000e+00 1.08092228e+00 4.80191555e-01 1.74823182e-01 5.31222962e-02 1.38607313e-02 3.19766197e-03 6.69528843e-04 1.30056887e-047 [6.0000000e+00 5.0000000e+00 4.0000000e+00 3.0000000e+00 2.00000000e+00 1.05162709e+00 4.39978944e-01 1.54452305e-01 4.81521757e-02 1.38759316e-02 3.79614482e-03 1.00389879e-03 2.59842593e-041 [6.00000000e+00 5.00000000e+00 4.00000000e+00 3.00000000e+00 2.00000000e+00 1.05125183e+00 4.39081179e-01 1.54173471e-01 4.82546415e-02 1.40218530e-02 3.88624541e-03 1.04610273e-03 2.76910435e-04] [6.00000000e+00 5.0000000e+00 4.0000000e+00 3.0000000e+00 2.00000000e+00 1.05143853e+00 4.39529996e-01 1.54312272e-01 4.82033850e-02 1.39490939e-02 3.84130557e-03 1.02501380e-03 2.68354863e-04] [6.25000000e+00 5.0000000e+00 3.75000000e+00 2.50000000e+00 2.50000000e+00 1.25337401e+00 4.15327729e-01 1.09703667e-01 2.56281147e-02 2.56281147e-02 5.59366486e-03 1.17666331e-03 2.43108776e-041 [6.25000000e+00 5.0000000e+00 3.75000000e+00 2.50000000e+00 2.50000000e+00 1.25331072e+00 4.14427377e-01 1.09625147e-01 2.57729041e-02 2.57729041e-02 5.69067855e-03 1.21736388e-03

```
2.57124201e-04]

[6.25000000e+00 5.00000000e+00 3.75000000e+00 2.50000000e+00

2.50000000e+00 1.25334221e+00 4.14877237e-01 1.09664129e-01

2.57006742e-02 2.57006742e-02 5.64228210e-03 1.19702451e-03

2.50097916e-04]]
```

dS =						
0.25	S0=4	S0=5	S0=6	S0=7	S0=8	S0=9 S0=10 S0=11 S0=12 S0=13 S0=14 S0=15 S0=16
Implicit	6	5	4	3	2	1.0812 8 .48066 9 .17506 9 .05311 58 01378 9 700314 973 0064 62376 01221
Explitcit	6	5	4	3	2	1.0805\bar{v}.479730.17459\bar{u}.05313\bar{u}\bar{w}1393\bar{v}\bar{w}03250\bar{w}0069\bar{u}\bar{w}\bar{w}01380
Crank-	6	5	4	3	2	1.0809 2 .48019 2 .17482 6 .0531223
Nicolson						
dS = 1	S0=4	S0=5	S0=6	S0=7	S0=8	S0=9 S0=10 S0=11 S0=12 S0=13 S0=14 S0=15 S0=16
Implicit	6	5	4	3	2	1.0516 3 .43997 9 .15445 2 .04815 22)1387 6 90379 6 1 2 0100 6 9002598
Explitcit	6	5	4	3	2	1.05125.439080.154176.04825060140209003886250104610002769
Crank-	6	5	4	3		2 1.051440.439530.15431 0 .04820 0 401394 9 100384 03 0010250
Nicolson						
10						
dS = 1.25	S0=4	S0=5	S0=6	S0=7	S0=8	S0=9 S0=10 S0=11 S0=12 S0=13 S0=14 S0=15 S0=16
Implicit	6.25	5	3.75	2.5	2.5	1.25337.415328.109704.02562810256281005599600117660002431
Explitcit	6.25	5	3.75	2.5	2.5	$1.2533 \\ 0.41442 \\ 0.10962 \\ 0.02577 \\ 0.902577 \\ 0.90569 \\ 0.80121 \\ 0.36002571 \\ 0.02577 \\ 0.90569 \\ 0.90569 \\ 0$
Crank- Nicolson	6.25	5	3.75	2.5	2.5	1.2533 \di .41487\vec{v}.10966\di.025700\tau25700\tau25700\tau2500564\div25119\vec{v}00\div202500

The followings are plots

```
In [28]: name = ["explicit method","implicit method","Crank-Nicolson"]
    Shape = ["s","v","o"]
    plt.figure(1,figsize = (8,6))
    for i in range(3):
        plt.plot(np.arange(4,17,1),optionC[i,:],marker = Shape[i],label = name[i])
    plt.legend()
    plt.title("American Call with dS = 0.25")
    plt.xlabel("SO")
    plt.ylabel("Option Prices")

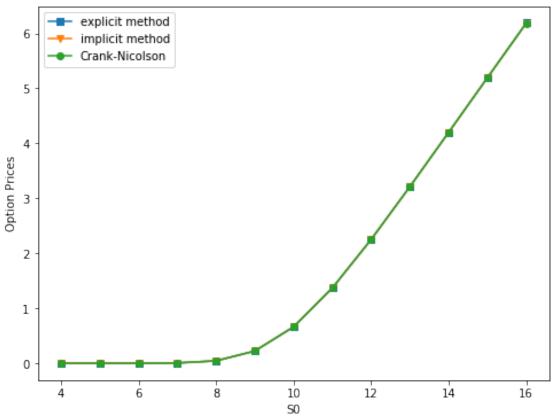
    plt.figure(2,figsize = (8,6))
    for i in range(3):
        plt.plot(np.arange(4,17,1),optionC[i+3,:],marker = Shape[i],label = name[i])
    plt.legend()
```

```
plt.title("American Call with dS = 1")
plt.xlabel("SO")
plt.ylabel("Option Prices")

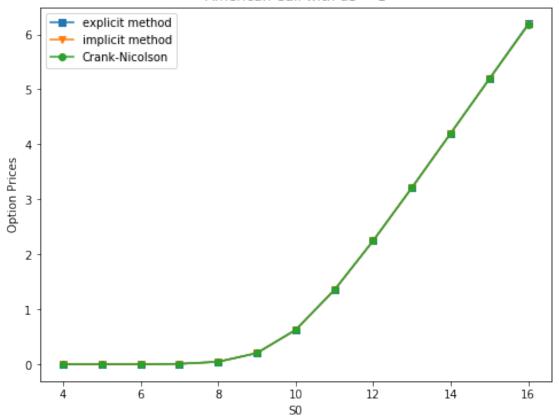
plt.figure(3,figsize = (8,6))
for i in range(3):
    plt.plot(np.arange(4,17,1),optionC[i+6,:],marker = Shape[i],label = name[i])
plt.legend()
plt.title("American Call with dS = 1.25")
plt.xlabel("SO")
plt.ylabel("Option Prices")
```

Out[28]: Text(0, 0.5, 'Option Prices')

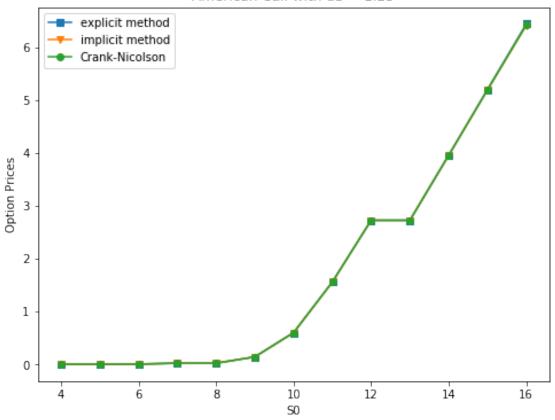
American Call with dS = 0.25



American Call with dS = 1



American Call with dS = 1.25

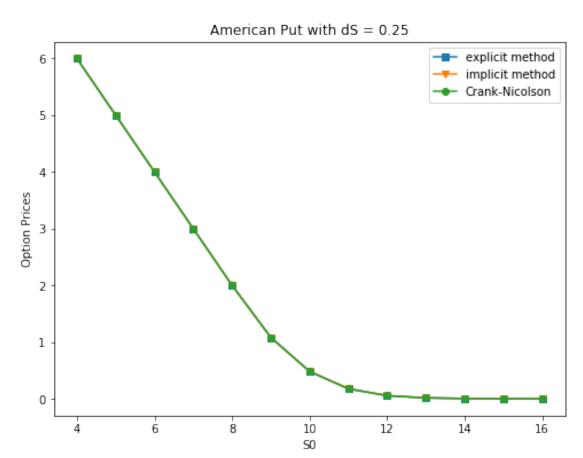


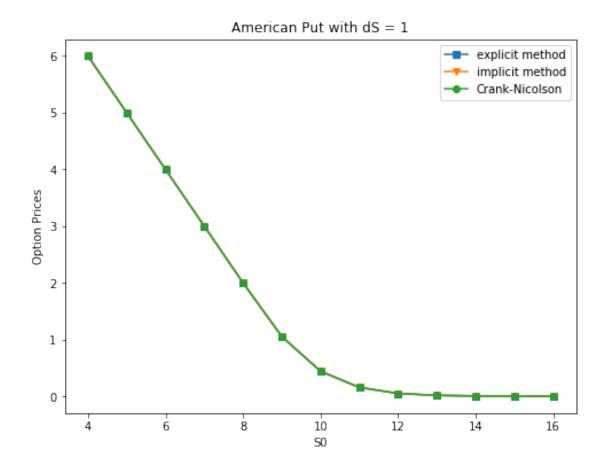
The plots show us that all the three methods are giving the very similar results since they are all overlaying on each other

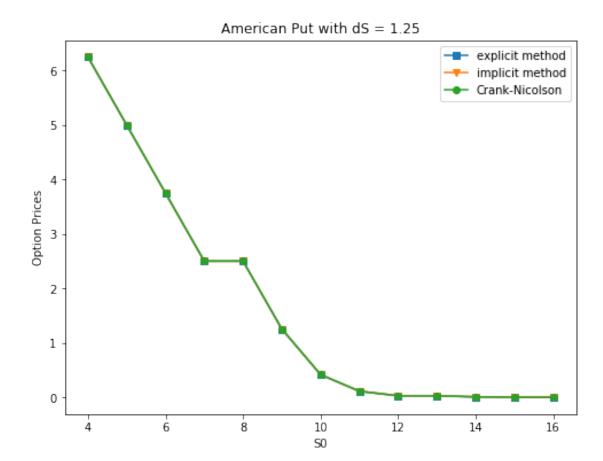
```
In [29]: plt.figure(4,figsize = (8,6))
         for i in range(3):
             plt.plot(np.arange(4,17,1),optionP[i,:],marker = Shape[i],label = name[i])
         plt.legend()
         plt.title("American Put with dS = 0.25")
         plt.xlabel("S0")
         plt.ylabel("Option Prices")
         plt.figure(5,figsize = (8,6))
         for i in range(3):
             plt.plot(np.arange(4,17,1),optionP[i+3,:],marker = Shape[i],label = name[i])
         plt.legend()
         plt.title("American Put with dS = 1")
         plt.xlabel("S0")
         plt.ylabel("Option Prices")
         plt.figure(6,figsize = (8,6))
         for i in range(3):
```

```
plt.plot(np.arange(4,17,1),optionP[i+6,:],marker = Shape[i],label = name[i])
plt.legend()
plt.title("American Put with dS = 1.25")
plt.xlabel("SO")
plt.ylabel("Option Prices")
```

Out[29]: Text(0, 0.5, 'Option Prices')







The plots show us that all the three methods are also giving the very similar results since they are all overlaying on each other