Question 2

For the part 1, we implement the constant time stepping function with fixed timesteps and nodes.

Const_stepping.m

```
function [result] = const timestep(S,N)
alpha parameter =0.8;
                         % constant related to alpha
r = .02;
                         % risk free rate
T = 1;
                         % time to expiry
K = 10;
                         % strike price
S0 = 10;
                         % initial stock price
delt = T/N;
                         % time interval
Large = 1e6;
                        % penalty coefficient
tolerance = 1/Large;
                        % toleranceeranceerance term
Vs = max(K-S.^2, S.^2 - K)';
m = length(S); %number of grids in row
alpha central = zeros(m,1);
beta central = zeros(m,1);
alpha forward = zeros(m, 1);
beta forward = zeros(m, 1);
alpha backward = zeros(m, 1);
beta backward = zeros(m,1);
alpha = zeros(m, 1);
beta = zeros(m,1);
for i = 2: m - 1
    %%central alpha and beta formula
    alpha central(i) = alpha parameter^2S(i)/((S(i) - S(i-1))*(S(i+1) - S(i-1))
1)))...
        -r*S(i)/(S(i+1) - S(i-1));
    beta central(i) = alpha parameter^2*S(i)/((S(i+1) - S(i))*(S(i+1) - S(i-1)))...
        +r*S(i)/(S(i+1) - S(i-1));
    %%forward alpha and beta formula
    alpha forward(i) = alpha parameter^2*S(i)/((S(i) - S(i-1))*(S(i+1) - S(i-1)));
    beta_forward(i) = alpha_parameter^2*S(i)/((S(i+1) - S(i))*(S(i+1) - S(i-1)))...
        + r*S(i)/(S(i+1) - \overline{S}(i)); % same as beta central
    %%backward alpha and beta formula
    alpha backward(i) = alpha parameter^2S(i) / ((S(i) - S(i-1))*(S(i+1) - S(i-1))*
        - r * S(i) / (S(i+1) - S(i));
    beta backward(i) = alpha parameter^2S(i)/((S(i+1) - S(i))*(S(i+1) - S(i-1)));
%% choosing parameter
for i = 2:m-1
   if(alpha central(i) >=0 && beta central(i) >=0)
       alpha(i) = alpha central(i);
       beta(i) = beta_central(i);
   elseif (alpha_forward(i) >=0 && beta_forward(i) >=0)
       alpha(i) = alpha forward(i);
       beta(i) = beta_forward(i);
   else
       alpha(i) = alpha backward(i);
```

```
beta(i) = beta_backward(i);
   end
end
%% CN-Rannacher time stepping
V 3 = zeros(m, N+1);
V_{init_3} = max(K - S.^2, S.^2 - K)';
V 3(:,1) = V init 3;
V_old_3 = V_init_3;
V_{\text{new}}3 = V_{\text{old}}3;
V \text{ new3n} = [];
for i = 1:2
    M matrix = [delt.*-alpha, delt.*(alpha + beta + r), delt.*-beta];
    M = spdiags(M matrix, [-1,0,1], m-1, m);
    M = full([M; zeros(1,m)]);
    I = eye(m);
    t = 1;
    while t > tolerance
        pv = Large *(V_new_3 < Vs);</pre>
        PV = diag(pv);
        rhs1 = V old 3 + PV * Vs; % RHS of the equation
        AP = sparse(spdiags(ones(m,1),0,m,m)+M+PV);
        [L3,U3,P3,Q3] = lu(AP);
        V_{new3n} = Q3 * ((L3*U3) \setminus (P3* rhs1)); % compute (Vn+1) (k+1)
        t = max(abs(V_new3n - V_new_3)./(max(ones(m,1), abs(V_new3n))));
        V_new_3 = V_new3n;
    end
    V old 3 = V new 3;
end
for i = 3:N
    M matrix = [delt.*-alpha/2, delt.*(alpha + beta + r)/2, delt.*-beta/2];
    M = spdiags(M_matrix, [-1,0,1], m-1, m);
    M = full([M; zeros(1, m)]);
    I = eye(m);
    t = 1;
    while t > tolerance
        pv = Large * (V_new_3 < Vs);
        PV = diag(pv);
        rhs2 = (I - M) * V old 3 + PV*Vs; % RHS of the equation
        AP2 = sparse(spdiags(ones(m,1),0,m,m)+M+PV);
        [L4,U4,P4,Q4] = lu(AP2);
        V_{new3n} = Q4 * ((L4*U4) \setminus (P4* rhs2)); % compute (Vn+1) (k+1)
        t = max(abs(V_new3n - V_new_3)./(max(1, abs(V_new3n))));
         % compute relative change
         V_new_3 = V_new3n;
    end
    V_old_3 = V_new_3;
X1 = sprintf('The option value for fully implict in N = %d, S= %d,
is: %s',N,length(S),V_new_3(S == S0));
disp(X1)
result = V new 3(S == S0);
end
```

The following method is used to get the table for different nodes and timesteps

```
K = 10;
N = 25;
S = [0:0.1*K:0.4*K,... %input S value
    0.45*K:0.05*K:0.8*K,...
    0.82*K:0.02*K:0.9*K,...
    0.91*K:0.01*K:1.1*K,...
    1.12*K:0.02*K:1.2*K,...
    1.25*K:.05*K:1.6*K,...
    1.7*K:0.1*K:2*K,...
    2.2*K, 2.4*K, 2.8*K,...
    3.6*K, 5*K, 7.5*K, 10*K];
% for node 62 and timestep 25
L = length(S);
V = const_timestep(S,N);
% for node 123 and timestep 50
N1 = 50;
S1 = movmean(S, 2);
S1 = [S, S1];
S1 = sort(S1);
S1 = S1(2:end);
L1 = length(S1);
V1 = const_timestep(S1,N1);
% for node 245 and timestep 100
N2 = 100;
S2 = movmean(S1, 2);
S2 = [S1, S2];
S2 = sort(S2);
S2 = S2(2:end);
L2 = length(S2);
V2 = const_timestep(S2,N2);
% for node 489 and timestep 200
N3 = 200;
S3 = movmean(S2, 2);
S3 = [S2,S3];
S3 = sort(S3);
S3 = S3(2:end);
L3 = length(S3);
V3 = const_timestep(S3,N3);
% for node 977 and timestep 400
N4 = 400;
S4 = movmean(S3, 2);
S4 = [S3, S4];
S4 = sort(S4);
S4 = S4(2:end);
L4 = length(S4);
V4 = const timestep(S4, N4);
T 1 =table([N;N1;N2;N3;N4],...
             [L;L1;L2;L3;L4],...
             [V;V1;V2;V3;V4],...
             [NaN; V1-V; V2-V1; V3-V2; V4-V3], ...
             [\,\text{NaN;NaN;}\,\,(\,\text{V1-V}\,)\,\,/\,\,(\,\text{V2-V1}\,)\,\,;\,\,(\,\text{V2-V1}\,)\,\,/\,\,(\,\text{V3-V2}\,)\,\,;\,\,(\,\text{V3-V2}\,)\,\,/\,\,(\,\text{V4-V3}\,)\,\,]\,\,)\,\,;
T_1.Properties.VariableNames ={'Timesteps','Node','Value','Change','Ratio'};
```

Result convergence table is shown below:

Timesteps	Node	Value	Change	Ratio
25	62	98.687	NaN	NaN
50	123	98.685	-0.0021605	NaN
100	245	98.684	-0.00059598	3.6252
200	489	98.684	-0.00013814	4.3144
400	977	98.684	-3.4116e-05	4.049

Table1: The option value in different timesteps and Node implemented by const delt method

From the table, we can see that the convergence of constant timestepping is around 4 and means quadratic convergence although there is a bit over fluctuate

```
figure(1);
plot(S_CN_R,V_CN_R);
xlabel('stock price')
ylabel('option value')
title('option value vs stock price in const timestep CN-R')
figure(2);
S_CN_R_2 = S_CN_R(2:end);
plot(S_CN_R_2,delta)
title('delta vs stock price in const timestep CN-R')
xlabel('stock price')
ylabel('delta')
```

Plot of option value vs stock prices:

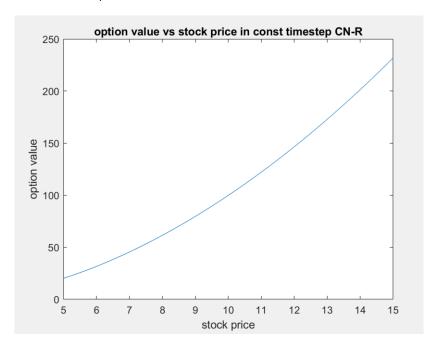


Figure 1. The option value vs tock price in const timestep CN-R

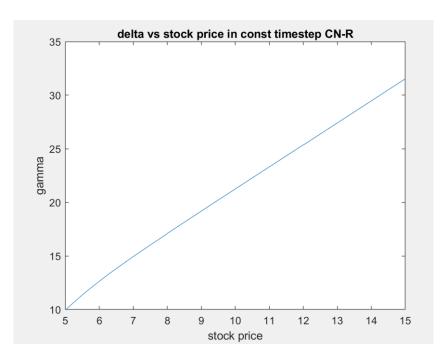


Figure 2. The delta vs tock price in const timestep CN-R

From the these two plots, we can see that the option price vs stock price plot is a little bit quadratic while the delta is not exactly linear, actually it is a little bit convex. Furthermore, I don't see any no oscillation.

Delt_selector.m

```
function [result,iteration] = delt_select(S,N)
alpha_parameter =0.8; % constant related to alpha
r = .02;
                         % risk free rate
T = 1;
                         % time to expiry
K = 10;
                         % strike price
S0 = 10;
                         % initial stock price
delt = T/N;
                         % time interval
Large = 1e6;
                         % penalty coefficient
tolerance = 1/Large;
                        % toleranceerance term
dnorm = 0.1;
Vs = max(K-S.^2, S.^2 - K)';
m = length(S); %number of grids in row
alpha_central = zeros(m,1);
beta_central = zeros(m,1);
alpha forward = zeros(m,1);
beta_forward = zeros(m,1);
alpha backward = zeros(m,1);
beta backward = zeros(m,1);
alpha = zeros(m,1);
beta = zeros(m, 1);
```

```
for i = 2: m - 1
         %%central alpha and beta formula
         alpha central(i) = alpha parameter^2S(i)/((S(i) - S(i-1))*(S(i+1) - S(i-1))
1)))...
                 -r*S(i)/(S(i+1) - S(i-1));
        beta central(i) = alpha parameter^2 S(i) / ((S(i+1) - S(i)) * (S(i+1) - S(i-1))) ...
                 +r*S(i)/(S(i+1) - S(i-1));
         \mbox{\ensuremath{\$}}\mbox{forward} alpha and beta formula
         alpha \ forward(i) \ = \ alpha \ parameter^2*S(i)/((S(i) \ - \ S(i-1))*(S(i+1) \ - \ S(i-1)));
         beta\_forward(i) = alpha\_parameter^2*S(i)/((S(i+1) - S(i))*(S(i+1) - S(i-1)))...
                 + r*S(i)/(S(i+1) - S(i)); % same as beta central
         %%backward alpha and beta formula
         alpha \ backward(i) = alpha \ parameter^2 + S(i) \ / \ ((S(i) - S(i-1)) + (S(i+1) - 
1))) ...
- r * S(i) / (S(i+1) - S(i));
         beta backward(i) = alpha parameter^2*S(i)/((S(i+1) - S(i))*(S(i+1) - S(i-1)));
end
%% choosing parameter
for i = 2:m-1
       if(alpha central(i) >=0 && beta central(i) >=0)
               alpha(i) = alpha_central(i);
               beta(i) = beta_central(i);
       elseif (alpha_forward(i) >=0 && beta forward(i) >=0)
               alpha(i) = alpha_forward(i);
               beta(i) = beta_forward(i);
      else
               alpha(i) = alpha_backward(i);
               beta(i) = beta backward(i);
      end
end
%% time step initialization
delt sum = delt;
delt old = delt;
%% CN-Rannacher time stepping
V 3 = zeros(m, N+1);
V init 3 = max(K - S.^2, S.^2 - K)';
V_3(:,1) = V_{init_3;}
V old 3 = V init 3;
V new 3 = V old \overline{3};
V = v = [];
%% for the first two implicit methods
for i = 1:2
         M matrix = [delt old.*-alpha, delt old.*(alpha + beta + r), delt old.*-beta];
        \overline{M} = \text{spdiags}(M \text{ matrix}, [-1, 0, 1], m-\overline{1}, m);
         M = full([M; zeros(1, m)]);
         I = eye(m);
         t = 1;
         while t > tolerance
                 pv = Large *(V_new_3 < Vs);</pre>
                 PV = diag(pv);
                 rhs1 = V_old_3 + PV * Vs; % RHS of the equation
                 AP = sparse(spdiags(ones(m, 1), 0, m, m) + M + PV);
                  [L3,U3,P3,Q3] = lu(AP);
                 V \text{ new3n} = Q3 * ((L3*U3) \setminus (P3* \text{ rhs1})); % compute (Vn+1) (k+1)
                 t = max(abs(V_new3n - V_new_3)./(max(ones(m,1), abs(V_new3n))));
                 V_new_3 = V_new3n;
```

```
end
    MaxRelChange = max(abs(V new 3 -
V_old_3)./(max(max(1,abs(V_new_3)),abs(V_old_3))));
    delt_new = (dnorm/MaxRelChange) *delt_old;
    delt sum = delt sum + delt new;
    delt old = delt new;
    V_old_3 = V_new_3;
end
iteration = 2;
%% for the rest, use CN-R method
while delt sum < T
    M matrix = [delt old.*-alpha/2, delt old.*(alpha + beta + r)/2, delt old.*-
beta/21:
    M = \text{spdiags}(M \text{ matrix}, [-1, 0, 1], m-1, m);
    M = full([M; zeros(1, m)]);
    I = eye(m);
    t = 1;
    while t > tolerance
        pv = Large * (V_new_3 < Vs);
        PV = diag(pv);
        rhs2 = (I - M) * V_old_3 + PV*Vs; % RHS of the equation
        AP2 = sparse(spdiags(ones(m, 1), 0, m, m) + M + PV);
        %AP2 = sparse(I + 0.5*M + PV);
        [L4,U4,P4,Q4] = lu(AP2);
V new3n = Q4 * ((L4*U4)\(P4* rhs2));% compute (Vn+1)(k+1)
        t = max(abs(V_new3n - V_new_3)./(max(1, abs(V_new3n))));
         % compute relative change
         V \text{ new } 3 = V \text{ new3n};
    end
    MaxRelChange = max(abs(V new 3 -
V_old_3)./(max(max(1,abs(V_new_3)),abs(V_old_3))));
    delt new = (dnorm/MaxRelChange) *delt old;
    delt_sum = delt_sum + delt_new;
    delt_old = delt_new;
    i = i + 1;
    %sprintf("the value of delt sum is %d and the price value is : %d ",delt sum,
V_new_3(S == S0))
    V_old_3 = V_new_3;
    iteration = iteration + 1;
end
 %% determine the last delt value
delt_old = T-(delt_sum - delt_new); %% last time step
if delt old >0
    M_matrix = [delt_old.*-alpha/2, delt_old.*(alpha + beta + r)/2, delt_old.*-
    M = spdiags(M matrix, [-1,0,1], m-1, m);
    M = full([M; zeros(1, m)]);
    I = eye(m);
    t = 1
    while t > tolerance
  pv = Large *(V_new_3 < Vs);</pre>
        PV = diag(pv);
        rhs2 = (I - M) * V old 3 + PV*Vs; % RHS of the equation
        AP2 = sparse(spdiags(ones(m, 1), 0, m, m) +M+PV);
        %AP2 = sparse(I + 0.5*M + PV);
        [L4,U4,P4,Q4] = lu(AP2);
```

```
V_{new3n} = Q4 * ((L4*U4) \setminus (P4* rhs2)); % compute (Vn+1)(k+1)
        t = max(abs(V new3n - V new 3)./(max(1, abs(V new3n))));
         % compute relative change
         V_new_3 = V_new3n;
    V_old_3 = V_new 3;
iteration = iteration + 1;
X1 = sprintf('The option value for fully implict in N = %d, S= %d,
is: %d', N, length(S), V_new_3(S == S0));
disp(X1)
result = V_new_3(S == S0);
The following is used to calculate the table
K = 10;
N = 25;
S = [0:0.1*K:0.4*K,... %input S value
    0.45*K:0.05*K:0.8*K,...
    0.82*K:0.02*K:0.9*K,...
    0.91*K:0.01*K:1.1*K,...
    1.12*K:0.02*K:1.2*K,...
    1.25*K:.05*K:1.6*K,...
    1.7*K:0.1*K:2*K,...
    2.2*K, 2.4*K, 2.8*K,..
    3.6*K, 5*K, 7.5*K, 10*K];
% for node 62
L = length(S);
[V_S,NS] = delt_select(S,N);
% for node 123
N1 = 50;
S1 = movmean(S, 2);
S1 = [S, S1];
S1 = sort(S1);
S1 = S1(2:end);
L1 = length(S1);
[V_S1,NS1] = delt_select(S1,N1);
% for node 245
N2 = 100;
S2 = movmean(S1, 2);
S2 = [S1, S2];
S2 = sort(S2);
S2 = S2(2:end);
L2 = length(S2);
[V S2,NS2] = delt select(S2,N2);
% for node 489 and timestep 200
N3 = 200;
S3 = movmean(S2, 2);
S3 = [S2, S3];
S3 = sort(S3);
S3 = S3(2:end);
L3 = length(S3);
[V_S3,NS3] = delt_select(S3,N3);
% for node 977 and timestep 400
N4 = 400;
S4 = movmean(S3, 2);
S4 = [S3, S4];
S4 = sort(S4);
S4 = S4(2:end);
L4 = length(S4);
[V_S4,NS4] = delt_select(S4,N4);
T S1 =table([NS;NS1;NS2;NS3;NS4],...
```

```
[L;L1;L2;L3;L4],...
[V_S;V_S1;V_S2;V_S3;V_S4],...
[NaN;V_S1-V_S;V_S2-V_S1;V_S3-V_S2;V_S4-V_S3],...
[NaN;NaN;(V_S1-V_S)/(V_S2-V_S1);(V_S2-V_S1)/(V_S3-V_S2);(V_S3-V_S2)/(V_S4-V_S3)]);
T S1.Properties.VariableNames = { 'Timesteps', 'Node', 'Value', 'Change', 'Ratio'};
```

Timesteps	Node	Value	Change	Ratio
19	62	98.687	NaN	NaN
21	123	98.685	-0.002006	NaN
26	245	98.684	-0.00055854	3.5916
30	489	98.684	-0.00014145	3.9485
32	977	98.684	-3.2703e-05	4.3254

Table 2: The option value in different timesteps and Node implemented by penalty method

From the table, we can see that the convergence of selected delt is around 4 and means quadratic convergence. And the option value do converge to 98.684. Compared to the constant timestepping, selected delt implemented by the penalty method is way more efficient, which can be seen from the timesteps. And these two gives almost exactly same result.

```
S CN R = S(S >= 5 \& S <= 15)';
V = V = V = V = 3 (S > 5 \& S < 15);
n = \overline{length(S CN R)};
delta = diff(V CN R)./diff(S CN R);
-((V_CN_R(2:n-1) - V_CN_R(1:n-2)) ./ (S_CN_R(2:n-1) - S_CN_R(1:n-2)))) ...
   ./ ((S_{CN_R(3:n)} - S_{CN_R(1:n-2))/2);
figure(1);
plot(S CN R, V CN R);
xlabel('stock price')
ylabel('option value')
figure(2);
S CN R 2 = S CN R(2:end);
plot(S CN R 2, delta)
xlabel('stock price')
ylabel('delta')
```

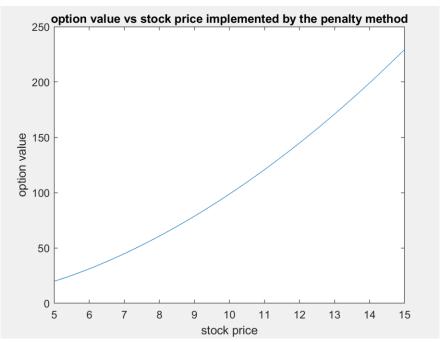


Figure 3. The option value vs tock price in penalty method CN-R

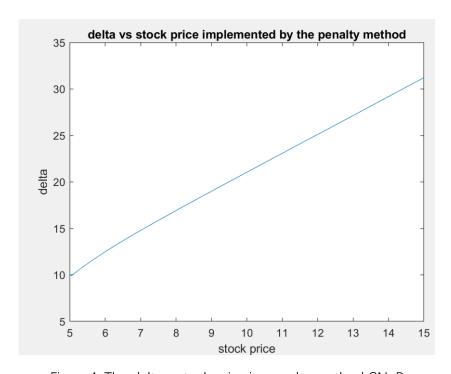


Figure 4. The delta vs tock price in penalty method CN-R

From the these two plots, we can see that the option price vs stock price plot is a little bit quadratic while the delta is not exactly linear, actually it is a little bit convex. Furthermore, I don't see any no oscillation. Compared to the two plots implemented by the const timestepping, I don't see a hugh difference between each other.