
Assignment 2 - Dupire formula and calibration of interest rate models

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In this project we will use the Dupire formula in two different versions (the first one using the derivatives of C and the second one with implied volatilities) for calibrating the volatility σ of a call option having observations for different strike prices K and maturity times T.

In the second part, we will proceed to calibrate the interest rate from STIBOR and swap rates using the Vasicek model for some market data. This calibration will be performed using Nelder-Mead simplex method (fminsearch) and BFGS quasi-Newton (fminunc).

Part 1

```
clc
clear
close all

load('LocVol.mat');

%
% Analytical solution. First of all, we will plot the analytical
% solution
% for the calibration problem in order to compare it with the results
% that
% we will obtain throughout the project. The plot is shown at the end
% of
% this part, together with the numerical results obtained.
%

figure(4)
[X,Y]=meshgrid(K,T);
SigmaAnalyt = 0.15+0.15*(0.5+2.*Y).*((X./100-1.2).^2)./(((X.^2)./(
(100.^2))+1.44));
surf(X,Y,SigmaAnalyt);
zlim([0 0.35])
xlabel("K")
ylabel("T")
zlabel("sigma")
title("Analytical solution")
caxis([0.00 0.3])
```

```
colormap(jet)
shading interp
colorbar

%
% Numerical solution for interior points. Here we calibrate the
% volatility for the interior points of the given data. We do that
% because
% of the fact that we will encounter some problems when adding
% boundary
% points, as we will see later on.
%

Tint = T(2:length(T)-1);
Kint = K(2:length(K)-1);
for i=2:size(C,1)-1
    for j=2:size(C,2)-1
        dCdTint(i-1,j-1) = (C(i+1,j)-C(i-1,j))/(T(i+1)-T(i-1));
        dCdKint(i-1,j-1) = (C(i,j+1)-C(i,j-1))/(K(j+1)-K(j-1));
    end
end

deltaK = 1;
for i=2:size(C,1)-1
    for j=2:size(C,2)-1
        dCdK2int(i-1,j-1) = (C(i,j+1)-2*C(i,j)+C(i,j-1))/
(deltaK)^2;
    end
end

for i=1:size(dCdKint,1)
    for j=1:size(dCdKint,2)
        SigmaNumInt(i,j) =
sqrt(2*(dCdTint(i,j)+(r-0)*Kint(j)*dCdKint(i,j)+0*C(i,j))/
(Kint(j)*Kint(j)*dCdK2int(i,j)));
    end
end

figure(1)
[XnumInt,YnumInt]=meshgrid(Kint,Tint);
surf(XnumInt,YnumInt,SigmaNumInt);
zlim([0 0.35])
xlabel("K")
ylabel("T")
zlabel("sigma")
title("Numerical solution for interior points")

caxis([0.0 0.3])
colormap(jet)
shading interp
colorbar

%
% Numerical solution. Now we proceed to compute the value of
```

```
%  $\sigma$  for the whole region (including the boundary points).
    However,
% as can be seen in the plot at the end of the section, we found out
    the
% problem that the values corresponding to the boundary of T=0.5
    somehow
% diverge towards negative values (in fact, a couple of points before
% computing the square root were negative, and an absolute value that
% should not be there was added in order to be able to at least obtain
    some
% results for the rest of the points).
%

for i=1:size(C,1) %i for T
    for j=1:size(C,2) %j for K
        if i==1 && j==1 %left bottom corner
            dCdT(i,j) = (-C(i+2,j)+4*C(i+1,j)-3*C(i,j))/(2*(T(i+1)-
T(i))); %taylor expansion fw with 2 order error
            dCdK(i,j) = (-C(i,j+2)+4*C(i,j+1)-3*C(i,j))/(2*(K(j+1)-
K(j)));
        elseif i==1 && j<size(C,2) && j>1 %bottom boundary
            dCdT(i,j) = (-C(i+2,j)+4*C(i+1,j)-3*C(i,j))/(2*(T(i+1)-
T(i)));
            dCdK(i,j) = (C(i,j+1)-C(i,j-1))/(K(j+1)-K(j-1));
        elseif i==1 && j==size(C,2) %right bottom corner
            dCdT(i,j) = (-C(i+2,j)+4*C(i+1,j)-3*C(i,j))/(2*(T(i+1)-
T(i)));
            dCdK(i,j) = (3*C(i,j)-4*C(i,j-1)+C(i,j-2))/(2*(K(j)-
K(j-1)));
        elseif i>1 && i<size(C,1) && j==1 %left side boundary
            dCdT(i,j) = (C(i+1,j)-C(i-1,j))/(T(i+1)-T(i-1));
            dCdK(i,j) = (-C(i,j+2)+4*C(i,j+1)-3*C(i,j))/(2*(K(j+1)-
K(j)));
        elseif i>1 && i<size(C,1) && j==size(C,2) %right side boundary
            dCdT(i,j) = (C(i+1,j)-C(i-1,j))/(T(i+1)-T(i-1));
            dCdK(i,j) = (3*C(i,j)-4*C(i,j-1)+C(i,j-2))/(2*(K(j)-
K(j-1)));
        elseif i==size(C,1) && j==1 %left up corner
            dCdT(i,j) = (3*C(i,j)-4*C(i-1,j)+C(i-2,j))/(2*(T(i)-
T(i-1)));
            dCdK(i,j) = (-C(i,j+2)+4*C(i,j+1)-3*C(i,j))/(2*(K(j+1)-
K(j)));
        elseif i==size(C,1) && j>1 && j<size(C,2) %upper boundary
            dCdT(i,j) = (3*C(i,j)-4*C(i-1,j)+C(i-2,j))/(2*(T(i)-
T(i-1)));
            dCdK(i,j) = (C(i,j+1)-C(i,j-1))/(K(j+1)-K(j-1));
        elseif i==size(C,1) && j==size(C,2) %right up corner
            dCdT(i,j) = (3*C(i,j)-4*C(i-1,j)+C(i-2,j))/(2*(T(i)-
T(i-1)));
            dCdK(i,j) = (-3*C(i,j)-4*C(i,j-1)+C(i,j-2))/(2*(K(j)-
K(j-1)));
        else
            dCdT(i,j) = (C(i+1,j)-C(i-1,j))/(T(i+1)-T(i-1));
            dCdK(i,j) = (C(i,j+1)-C(i,j-1))/(K(j+1)-K(j-1));
```

```
        end
    end
end

deltaK = 1;
for i=1:size(C,1)
    for j=1:size(C,2)
        if j==1
            dCdK2(i,j) = (2*C(i,j)-5*C(i,j+1)+4*C(i,j+2)-C(i,j+3))/
((deltaK)^2);
        elseif j==size(C,2)
            dCdK2(i,j) = (2*C(i,j)-5*C(i,j-1)+4*C(i,j-2)-C(i,j-3))/
((deltaK)^2);
        else
            dCdK2(i,j) = (C(i,j+1)-2*C(i,j)+C(i,j-1))/((deltaK)^2);
        end
    end
end

for i=1:size(dCdK,1)
    for j=1:size(dCdK,2)
        SigmaNum(i,j) =
sqrt(abs(2*(dCdT(i,j)+(r-0)*K(j)*dCdK(i,j)+0*C(i,j))/
(K(j)*K(j)*dCdK2(i,j)))));
    end
end

figure(2)
[Xnum1,Ynum1]=meshgrid(K,T);
surf(Xnum1,Ynum1,SigmaNum)
zlim([0 0.35])
xlabel("K")
ylabel("T")
zlabel("sigma")
title("Numerical solution")

caxis([0.0 0.3]);
colormap(jet)
shading interp
colorbar

%
% Numerical solution with implied volatilities. As we have mentioned
% before, the obtained volatilities were quite innacurate for boundary
% points and for larger values of trike price K, due to the fact
% that the values of C are so small there. So, the version of the
% Dupier
% formula with implied volatilities (that we will compute using the
% program
% that we developed in the previous assignment) will is used in this
% part.
% The resulting plot is shown at the end of the part 1 section.
%
```

```
for i=1:size(C,1)
    for j=1:size(C,2)
        Timp = T(i);
        t = 0;

        myfun = @(sigmaImp,Cimp,Kimp,S0)
            normcdf(1/(sigmaImp*sqrt((Timp-t)))*(log(S0/
            Kimp)+(r-0+0.5*sigmaImp*sigmaImp)*(Timp-t)))*S0*exp(-0*(Timp-
            t))-normcdf(1/(sigmaImp*sqrt((Timp-t)))*(log(S0/
            Kimp)+(r-0-0.5*sigmaImp*sigmaImp)*(Timp-t)))*Kimp*exp(-r*(Timp-t))-
            Cimp;

        Kimp = K(j);
        Cimp = C(i,j);

        fun = @(sigmaImp) myfun(sigmaImp,Cimp,Kimp,S0);

        sigmaImp(i,j) = fzero(fun,0);
        d1(i,j) = 1/(sigmaImp(i,j)*sqrt(Timp))*(log(S0/
            Kimp)+(r-0+0.5*sigmaImp(i,j)*sigmaImp(i,j))*Timp);
        d2(i,j) = 1/(sigmaImp(i,j)*sqrt(Timp))*(log(S0/
            Kimp)+(r-0-0.5*sigmaImp(i,j)*sigmaImp(i,j))*Timp);
    end
end

for i=1:size(sigmaImp,1)
    for j=1:size(sigmaImp,2)
        if i==1 && j==1
            dsdT(i,j) = (sigmaImp(i+1,j)-sigmaImp(i,j))/(T(i+1)-T(i));
            dsdK(i,j) = (sigmaImp(i,j+1)-sigmaImp(i,j))/(K(j+1)-K(j));
        elseif i==1 && j<size(sigmaImp,2) && j>1
            dsdT(i,j) = (sigmaImp(i+1,j)-sigmaImp(i,j))/(T(i+1)-T(i));
            dsdK(i,j) = (sigmaImp(i,j+1)-sigmaImp(i,j-1))/(K(j+1)-
            K(j-1));
        elseif i==1 && j==size(sigmaImp,2)
            dsdT(i,j) = (sigmaImp(i+1,j)-sigmaImp(i,j))/(T(i+1)-T(i));
            dsdK(i,j) = (sigmaImp(i,j)-sigmaImp(i,j-1))/(K(j)-K(j-1));
        elseif i>1 && i<size(sigmaImp,1) && j==1
            dsdT(i,j) = (sigmaImp(i+1,j)-sigmaImp(i-1,j))/(T(i+1)-
            T(i-1));
            dsdK(i,j) = (sigmaImp(i,j+1)-sigmaImp(i,j))/(K(j+1)-K(j));
        elseif i>1 && i<size(sigmaImp,1) && j==size(sigmaImp,2)
            dsdT(i,j) = (sigmaImp(i+1,j)-sigmaImp(i-1,j))/(T(i+1)-
            T(i-1));
            dsdK(i,j) = (sigmaImp(i,j)-sigmaImp(i,j-1))/(K(j)-K(j-1));
        elseif i==size(sigmaImp,1) && j==1
            dsdT(i,j) = (sigmaImp(i,j)-sigmaImp(i-1,j))/(T(i)-T(i-1));
            dsdK(i,j) = (sigmaImp(i,j+1)-sigmaImp(i,j))/(K(j+1)-K(j));
        elseif i==size(sigmaImp,1) && j>1 && j<size(sigmaImp,2)
            dsdT(i,j) = (sigmaImp(i,j)-sigmaImp(i-1,j))/(T(i)-T(i-1));
            dsdK(i,j) = (sigmaImp(i,j+1)-sigmaImp(i,j-1))/(K(j+1)-
            K(j-1));
        elseif i==size(sigmaImp,1) && j==size(sigmaImp,2)
            dsdT(i,j) = (sigmaImp(i,j)-sigmaImp(i-1,j))/(T(i)-T(i-1));
            dsdK(i,j) = (sigmaImp(i,j)-sigmaImp(i,j-1))/(K(j)-K(j-1));
```

```
        else
            dsdT(i,j) = (sigmaImp(i+1,j)-sigmaImp(i-1,j))/(T(i+1)-
T(i-1));
            dsdK(i,j) = (sigmaImp(i,j+1)-sigmaImp(i,j-1))/(K(j+1)-
K(j-1));
        end
    end
end

deltaK = 1;
for i=1:size(sigmaImp,1)
    for j=1:size(sigmaImp,2)
        if j==1
            dsdK2(i,j) = (2*sigmaImp(i,j)-5*sigmaImp(i,j
+1)+4*sigmaImp(i,j+2)-sigmaImp(i,j+3))/(deltaK)^2;
        elseif j==size(C,2)
            dsdK2(i,j) =
            (2*sigmaImp(i,j)-5*sigmaImp(i,j-1)+4*sigmaImp(i,j-2)-
sigmaImp(i,j-3))/(deltaK)^2;
        else
            dsdK2(i,j) = (sigmaImp(i,j
+1)-2*sigmaImp(i,j)+sigmaImp(i,j-1))/(deltaK)^2;
        end
    end
end

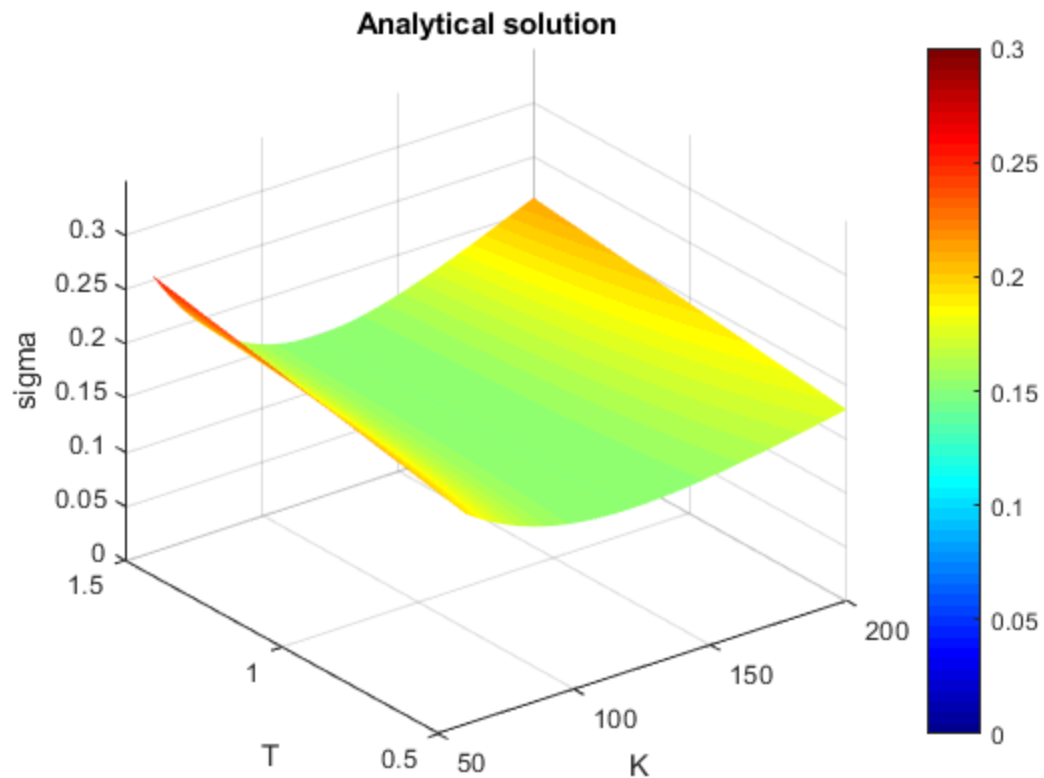
for i=1:size(dsdK,1)
    for j=1:size(dsdK,2)
        SigmaNumImp(i,j) =
        sqrt((sigmaImp(i,j)^2+2*sigmaImp(i,j)*T(i)*(dsdT(i,j)+r*K(j)*dsdK(i,j)))/
(1+2*d1(i,j)*K(j)*sqrt(T(i))*dsdK(i,j)+K(j)*K(j)*T(i)*(d1(i,j)*d2(i,j)*dsdK(i,j)*d
        end
    end

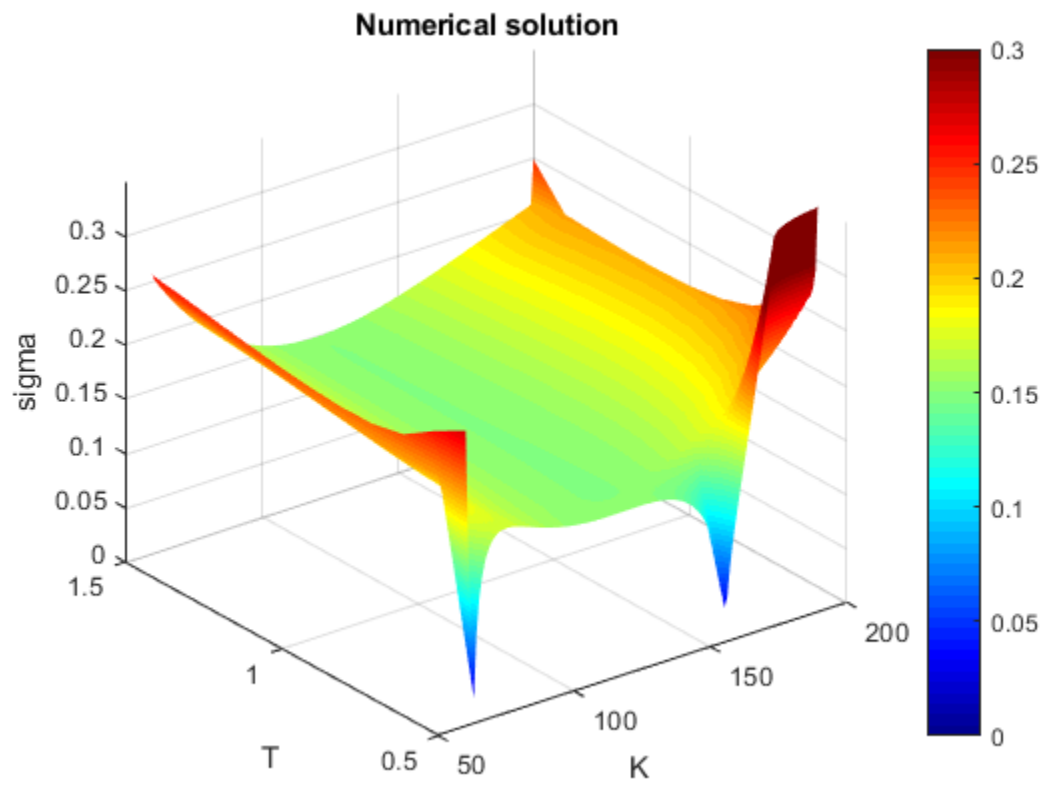
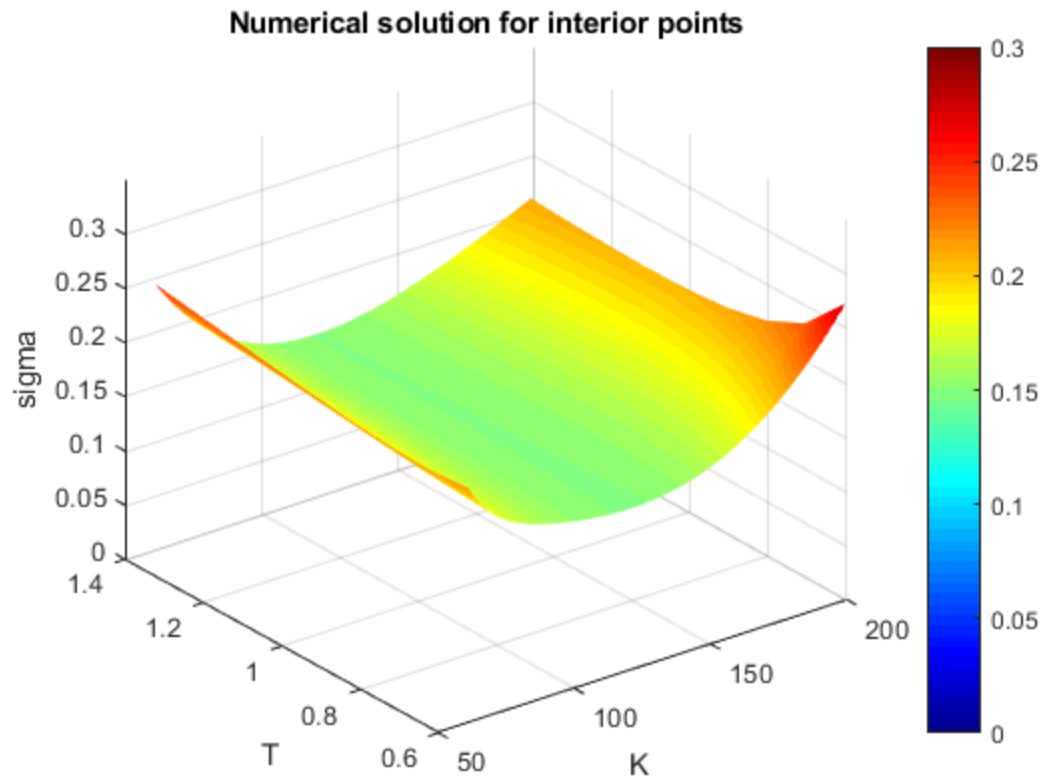
figure(3)
[Xnum2,Ynum2]=meshgrid(K,T);
surf(Xnum2,Ynum2,SigmaNumImp);
zlim([0 0.35])
xlabel("K")
ylabel("T")
zlabel("sigma")
title("Numerical solution with implied volatilities")

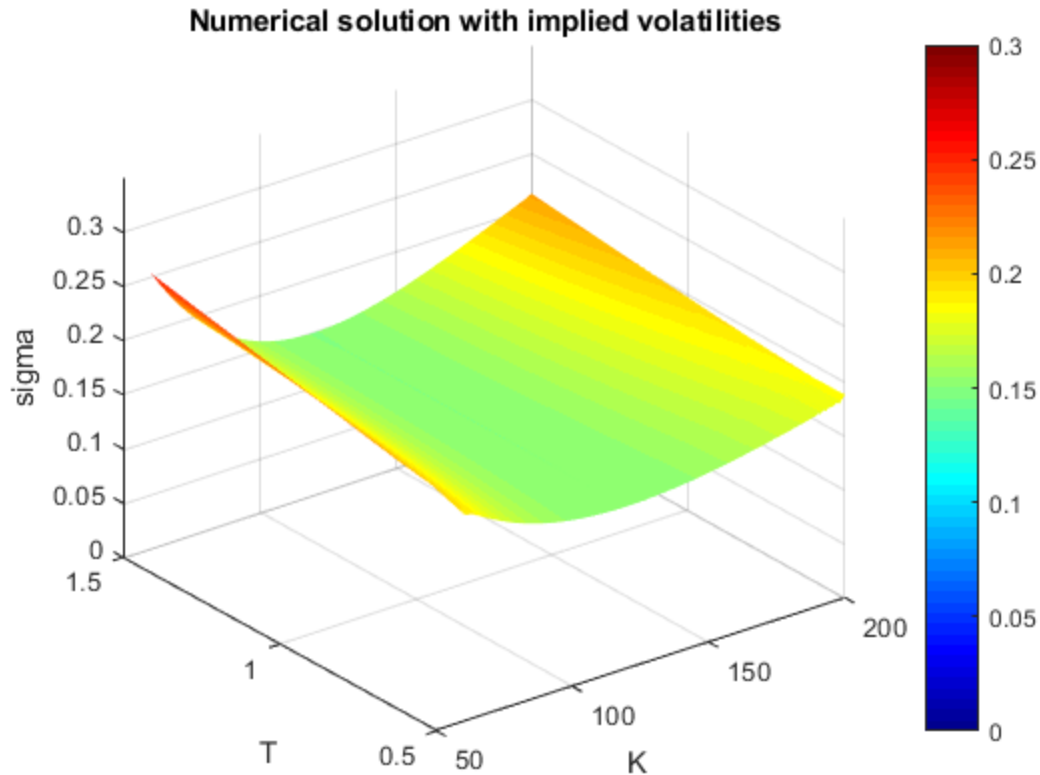
caxis([0.00 0.3])
colormap(jet)
shading interp
colorbar

%
% As it can be seen from the following plots, we can consider all of
% our
% results to be quite accurate for the interior points, where they are
% really close to the analytical solution (in fact the calibration
% using
```

```
% implied volatilities gave accurate results for all points (including  
% those with large strike K). However, as mentioned before, the second  
plot  
% shows some weird results that should not be correct at all. We do not  
know  
% the reason of them because we have checked the computed derivatives  
and  
% they don't give any strange results, so we think that it might be a  
% problem with the data or with the formula that makes the numerical  
errors  
% to explode around those points.  
%
```







Part 2

In this part, we will calibrate the simply-compounded interest rates and swap rates using the Vasicek model.

In this part we calibrate the parameters $\theta = \{k, \phi, \sigma, r_0\}$ by calling the functions `fminsearch` and `fminunc`.

```
clear all;
```

```
x0 = [0.9572 0.4854 0.8003 0.1419];  
%x0 = [1 1 1 1];  
T1 = [1 7 30 60 90 180]; %in days  
T1 = T1/360; % in years  
Lmarket = [0.0300 0.0302 0.0307 0.0314 0.0320 0.0335];  
  
T2 = [1 2 3 4 5 6 7 8 9 10]; %in years swap maturities time  
Smarket = [0.0351 0.0368 0.0375 0.0379 0.0381 0.0383 0.0384 0.0385  
0.0385 0.0386];  
swaptime = 0.25;  
fun = @(x)cirmodel(x,T1,T2,Lmarket,Smarket,swaptime);  
x1 = fminsearch(fun,x0); %from fminsearch  
x2 = fminunc(fun, x0); %from fminunc  
  
for i = 1:1:size(T1,2)  
    tau(i) = T1(i) - 0;  
    B(i) = (1 - exp(-x1(2)* tau(i)))/x1(2);
```

```
A(i) = (x1(1) - (x1(3)*x1(3)/(2*x1(2)*x1(2)))) * (B(i) - tau(i)) -  
(x1(3)*x1(3)*B(i)*B(i)/(4*x1(2)));  
Z(i) = exp(A(i) - B(i)*x1(4));  
L(i) = (1 - Z(i))/(tau(i)*Z(i));  
tau2(i) = T1(i) - 0;  
B2(i) = (1 - exp(-x2(2)* tau(i)))/x2(2);  
A2(i) = (x2(1) - (x2(3)*x2(3)/(2*x2(2)*x2(2)))) * (B(i) - tau(i))  
- (x2(3)*x2(3)*B(i)*B(i)/(4*x2(2)));  
Z2(i) = exp(A(i) - B(i)*x2(4));  
L2(i) = (1 - Z(i))/(tau(i)*Z(i));  
end  
  
for j = 1:size(T2,2)  
    tau(j) = T2(j) - 0;  
    B(j) = (1 - exp(-x1(2)* tau(j)))/x1(2);  
    A(j) = (x1(1) - (x1(3)*x1(3)/(2*x1(2)*x1(2)))) * (B(j) - tau(j)) -  
(x1(3)*x1(3)*B(j)*B(j)/(4*x1(2)));  
    Z(j) = exp(A(j) - B(j)*x1(4));  
    n = 1;  
    for k = swaptime:swaptime:T2(j)  
        Bj(n) = (1 - exp(-x1(2)*k))/x1(2);  
        Aj(n) = (x1(1) - (x1(3)*x1(3)/(2*x1(2)*x1(2)))) * (Bj(n) - k)  
- (x1(3)*x1(3)*Bj(n)*Bj(n)/(4*x1(2)));  
        Zj(n) = exp(Aj(n) - Bj(n)*x1(4));  
        n = n + 1;  
    end  
  
    S(j) = (1 - Z(j))/(swaptime * sum(Zj));  
  
    tau2(j) = T2(j) - 0;  
    B2(j) = (1 - exp(-x2(2)* tau(j)))/x2(2);  
    A2(j) = (x2(1) - (x2(3)*x2(3)/(2*x2(2)*x2(2)))) * (B(j) - tau(j))  
- (x2(3)*x2(3)*B(j)*B(j)/(4*x2(2)));  
    Z2(j) = exp(A(j) - B(j)*x2(4));  
    n = 1;  
  
    for k = swaptime:swaptime:T2(j)  
        Bj2(n) = (1 - exp(-x2(2)*k))/x2(2);  
        Aj2(n) = (x2(1) - (x2(3)*x2(3)/(2*x2(2)*x2(2)))) * (Bj(n) - k)  
- (x2(3)*x2(3)*Bj(n)*Bj(n)/(4*x2(2)));  
        Zj2(n) = exp(Aj(n) - Bj(n)*x2(4));  
        n = n + 1;  
    end  
  
    S2(j) = (1 - Z(j))/(swaptime * sum(Zj));  
end  
  
my_fig = figure('position', [0, 0, 700, 500]);  
  
subplot(2,1,1);  
plot(T1*360,L*100);  
hold on  
plot(T1*360,L2*100,'g:');  
plot(T1*360, Lmarket*100,'rd','MarkerSize',10);
```

```
xlim([0 T1(end)*360]);
ylim([2 5]);
xlabel('time(days)');
ylabel('yield(%)');
legend('Interest rate(fminsearch)', 'Interest rate(fminunc)', 'Market
data')
title('Yield curve for the simply-compounded interest rates');
hold off

subplot(2,1,2);
plot(T2,S*100);
hold on
plot(T2,S2*100,'g:');
plot(T2, Smarket*100,'rd','MarkerSize',10);
xlim([0 T2(end)]);
ylim([2 5]);
xlabel('time(year)');
ylabel('yield(%)');
legend('Swap rate(fminsearch)', 'Swap rate(fminunc)', 'Market data')
title('Yield curve for the swap rates');
hold off

%saveas(my_fig,'yield_plot2.png');

%
% The different initial guesses x0 will give different results when
% calling
% fminsearch and fminunc. But when calculating back the simply-
% compounded
% interest rate and swap rates, they give the same result.
% This might be that the fminsearch and fminunc found local minimizers
% and
% the value are both close to 0. The convergency rates are fast, but
% fminsearch use Nelder-Mead simplex method that does not require the
% minimizing problem to be differentiable and the error function can
% be
% discontinuous. While the fminunc function use the classic Newton
% method
% assuming that it is differentiable. fminsearch tend to give more
% robust
% results while fminunc might get stucked at local minimum. Finally,
% we
% can confirm that the obtained results for both methods are very
% accurate
% as they match almost perfectly the provided market data this is
% shown in
% the plots at the end of the report.
%
```

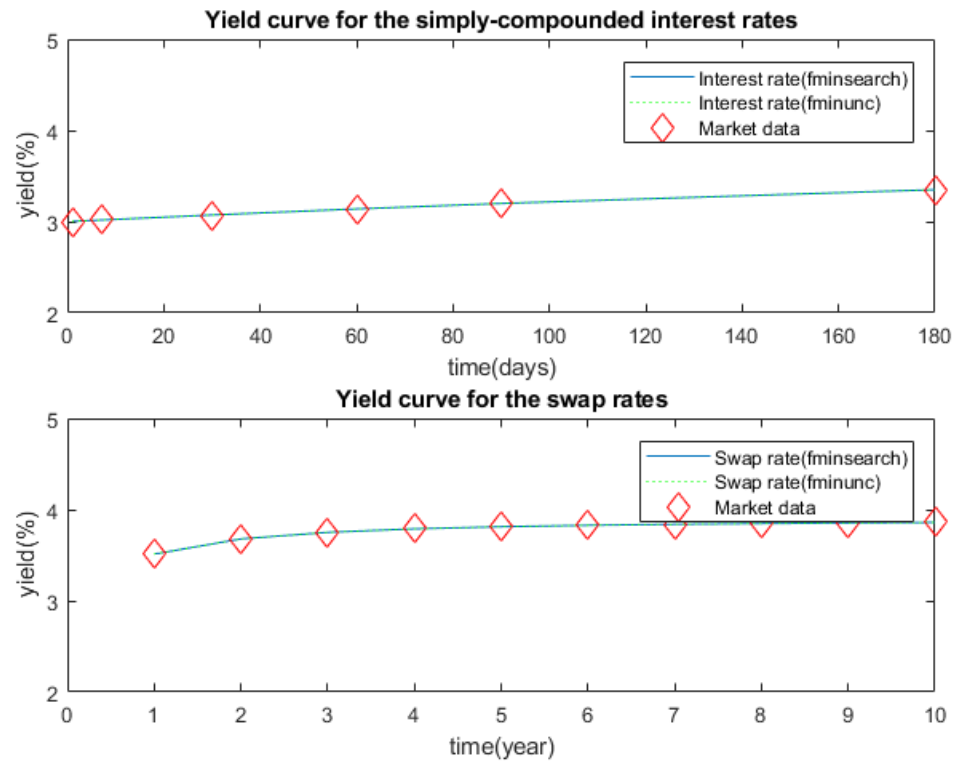
dbtype('cirmodel.m');

Local minimum found.

Optimization completed because the size of the gradient is less than the default value of the optimality tolerance.

```
1    function y_out = cirmodel(x,T1,T2,Lmarket,Smaket,swaptime)
2    %here x is vector input,
3    %x(1) = phi
4    %x(2) = k
5    %x(3) = sigma (volatility)
6    %x(4) = r0(present interest rate)
7
8    %T1 = [1 7 30 60 90 180]; %in days
9    %T1 = T1/360;
10   %Lmarket = [0.0300 0.0302 0.0307 0.0314 0.0320 0.0335];%,market
    libor rate
11
12   %B = (1 - exp(-x(2)* tauti))/x(2);%need to get tauti
13
14   for i = 1:size(T1,2)
15       tau(i) = T1(i) - 0;
16       B(i) = (1 - exp(-x(2)* tau(i)))/x(2);
17       A(i) = (x(1) - (x(3)*x(3)/(2*x(2)*x(2)))) * (B(i) - tau(i))
- (x(3)*x(3)*B(i)*B(i)/(4*x(2)));
18       Z(i) = exp(A(i) - B(i)*x(4));
19       L(i) = (1 - Z(i))/(tau(i)*Z(i));
20       Ldiff2(i) = ((L(i) - Lmarket(i))/Lmarket(i))^2;
21   end
22
23   %T2 = [1 2 3 4 5 6 7 8 9 10];%in years swap maturities time
24   %Smaket = [0.0351 0.0368 0.0375 0.0379 0.0381 0.0383 0.0384
0.0385 0.0385 0.0386];
25
26   %swaptime = 3/12;
27
28   for j = 1:size(T2,2)
29
30       tau(j) = T2(j) - 0;
31       B(j) = (1 - exp(-x(2)* tau(j)))/x(2);
32       A(j) = (x(1) - (x(3)*x(3)/(2*x(2)*x(2)))) * (B(j) - tau(j))
- (x(3)*x(3)*B(j)*B(j)/(4*x(2)));
33       Z(j) = exp(A(j) - B(j)*x(4));
34       n = 1;
35       for k = swaptime:swaptime:T2(j)
36
37           Bj(n) = (1 - exp(-x(2)*k))/x(2);
38           Aj(n) = (x(1) - (x(3)*x(3)/(2*x(2)*x(2)))) * (Bj(n) - k)
- (x(3)*x(3)*Bj(n)*Bj(n)/(4*x(2)));
39           Zj(n) = exp(Aj(n) - Bj(n)*x(4));
40           n = n + 1;
41
42       end
43
```

```
44     S(j) = (1 - Z(j))/(swaptime * sum(Zj));  
45     Sdiff2(j) = ((S(j) - Smarket(j))/Smarket(j))^2;  
46 end  
47  
48 y_out = sum(Ldiff2) + sum(Sdiff2);  
49 end
```



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