

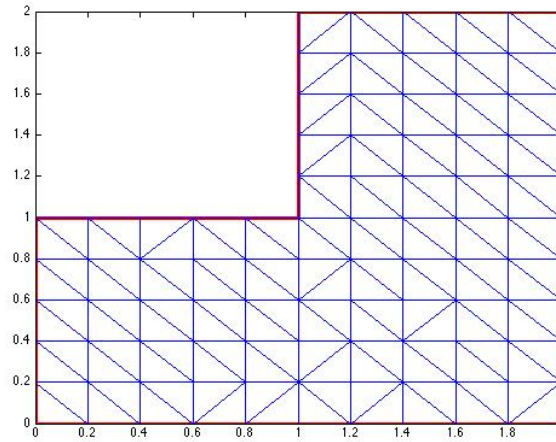
## HomeWork 8

### 1. The matlab code for creating triangular finite element grids :

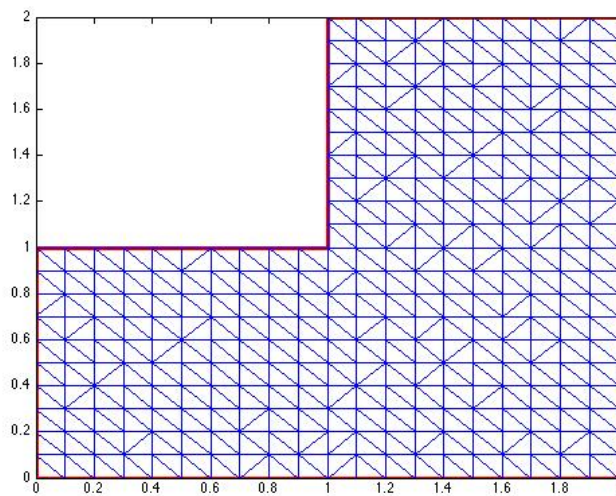
```
function [ sort, DT_I, P, Val, DT,M ] = generate_grid( h )
figure()
axis([0 2 0 2]);
axis equal;
M = floor(2/h) - 1;
sort = [];
Val = [];
P = [];
for i = 0:h:2
    for j = 0:h:2
        P = [P;j i];
    end
end
total_number = size(P,1);
n1 = 1;
n2 = M+2;
n3 = (M+2)*(M+2);
n4 = (M+2)*(M+1)+floor((M+2)/2)+1;
n5 = (M+2)*floor((M+2)/2)+floor((M+2)/2)+1;
n6 = (M+2)*floor((M+2)/2)+1;
for id = 1:total_number
    if (mod(id,M+2) == 1) && (id <= n6)
        sort = [sort;1];
        Val = [Val;1];
    elseif mod(id,M+2) == 0
        sort = [sort;1];
        Val = [Val;0];
    elseif mod(id,M+2) <= floor((M+2)/2) && mod(id,M+2) >= 1 && id > n5
        sort = [sort;2];
        Val = [Val; NaN];
    else
        sort = [sort;0];
        Val = [Val; NaN];
    end
end
C = [n1 n2;n2 n3;n3 n4;n4 n5;n5 n6;n6 n1];
plot(P(C'),P(C'+size(P,1)),'-r','LineWidth', 3);
DT = delaunayTriangulation(P, C);
IO = isInterior(DT);
hold on
triplot(DT(IO, :),DT.Points(:,1), DT.Points(:,2),'LineWidth', 1)
hold off
DT_I = DT(IO,:);
end
```

**Explanation:** In the program, the matrix *sort* stores the kinds of points, i.e., 1 represents the known points, 0 represents the unknown points and 2 represents the points in the left upper corner outside the L-shape region. The matrix *Val* stores the values of points and the values of unknown points are all *NaN* at first. And the matrix *P* stores the coordinates of all points. The information of these three matrixes corresponds to each other. The triangular finite elements are produced using matlab command *delaunayTriangulation()*. These triangles are produced with uniform size and random direction. Each row in the returned matrix *DT* could represent an element(triangle) and it stores three values which are the id of points within this element. These ids can help figure out all the local matrixes. The parameter of the function *h* represents the size of each element.

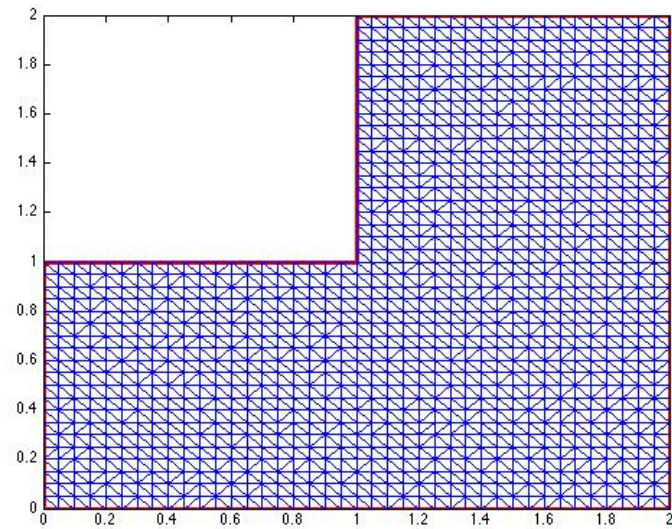
**Result:** Figure 1 to Figure 4 are the images of the finite element grids with  $h = 0.2$ ,  $h = 0.1$ ,  $h = 0.05$ ,  $h = 0.03$ .



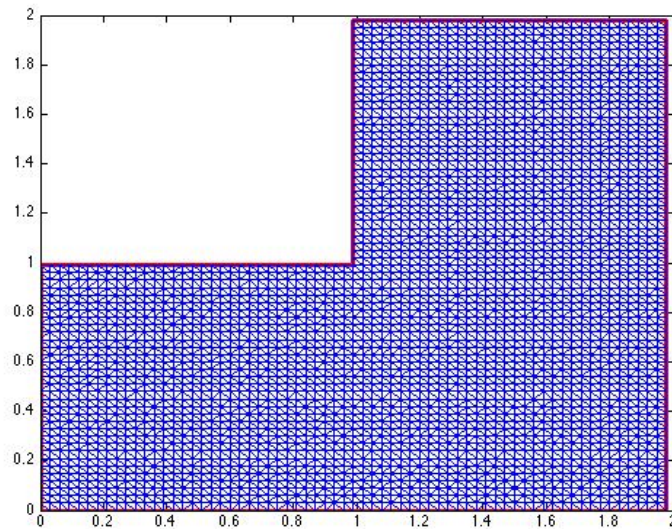
**Figure 1. The finite element grids with  $h = 0.2$**



**Figure 2. The finite element grids with  $h = 0.1$**



**Figure 3. The finite element grids with  $h = 0.05$**



**Figure 4. The finite element grids with  $h = 0.03$**

This is my first version of function **generate\_grid**. However, I have to deal with the points outside the L-region and it's troublesome. Hence, I wrote another version and I only produce points inside the L-region. The code is shown as below,

```
function [ sort, P, Val, DT, DT_I ] = generate_grid( h )
figure()
axis([0 2 0 2]);
axis equal;
M = floor(2/h);
h = 2/M;
sort = [];
```

```

Val = [];

P = [];
for i = 0:h:1
    for j = 0:h:2
        P = [P;j,i];
    end
end
for i = 1+h:h:2
    for j = 1:h:2
        P = [P;j,i];
    end
end
total_number = size(P,1);
for id = 1:total_number
    if P(id,1) == 0 && P(id, 2) == 0
        n1 = id;
    elseif P(id,1) == 2 && P(id,2) == 0
        n2 = id;
    elseif P(id,1) == 2 && P(id,2) == 2
        n3 = id;
    elseif P(id,1) == 1 && P(id,2) == 2
        n4 = id;
    elseif P(id,1) == 1 && P(id,2) == 1
        n5 = id;
    elseif P(id,1) == 0 && P(id,2) == 1
        n6 = id;
    end
    if P(id,1) == 0 && P(id,2) <= 1
        sort = [sort;1];
        Val = [Val;1];
    elseif P(id,1) == 2
        sort = [sort;1];
        Val = [Val;0];
    else
        sort = [sort;0];
        Val = [Val;NaN];
    end
end
C = [n1 n2;n2 n3;n3 n4;n4 n5;n5 n6;n6 n1];
plot(P(C'),P(C'+size(P,1)),'-r','LineWidth', 3);
DT = delaunayTriangulation(P, C);
IO = isInterior(DT);
hold on
triplot(DT(IO, :),DT.Points(:,1), DT.Points(:,2),'LineWidth', 1)
hold off
DT_I = DT(IO,:);
end

```

## 2. The Whole Process:

(1) Using function **Diff\_Triangle** to figure out the geometric coefficients  $a_i, b_i, c_i$ . The coefficient  $a_i$  is used to help figure out the right hand side. However, in this case, the Laplace's equation is  $\nabla^2 u = 0$ , hence,  $f = 0$ . In addition, the Neumann condition  $\partial u / \partial n = 0$  is automatically satisfied, so the r.h.s of each local matrix are all zero vectors and  $a_i$  is not needed.

### The matlab code of function Diff\_Triangle:

```
function [b,c,Area_A] = Diff_Triangle( x,y )
Area = x(2)*y(3) - x(3)*y(2) + x(3)*y(1) - x(1)*y(3) + x(1)*y(2) - x(2)*y(1);
Area_A = abs(Area)/2;
b = zeros(3,1);
c = zeros(3,1);
for i = 1:3
    i1 = mod(i,3) + 1;
    i2 = mod(i1,3) + 1;
    b(i) = (y(i1) - y(i2))/Area;
    c(i) = (x(i2) - x(i1))/Area;
end
end
```

**EXPLANATION:** The returned value  $Area\_A$  is  $A$  appears in the textbook,  $b(i)$  is just  $b_i/2A$  and  $a(i)$  is just  $a_i/2A$ . The input vectors  $x$  and  $y$  store the coordinates of the three points.

(2) Figure out the local matrix using the geometrix coefficients. The r.h.s are all zeros.

### The matlab code of functin Local\_Matrix used to construct local matrixes for each element:

```
function [ A ] = Local_Matrix(x,y)
[b,c,Area_A] = Diff_Triangle(x,y);
A = zeros(3,3);
for i = 1:3
    for j = 1:3
        A(i,j) = Area_A*(b(i)*b(j) + c(i)*c(j));
    end
end
end
```

(3) Combine all the local matrixes into a global matrix, apply the boundary conditions into r.h.s and modify the global matrix. The main idea of the combination process is  $a_{g[i]g[j]} \leftarrow a_{g[i]g[j]} + a_{ij}$ , where  $g[i]$  and  $g[j]$  is the id of the three points in the global matrix. And the main idea of dealing with boundary condition is that if  $u_i$  is known, then set all elements in the  $i$ th row and the  $i$ th column 0 and set element  $a_{ii} = 1$ , in addition, set the  $i$ th element in r.h.s equal to  $u_i$  and

subtract  $a_{ji}u_i$  from the  $j$ th element in r.h.s unless  $u_j$  is known as well. I wrote two versions of the combination process.

### The first version:

```
function [ Ag, fg ] = Global_Matrices(ver, Al, Ag, fg)
for row = 1:3
    for col = 1:3
        if ver(col).sort == 1
            if col ~= row
                if(ver(row).sort ~= 1)
                    fg(ver(row).id) = fg(ver(row).id) - Al(row, col)*ver(col).value;
                end
                Ag(ver(row).id, ver(col).id) = 0;
            else
                Ag(ver(row).id, ver(col).id) = 1;
            end
            fg(ver(col).id) = ver(col).value;
        elseif ver(row).sort == 1
            Ag(ver(row).id, ver(col).id) = 0;
            fg(ver(row).id) = ver(row).value;
        else
            Ag(ver(row).id, ver(col).id) = Ag(ver(row).id, ver(col).id) + Al(row, col);
        end
    end
end
end
```

**EXPLANATION:** This function combines the local matrix while modifying the global matrix and r.h.s at the same time. The input consists of Al(the local matrix), fl(the local r.h.s), Ag(the global matrix before the combination), fg(the r.h.s before the combination) and ver which stores the corresponding ids, sorts and values of the three points in the local matrix.

### The second version:

```
function [ Ag, fg ] = Global_Matrices1(ver, Al, fl, Ag, fg)
for i = 1:3
    fg(ver(i).id) = fg(ver(i).id) + fl(i);
    for j = 1:3
        Ag(ver(i).id, ver(j).id) = Ag(ver(i).id, ver(j).id) + Al(i, j);
    end
end
end
```

**EXPLANATION:** The function **Global\_Matrices** finishes the combination process by traversing all the local matrixes. Then use the codes below to deal with boundary conditions. The matrix D stores the ids of the known points.

```
D = [];
for i = 1:total_number
    if sort(i, 1) == 1
```

```

        fg(:,1) = fg(:,1) - Ag(:,i).*Val(i);
        Ag(i,:) = 0;
        Ag(:,i) = 0;
        Ag(i,i) = 1;
        D = [D;i];
    end
end
for i = 1:size(D,1)
    fg(D(i)) = Val(D(i));
end
end

```

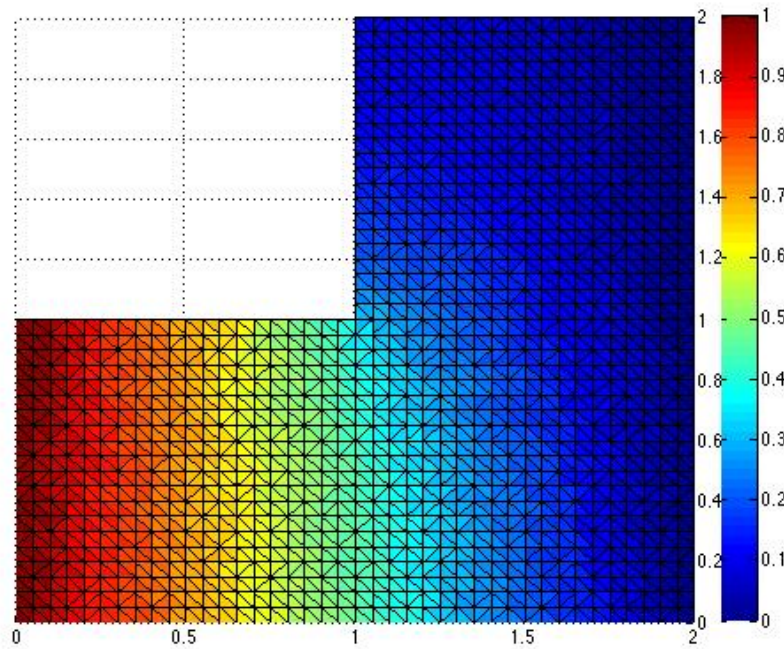
I wrote the function **get\_Aandb** to convert the finite element grids problem into a solving linear system  $Ax=b$  problem using the three functions above. The function will return the global matrix Ag and the r.h.s.

```

function [ Ag, fg ] = get_Aandb( DT_I, sort, P, Val )
    element_number = size(DT_I,1);
    total_number = size(P,1);
    M = sqrt(total_number) - 2;
    Ag = zeros(total_number, total_number);
    fg = zeros(total_number,1);
    for i = 1:element_number
        x = zeros(3,1);
        y = zeros(3,1);
        vert(3,1) = struct('id',[],'value',[],'sort',[]);
        for j = 1:3
            x(j) = P(DT_I(i,j),1);
            y(j) = P(DT_I(i,j),2);
            vert(j).id = DT_I(i,j);
            vert(j).sort = sort(vert(j).id);
            vert(j).value = Val(vert(j).id);
        end
        Al = Local_Matrix(x,y);
        [ Ag, fg ] = Global_Matrices(vert, Al, Ag, fg);
    end
    del = [];
    for i = floor((M+2)/2)+1:M+1
        for j = 1:floor((M+2)/2)
            t = i*(M+2) + j;
            del = [del;t];
        end
    end
    Ag(del,:) = [];
    Ag(:,del) = [];
    fg(del) = [];
end

```

Using these codes, solved the unknowns in these finite element grids with size  $h = 0.05$ , and then visualized the grids with command `trisurf(DT(:, :), DT.Points(:, 1), DT.Points(:, 2), res)`. Here, `res` is the solution of all the unknowns. I got the visualization as Figure 5 shows.



**Figure 5. The visualization of finite element grids with  $h = 0.05$**

3. Compared the Finite Element Method with the Five-Point Finite Difference Method with different values of  $h$  using `bigc` method.

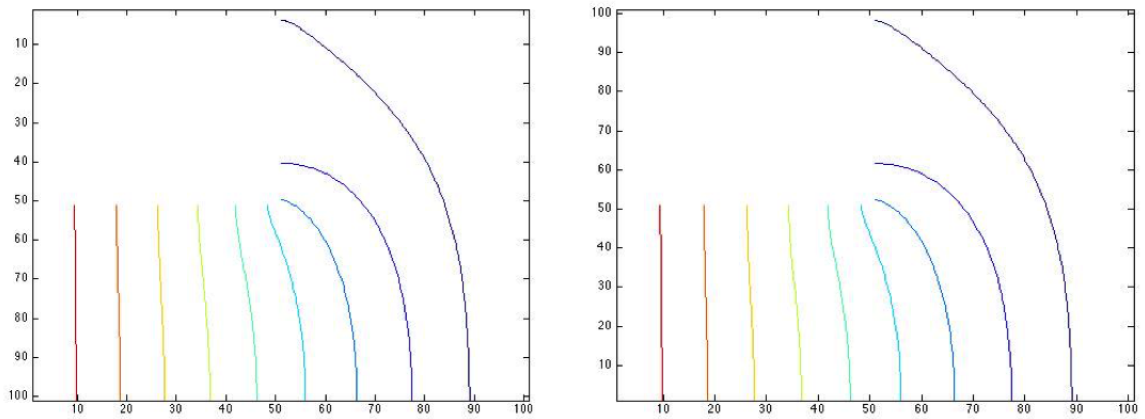
**Table 1. The comparisons of cputimes**

CPUTIME	$h=0.02$	$h=0.03$	$h=0.05$
FE	81.8700	11.68	1.3
FD	71.93	9.45	1.04

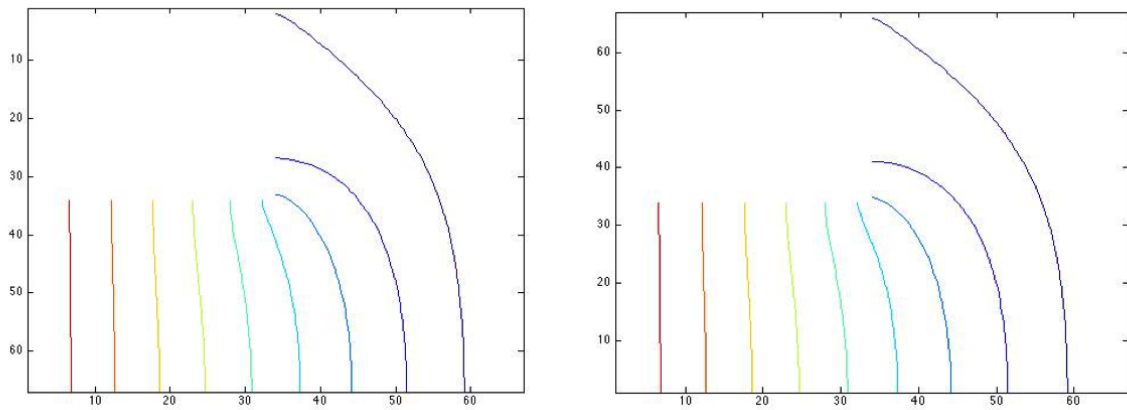
Obviously, with the same number of elements, the method FE needs more time than the FD method.

The figures below are the iso-value lines for both FE and FD methods with  $h = 0.02$ ,  $h = 0.03$ ,  $h = 0.05$ . The figures on the left side are the iso-value lines using FD and those on the right side are those using FE.

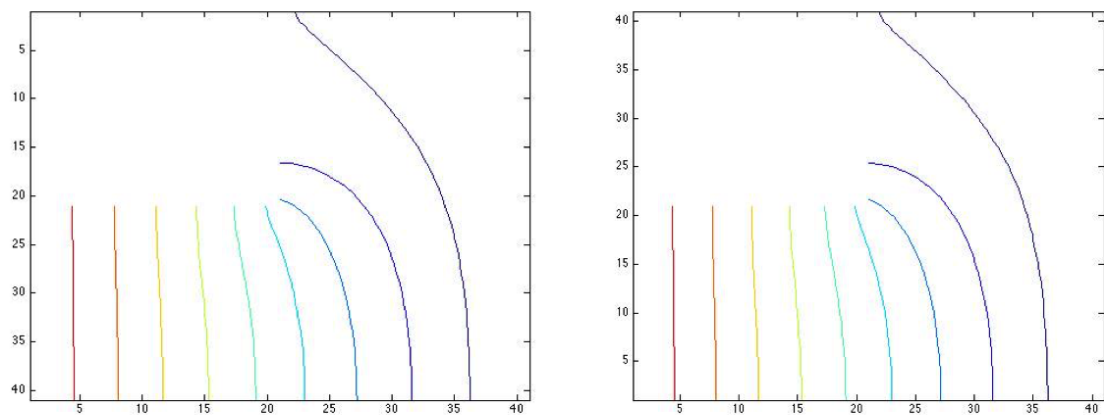




**Figure 6. The iso-value lines for FD and FE with  $h = 0.02$**



**Figure 7. The iso-value lines for FD and FE with  $h = 0.03$**



**Figure 8. The iso-value lines for FD and FE with  $h = 0.05$**

As these figures show, the two results of the two methods are almost the same.

	1	2	3	4	5	6	7	8	9	10	11
1	NaN	NaN	NaN	NaN	NaN	0.1083	0.1025	0.0864	0.0621	0.0324	0
2	NaN	NaN	NaN	NaN	NaN	0.1140	0.1077	0.0904	0.0648	0.0337	0
3	NaN	NaN	NaN	NaN	NaN	0.1321	0.1240	0.1028	0.0728	0.0376	0
4	NaN	NaN	NaN	NaN	NaN	0.1666	0.1532	0.1239	0.0862	0.0440	0
5	NaN	NaN	NaN	NaN	NaN	0.2279	0.1983	0.1535	0.1039	0.0523	0
6	1	0.8820	0.7627	0.6400	0.5082	0.3484	0.2587	0.1880	0.1235	0.0613	0
7	1	0.8826	0.7644	0.6446	0.5221	0.3990	0.3000	0.2163	0.1409	0.0696	0
8	1	0.8840	0.7679	0.6518	0.5367	0.4256	0.3259	0.2364	0.1541	0.0761	0
9	1	0.8854	0.7713	0.6582	0.5473	0.4409	0.3415	0.2494	0.1631	0.0806	0
10	1	0.8865	0.7736	0.6623	0.5536	0.4490	0.3500	0.2566	0.1682	0.0832	0
11	1	0.8869	0.7744	0.6636	0.5556	0.4516	0.3526	0.2590	0.1699	0.0841	0

11x11 double

	1	2	3	4	5	6	7	8	9	10	11
1	NaN	NaN	NaN	NaN	NaN	0.1045	0.0990	0.0834	0.0599	0.0313	0
2	NaN	NaN	NaN	NaN	NaN	0.1100	0.1040	0.0873	0.0626	0.0326	0
3	NaN	NaN	NaN	NaN	NaN	0.1276	0.1197	0.0993	0.0704	0.0364	0
4	NaN	NaN	NaN	NaN	NaN	0.1607	0.1479	0.1198	0.0834	0.0426	0
5	NaN	NaN	NaN	NaN	NaN	0.2192	0.1915	0.1487	0.1008	0.0508	0
6	1.0000	0.8803	0.7594	0.6346	0.4997	0.3333	0.2502	0.1826	0.1202	0.0598	0
7	1.0000	0.8811	0.7613	0.6397	0.5153	0.3904	0.2932	0.2115	0.1378	0.0681	0
8	1.0000	0.8826	0.7651	0.6478	0.5314	0.4197	0.3206	0.2324	0.1514	0.0747	0
9	1.0000	0.8842	0.7689	0.6547	0.5431	0.4363	0.3373	0.2460	0.1607	0.0794	0
10	1.0000	0.8854	0.7715	0.6592	0.5498	0.4450	0.3462	0.2535	0.1660	0.0821	0
11	1.0000	0.8858	0.7724	0.6607	0.5520	0.4478	0.3491	0.2560	0.1677	0.0830	0

**Figure 9. The comparison of the solutions for FE and FD with same number of elements, the upper one is for FD and the below one is for FE.**

As Figure 9 shows, the two solutions are nearly the same. Hence, FE and FD are both effective approaches.

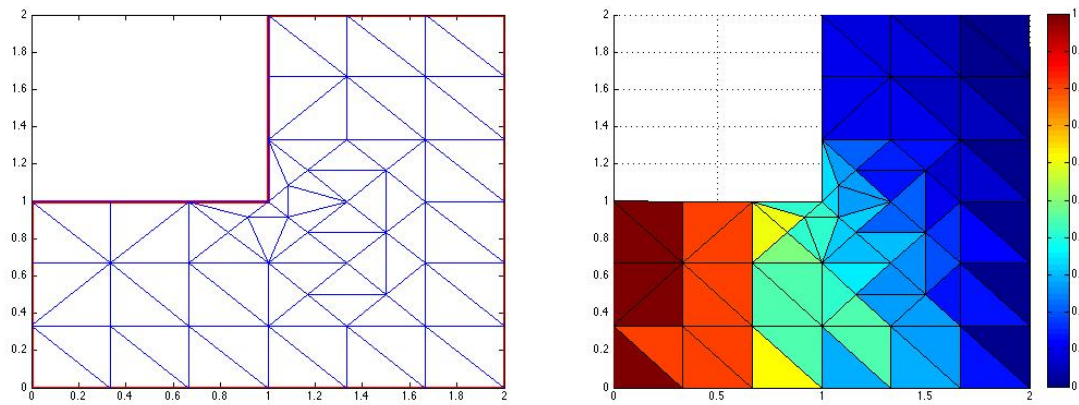
4. At first, read the data file using command **textread('point.text')** and there are triangles formed outside the L-shape in the upper left quadrant. Then I used the following commands to remove these triangles.

```
DT = delaunayTriangulation(P1, C);
IO = isInterior(DT);
hold on
triplot(DT(IO, :), DT.Points(:,1), DT.Points(:,2), 'LineWidth', 1);
hold off
```

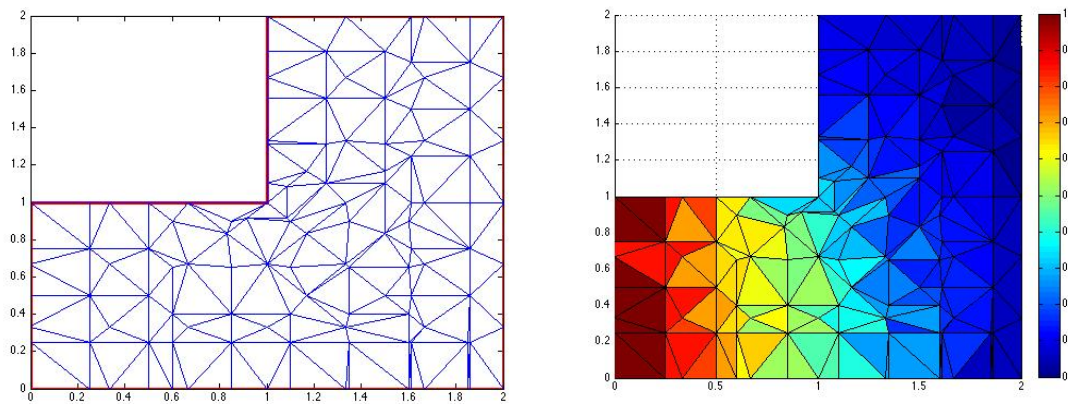
The command **isInterior** is to find the triangles inside the L-shape. And the command **triplot** is to visualize these triangles.

Using the functions above again, the temperatures of these points can be approximated. The triangles and the visualization of results using the FE method are shown in Figure 10. I found the solutions are not smooth especially in the center corner of the L-shape region where the meshes are unstructured. The reason is that the meshes are sparse and these in the center corner are unstructured which leads to most errors. Hence, I think the most errors occur in the center region and I added points everywhere especially in region around the center corner. I

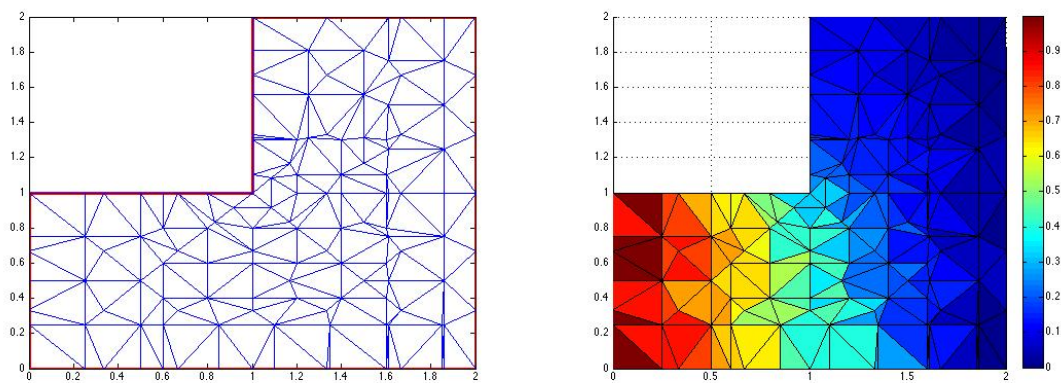
repeat the process again and again, the new meshes and visualizations are shown from Figure 11 to Figure 14.



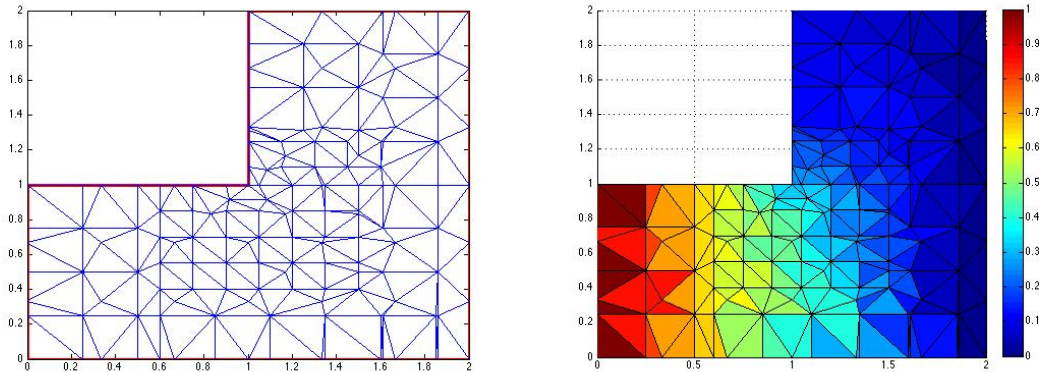
**Figure 10.**



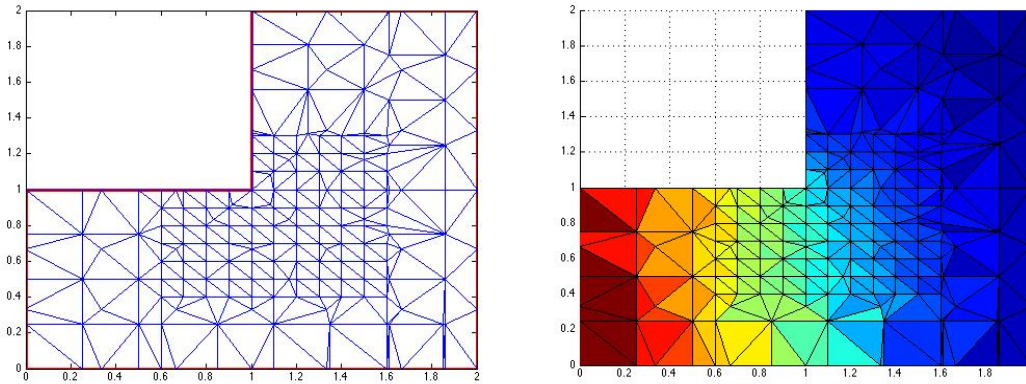
**Figure 11.**



**Figure 12.**



**Figure 13.**



**Figure 14.**

As the mesh in Figure 11 shows, I added points everywhere uniformly at first and then I stop adding points outside the center region and repeat adding points the center region as the meshes form Figure 12 to Figure 14 show. As a result, the solution becomes more and more accurate and the major error is reduced significantly. I judge the accuracy of solution by observing the visualizations. I implemented the visualization by using command - **trisurf(DT\_I,DT.Points(:,1), DT.Points(:,2), res)**. Here, res is solution of Finite Element Method. The colors in the visualization can represent the temperatures or values of points or triangles as the color bar in the figures show. The true result of this heat problem should seem like the visualization of Figure 5. The colors should have a steep tread that changes from dark red to dark blue gradually and smoothly. As Figure 10 to Figure 14 show, the visualization becomes more and more smooth. And the trend becomes more and more obvious. Especially, the trend can't be capatured in the center region as Figure 10 shows at first. After adding points, the colors in that region becom smooth and the trend becomes

obvious and steep. The visualization of Figure 14 is close to that of Figure 5, hence, the most error has been reduced and the solution has been more accurate.