

Working with user subroutines within MSC.Marc

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1 Prerequisites

In order to work with the MSC.Marc's subroutines, a compatible Fortran compiler should be installed. This can be done in a number of steps. First of all, Visual Studio (NOT the express edition) and Intel Visual Fortran were needed to be installed; the former (version 2017) was acquired from [Microsoft Imagine](#)¹, and a student version of the latter (version Parallel Studio XE 2018 (Update 1)) from [Intel Software](#). Once both were installed, the following steps were taken:

1. The folder of the installed Intel Visual Fortran was added to the windows environment.
2. Both codes were linked using the following code:

```
"C:\Program Files (x86)\IntelSWTools\compilers_and_libraries_2018\windows\bin\compilervars" intel64 vs20172
```

In order to test the compiler, the following simple code was written in a file named `helloworld.f`.

```
1 program helloworld
2 print *, "Hello world!"
3 end program helloworld
```

Then, it was compiled using the following command:

```
ifort helloworld.f
```

This command generated an executive file named `helloworld.exe`. Once the codes were installed and linked, as the next step, they should be tested within MSC.Marc. This was done by running the **example 2.35** from Marc's manual (*volume E*), in which a non-uniform load is defined within a user subroutine named `FORCEM`³. The simulation ran successfully.

In the following section, a series of simulations will be discussed for working with user subroutines.

2 Case Study

Basic System In this simulation, the goal is to apply a uniform load upon a square deformable material. The plate has a side length of $L = 10$, and a height of $h = 1$. The material's mechanical properties were assigned to be $E = 2 \times 10^5$ and $\nu = 0.3$. The position of the plate was fixed at its bottom nodes, and a uniform pressure of $p = 2000$ was applied to the top faces of the plate.

First the plate was generated, and divided into 20,20, and 3 sections along the x, y , and z axes, respectively. Following the subdivision, the `sweep_all` command was invoked, in order to make sure there are no duplicates. Then, a material with finite stiffness was generated with the assigned mechanical properties. It should be noted that the default value of `Mass Density = 0` was not altered. In order to fix the position of the bottom nodes, the `Fixed Displacement`, and for applying the pressure, the `Face Load` options were selected from the `Boundary Conditions` menu. Then, a `Structural` job was defined, and the `Linear Elastic Analysis` was turned on. Then, the job was submitted.

Moving Load Based on the prepared basic system, the next simulation was designed with the goal to control the applied pressure using a user subroutine: the position of the applied pressure is time dependent. In this study:

- The distributed load is applied within a square with a length of $b = 2$.
- The applying load moves with a velocity of $\sqrt{2}$ along the diagonal.
- The simulation runs during 10 time units with 20 equally sized increments.

In order to perform this simulation, a number of changes were applied:

1. The value of the applied pressure was set to 1, and the `Method` was changed to `User Sub. Forcem`.
2. From the `Loadcases` menu, a new `Structural Static` load case was created: the `Total Loadcase Time` and `#Steps` were assigned to be 10 and 20, respectively.
3. The moving load was controlled via a user subroutine. The subroutine was saved in a file (see appendix A for the code). Then, the file was loaded in the `Run Job` dialogue.

User Defined Material (UDM) Following the basic system, another simulation was designed for defining the material's mechanical behavior via an user subroutine. This was done by defining the mechanical responses of the simulated material in different regimes.

¹Older versions can be freely downloaded from [VisualStudio website](#).

²In order to make sure that this link is established whenever Marc is running, this command was added to the shortcut of the Marc Mentat.

³This user subroutine is extensively discussed in the MSC.Marc's manual *volume D*, pp. 62-66.

1. **Elastic:** the material was modeled as an Elastic-Plastic Anisotropic one, and the **User Subs.** `Hooklw/Anelas` was activated for defining the material using the corresponding subroutine.

The `Hooklw` was used in this study. The Hooke's law is written as $[\sigma] = [C][\epsilon]$; the stiffness tensor $[C]$ should be defined in the `Hooklw` subroutine. I tried two methods, (1) defining a compliance tensor $[S]$, as in $[\epsilon] = [S][\sigma]$, and using $[C] = [S]^{-1}$ (see Appendix B), and (2) defining $[C]$ as a function of the Lamé constants λ and μ (see Appendix C). Both methods resulted in the same values as the **Elastic-Plastic Isotropic** material, which was used in the basic system.

2. **Plastic:** the mechanical behavior was formulated within the `Wkslp` subroutine. I tried a sample of the `MSC.Marc`, called `barlat_yld2004_18p`. In the simulation, I set the plastic behavior to be read from the user subroutine, and I wrote the file according to the original formulation of $\sigma = 646(1 + \varepsilon)^{0.227}$. The simulation ran successfully, and the results were the same as the original ones. See Appendix D for the subroutine's code.

In order to activate both subroutines, each ones were written in a file, and a using the `#include` command, they were summoned in another file; then, this final file was called within the `MSC.Marc`.

Neural Network-based UDM In another attempt, the simulation described in 2 was performed by defining the material via a trained neural network. To do so, the stress-strain curve was imported in an in-house MATLAB code. Then, the trained network was converted into Fortran 90 language, and implemented within the `Wkslp` subroutine. The subroutine is presented in Appendix E; however, it should be noted that the code might not be the most suitable one to be used. Perhaps, the `urpflo` subroutine is a more appropriate one, which does not take care of the elastic regime of the deformation; in other words, the material is treated as a rigid-perfectly plastic one.

The preparation of the subroutine has been done in three stages: (1) running a NN trainer in MATLAB (Appendix), (2) saving the trained NN (Appendix), and (3) converting the saved NN into Fortran 90 (Appendix).

A Moving Load

```

1  subroutine forcem( press , p , d , nn , n )
2  #ifdef _IMPLICITNONE
3      implicit none
4  #else
5      implicit logical ( a-z )
6  #endif
7  #include "creeps"
8      dimension p(3) , d(3) , n(2)
9      real*8 b , v , dist
10     real*8 xc , xmax , xmin , yc , ymax , ymin
11
12     b = 2.0
13     v = sqrt(2.0)
14     dist = v*(cptim+timinc)
15     xc = 0.5*sqrt(2.0)*dist

```

```

16     yc = 0.5*sqrt(2.0)*dist
17
18     xmin = xc - b/2.0
19     xmax = xc + b/2.0
20     ymin = yc - b/2.0
21     ymax = yc + b/2.0
22
23     press = 0.0
24     if ( p(1) .le. xmax .and. p(1) .ge. xmin ) then
25     if ( p(2) .le. ymax .and. p(2) .ge. ymin ) then
26         press = 2000.0
27     endif
28     endif
29
30     return
31     end

```

It should be noted that the conditions in lines 24 and 25 could not be combined, due to some restrictions on the installed compiler.

B User material: Elastic 1

```

1  subroutine hooklw(m,nn,kcus,b,ngens,dt,dtdl,e,
2      pr,ndi,nshear,
3      * imod,rprops,iprops)
4  #ifdef _IMPLICITNONE
5      implicit none
6  #else
7      implicit logical (a-z)
8  #endif
9      real*8 b, dt, dtdl, e
10     integer imod, iprops, kcus, m, ndi, ngens, nn,
11     nshear
12     real*8 pr, rprops
13     dimension b(ngens,ngens),dt(*),dtdl(*),rprops
14     (*),iprops(*),
15     * kcus(2),m(2),e(*),pr(*)
16
17     real*8 Ym, nu, g
18
19     real*8 ex, ey, ez, uxy, uyz, uzx, gxy, gyx, gzx
20     real*8 sum, d
21     dimension sum(20)
22
23     imod = 1
24
25 c material constants
26     Ym = 200000.
27     nu = 0.3
28     g = Ym/2./(1.+nu)
29
30 c modified for anisotropic material
31     ex = Ym
32     ey = Ym
33     ez = Ym
34     uxy = nu
35     uyz = nu
36     uzx = nu
37     gxy = g
38     gyx = g
39     gzx = g
40
41 c compliance tensor
42     b(1,1)=1./ex
43     b(2,2)=1./ey
44     b(3,3)=1./ez
45     b(2,1)=-uxy/ex
46     b(3,2)=-uyz/ey
47     b(1,3)=-uzx/ez
48     b(3,1)=b(1,3)
49     b(1,2)=b(2,1)
50     b(2,3)=b(3,2)
51     b(4,4)=1./gxy

```

```

49      b(5,5)=1./gyz
50      b(6,6)=1./gzx
51
52 c    stiffness tensor: b
53      call invert(b,ngens,sum,0,d,ngens)
54
55      return
56 end

```

C User material: Elastic 2

```

1      subroutine hooklw(m,nn,kcus,b,ngens,dt,dtdl,e,
2      pr,ndi,nshear,
3      * imod,rprops,iprops)
4      #ifdef _IMPLICITNONE
5      implicit none
6      #else
7      implicit logical (a-z)
8      #endif
9      real*8 b, dt, dtdl, e
10     integer imod, iprops, kcus, m, ndi, ngens, nn,
11     nshear
12     real*8 pr, rprops
13     dimension b(ngens,ngens),dt(*),dtdl(*),rprops
14     (*),iprops(*),
15     * kcus(2),m(2),e(*),pr(*)
16
17     real*8 Ym, nu, lambda, mu
18     integer i, j
19     imod = 1
20
21     Ym = 200000.0
22     nu = 0.3
23
24     lambda = Ym*nu/(1.+nu)/(1.-2.*nu)
25     mu = Ym/(1.+nu)/2.
26
27     b = 0.
28     b(1,1) = 2.*mu + lambda
29     b(2,2) = b(1,1)
30     b(3,3) = b(1,1)
31     b(4,4) = mu
32     b(5,5) = b(4,4)
33     b(6,6) = b(4,4)
34     b(1,2) = lambda
35     b(1,3) = b(1,2)
36     b(2,1) = b(1,2)
37     b(2,3) = b(1,2)
38     b(3,1) = b(1,2)
39     b(3,2) = b(1,2)
40
41     return
42 end

```

D User material: Plastic

```

1      subroutine wkslp(m,nn,kcus,matus,slope,ebarp,
2      egrate,stryt,dt,
3      * ifirst)
4      #ifdef _IMPLICITNONE
5      implicit none
6      #else
7      implicit logical (a-z)
8      #endif
9      stryt = 646.*(0.025+ebarp)**0.227
10     slope = 646.*0.227*(0.025+ebarp)**(1.-0.227)
11     return
12 end

```

E User material: Neural Network

```

1      module NN_funcs

```

```

2      #ifdef _IMPLICITNONE
3      implicit none
4      #else
5      implicit logical (a-z)
6      #endif
7      contains
8
9      function mapminmax_apply(x, xoffset, gain, ymin)
10     result(y)
11     !implicit none
12     real*8, intent(in) :: x, xoffset, gain, ymin
13     real*8 :: y
14     y = x-xoffset
15     y = y*gain
16     y = y+ymin
17     end function mapminmax_apply
18
19     function tansig_apply(n,i,j) result(y)
20     !implicit none
21     INTEGER i,j
22     real*8 n(i,j)
23     real*8 y(i,j)
24     y = 2. / (1. + exp(-2.*n)) - 1.
25     end function tansig_apply
26
27     function mapminmax_reverse(y, xoffset, gain,
28     ymin, j) result(x)
29     !implicit none
30     integer j
31     real*8, intent(in) :: y(j), xoffset, gain, ymin
32     real*8 x(j)
33     x = y-ymin
34     x = x/gain
35     x = x+xoffset
36     end function mapminmax_reverse
37
38     end module NN_funcs
39
40     !=====
41     subroutine wkslp(m,nn,kcus,matus,slope,ebarp,
42     egrate,stryt,dt,
43     * ifirst)
44     #ifdef _IMPLICITNONE
45     implicit none
46     #else
47     implicit logical (a-z)
48     #endif
49     real*8 dt, ebarp, egrate
50     integer ifirst, kcus, m, matus, nn
51     real*8 slope, stryt
52     dimension matus(2),kcus(2)
53     ! In order to calculate the slope, stress will be
54     ! calculated at two strains:
55     ! (1) the current strain, and (2) strain + deps.
56     ! Their difference divided by deps
57     ! would be reported as the slope.
58     ! "deps" needs to possess a small value. In here,
59     ! arbitrarily, it is set to be 0.01.
60     real*8 deps, nexttyld
61     deps = 0.01
62
63     call NeuralNet(stryt,ebarp)
64     call NeuralNet(nexttyld,ebarp + deps)
65     slope = (stryt - nexttyld) / deps
66     return
67     end
68
69     !=====
70     subroutine NeuralNet(stress,strain)
71     use NN_funcs
72     #ifdef _IMPLICITNONE
73     implicit none
74     #else
75     implicit logical (a-z)
76     #endif
77     !Constants Definition
78     real*8 x_xoffset, x_gain, x_ymin
79     real*8 y_xoffset, y_gain, y_ymin

```

```

72 !Input
73     real*8 Xp1, strain
74 !Layer1
75     INTEGER, PARAMETER :: b1i=2,b1j=1
76     real*8, dimension(b1i,b1j) :: a1, b1, IW1_1,
       nn_sum
77 !Layer2
78     INTEGER, PARAMETER :: b2i=2,b2j=1
79     real*8, dimension(b2j) :: a2, b2, stress
80     real*8 LW2_1(b2i)
81 !Assign_values
82     x_xoffset = 0.
83     x_gain = 2.
84     x_ymin = -1.
85     y_xoffset = 646.
86     y_gain = 0.0181690135872346
87     y_ymin = -1.
88 !Layer_1
89     DATA b1 / 0.2344368269302832819,
90               1.9855625577702908924 /
91     DATA IW1_1 / 0.15802878281862445253,
92                 0.51641932364727805016 /
93 !Layer_2
94     b2 = -6.0679754121327071914
95     DATA LW2_1 / 5.2036136075577150706,
96                 5.1936881732563282554 /
97 !Simulation
98     Xp1 = strain
99 !Input_1
100     Xp1 = mapminmax_apply(Xp1, x_xoffset, x_gain,
101                          x_ymin)
102 !Layer_1
103     nn_sum = b1 + IW1_1 * Xp1
104     a1 = tansig_apply(nn_sum, b2i, b2j);
105 !Layer_2
106     a2 = b2 + matmul(LW2_1, a1)
107 !Output_1
108     stress = mapminmax_reverse(a2, y_xoffset, y_gain,
109                               y_ymin, b2j);
110
111     return
112 end

```

F User material: Neural Network: NN Trainer

```

1 clear; close all;
2 %% Input data
3 sz = 100;    %% of sample size
4 var = 1;    %% of variables
5 eps = linspace(0,1,sz);
6 k = 646; n = 0.227;
7 sig = k * (1+eps).^n; % + 10*rand(var,sz);
8 %% Train the NN
9 x = eps(:)';
10 y = sig(:)';
11 nn = 3;
12 net = feedforwardnet([4 3 2]);
13 % net = cascadeforwardnet(nn);
14 [net,tr] = train(net,x,y);
15 %% Extract the trained NN
16 genFunction(net, 'netFcn');
17 %% Test the NN
18 eps_test = .35;linspace(0,1,sz/10);
19 x = eps_test(:)';
20 % y_NN = net(x);
21 y_NN = netFcn(x);
22 %% Visualize the comparison
23 % y_NN = netFcn(x);
24 figure;
25 plot(eps, sig, '-');
26 hold on
27 plot(eps_test, y_NN, '*');

```

G User material: Neural Network: Trained NN

```

1 function [Y,Xf,Af] = netFcn(X,~,~)
2 %NETFCN neural network simulation function.
3 %
4 % Generated by Neural Network Toolbox function
   genFunction, 23-Aug-2018 15:31:29.
5 %
6 % [Y] = netFcn(X,~,~) takes these arguments:
7 %
8 % X = 1xTS cell, 1 inputs over TS timesteps
9 % Each X{1,ts} = 1xQ matrix, input #1 at timestep
   ts.
10 %
11 % and returns:
12 % Y = 1xTS cell of 1 outputs over TS timesteps.
13 % Each Y{1,ts} = 1xQ matrix, output #1 at timestep
   ts.
14 %
15 % where Q is number of samples (or series) and TS is
   the number of timesteps.
16
17 %#ok<*RPMTO>
18
19 % ===== NEURAL NETWORK CONSTANTS =====
20
21 % Input 1
22 x1_step1.xoffset = 0;
23 x1_step1.gain = 2;
24 x1_step1.ymin = -1;
25
26 % Layer 1
27 b1 =
   [-4.2331650311832484945;-1.8534005300867182342;0.3717
28
29 IW1_1 =
   [4.3486852475527291162;1.7811488781001498793;0.5534845
30
31 % Layer 2
32 b2 =
   [-2.7505303206879916367;-0.31118817730058662141;2.426
33
34 LW2_1 = [0.18059897243039635395
   0.44654030386170651123 1.6797400342346737734
   -1.7649989328324051652;0.66561540831778609473
   -0.29549659468472666557 -1.1562732833758548878
   0.052967542915231445588;0.021583755501849426206
   -1.0449273951358553081 0.064952170212388068982
   -1.3557824303581875736];
35
36 % Layer 3
37 b3 = [1.3319933641381176415;1.6671150760804414048];
38 LW3_2 = [-1.5508478424855232092
   0.53242300257854990875
   0.81836946123260534414;1.3440984454581468288
   -1.3071802986342959674 -1.1884459257320574288];
39
40 % Layer 4
41 b4 = 0.20836408570433107013;
42 LW4_3 = [-0.88808615067962504153
   1.6565370363349731786];
43
44 % Output 1
45 y1_step1.ymin = -1;
46 y1_step1.gain = 0.0181690135872346;
47 y1_step1.xoffset = 646;
48
49 % ===== SIMULATION =====
50
51 % Format Input Arguments
52 isCellX = iscell(X);
53 if ~isCellX

```

```

52 X = {X};
53 end
54
55 % Dimensions
56 TS = size(X,2); % timesteps
57 if ~isempty(X)
58 Q = size(X{1},2); % samples/series
59 else
60 Q = 0;
61 end
62
63 % Allocate Outputs
64 Y = cell(1,TS);
65
66 % Time loop
67 for ts=1:TS
68
69     % Input 1
70     Xp1 = mapminmax_apply(X{1,ts},x1_step1);
71
72     % Layer 1
73     a1 = tansig_apply(repmat(b1,1,Q) + IW1.1*Xp1);
74
75     % Layer 2
76     a2 = tansig_apply(repmat(b2,1,Q) + LW2.1*a1);
77
78     % Layer 3
79     a3 = tansig_apply(repmat(b3,1,Q) + LW3.2*a2);
80
81     % Layer 4
82     a4 = repmat(b4,1,Q) + LW4.3*a3;
83
84     % Output 1
85     Y{1,ts} = mapminmax_reverse(a4,y1_step1);
86 end
87
88 % Final Delay States
89 Xf = cell(1,0);
90 Af = cell(4,0);
91
92 % Format Output Arguments
93 if ~isCellX
94 Y = cell2mat(Y);
95 end
96 end
97
98 % ===== MODULE FUNCTIONS =====
99
100 % Map Minimum and Maximum Input Processing Function
101 function y = mapminmax_apply(x,settings)
102 y = bsxfun(@minus,x,settings.xoffset);
103 y = bsxfun(@times,y,settings.gain);
104 y = bsxfun(@plus,y,settings.ymin);
105 end
106
107 % Sigmoid Symmetric Transfer Function
108 function a = tansig_apply(n,~)
109 a = 2 ./ (1 + exp(-2*n)) - 1;
110 end
111
112 % Map Minimum and Maximum Output Reverse-Processing Function
113 function x = mapminmax_reverse(y,settings)
114 x = bsxfun(@minus,y,settings.ymin);
115 x = bsxfun(@rdivide,x,settings.gain);
116 x = bsxfun(@plus,x,settings.xoffset);
117 end

```

H User material: Neural Network: NN Convertor

```

1 clear all; clc;
2 seed_file_name = 'netFcn.m';
3 %% Read the netFcn.m file

```

```

4 fid = fopen(seed_file_name,'r');
5 netFcn = textscan(fid,'%s','whitespace',' ','
    delimiter',' \n');
6 netFcn = netFcn{:};
7 fclose(fid);
8 %% Find the number of hidden layers
9 idx = find(strcmp(netFcn, '% Output 1'));
10 idx = idx - 4;
11 L_fin = netFcn{idx};
12 idx = ismember(L_fin, '% Layer ');
13 L_fin(idx) = [];
14 L_fin = str2num(L_fin);
15 %% Collect the network constants
16 %x_step
17 idx = find(strcmp(netFcn, '% Input 1'));
18 for i = 1:3
19     x_step(i) = read_data(netFcn{idx+i});
20 end
21
22 %y_step
23 idx = find(strcmp(netFcn, '% Output 1'));
24 for i = 1:3
25     y_step(i) = read_data(netFcn{idx+i});
26 end
27
28 % Layers
29 idx = find(strcmp(netFcn, '% Layer 1'));
30 const{L_fin,2} = {};
31 for i = 1:L_fin
32     const{i,1} = read_data(netFcn{idx+1}); %b
33     const{i,2} = read_data(netFcn{idx+2}); %W
34     idx = idx + 4;
35 end
36 %% Prepare for writing a Fortran subroutine
37 % Header
38 text{1,1} = '      include ''NN_Funcs.f'' ';
39 text{2,1} = '      subroutine NeuralNet(stress ,
    strain)';
40 text{3,1} = '      use NN_funcs';
41 text{4,1} = '#ifdef _IMPLICITNONE';
42 text{5,1} = '      implicit none';
43 text{6,1} = '#else';
44 text{7,1} = '      implicit logical (a-z)';
45 text{8,1} = '#endif';
46 % Definer
47 text{9,1} = '      real*8 x_xoffset , x_gain ,
    x_ymin';
48 text{10,1} = '      real*8 y_xoffset , y_gain ,
    y_ymin';
49 text{11,1} = '      real*8 a0 , strain';
50 text{12,1} = '      real*8 , dimension(1) :: stress '
    ;
51 idx = size(text,1);
52 formatL1 = '      real*8 , dimension(%i,1) :: b%i ,
    a%i';
53 formatL2 = '      real*8 , dimension(%i,%i) :: W%i
    ' ;
54 Layers_size(1) = 1;
55 Layers_size(L_fin+1) = 1;
56 for i=2:L_fin
57     Layers_size(i) = size(const{i-1},1);
58     text{idx+1,1} = sprintf(formatL1,Layers_size(i) ,
    i-1,i-1);
59     text{idx+2,1} = ...
60         sprintf(formatL2,Layers_size(i),Layers_size(
    i-1),i-1);
61     idx = idx + 2;
62 end
63 formatL1 = '      real*8 , dimension(%i) :: b%i , a
    %i';
64 formatL2 = '      real*8 , dimension(%i) :: W%i';
65 text{idx+1,1} = sprintf(formatL1,Layers_size(L_fin
    +1),i,i);
66 text{idx+2,1} = sprintf(formatL2,Layers_size(i),i);
67 % Steps (x and y)
68 idx = size(text,1);

```



```

69 text{idx+1,1} = [ '          x_offset = ', num2str(
    x_step(1))];
70 text{idx+2,1} = [ '          x_gain = ', num2str(x_step
    (2))];
71 text{idx+3,1} = [ '          x_ymin = ', num2str(x_step
    (3))];
72 text{idx+4,1} = [ '          y_offset = ', num2str(
    y_step(1))];
73 text{idx+5,1} = [ '          y_gain = ', num2str(y_step
    (2))];
74 text{idx+6,1} = [ '          y_ymin = ', num2str(y_step
    (3))];
75 % bs and Ws
76 idx = size(text,1);
77 formatLbW = '          data %s%i / %s /';
78 for i=1:L_fin
79     b_string = sprintf('%20f, ' , const{i,1}');
80     b_string = b_string(1:end-2);
81     W_temp = const{i,2};
82     W_string = sprintf('%20f, ' , W_temp(:)');
83     W_string = W_string(1:end-2);
84
85     text{idx+1,1} = sprintf(formatLbW, 'b', i, b_string
    );
86     text{idx+2,1} = sprintf(formatLbW, 'W', i, W_string
    );
87     idx = idx + 2;
88 end
89 % Simulation Section
90 idx = size(text,1);
91 text{idx+1,1} = ...
92     '          a0 = mapminmax_apply(strain , x_offset ,
    x_gain , x_ymin)';
93 formatLa = '          a%i = tansig_apply (b%i+W%i*a%i
    ,%i,%i)';
94 idx = size(text,1);
95 for i=1:L_fin-1
96     text{idx+1,1} = sprintf(formatLa,i,i,i,i-1,
    Layers_size(i+1),1);
97     idx = idx + 1;
98 end
99 formatLafin = '          a%i = b%i + matmul(W%i,a%i)';
100 text{idx+1,1} = sprintf(formatLafin , L_fin , L_fin ,
    L_fin , i);
101 formatLY = ...
102     '          stress = mapminmax_reverse(a%i , y_offset
    , y_gain , y_ymin , 1)';
103 text{idx+2,1} = sprintf(formatLY , L_fin);
104 % Ending
105 idx = size(text,1);
106 text{idx+1,1} = '          return';
107 text{idx+2,1} = '          end';
108 %% Write down the subroutine file named NeuralNet.f
109 fid = fopen('NeuralNet.f','w');
110 for i = 1:size(text,1)
111     fprintf(fid , '%s\n',text{i});
112 end
113 fclose(fid);

```

This code requires the following function.

```

1 function number = read_data(string)
2 %read_data
3 idx_temp1 = ismember(string , '=' );
4 idx_temp2 = find(idx_temp1,1);
5 idx_temp1(1:idx_temp2) = 1;
6 string(idx_temp1) = [];
7 number = str2num(string);
8 end

```

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```

2 subroutine NeuralNet(stress , strain)
3     use NN_funcs
4 #ifdef _IMPLICITNONE
5     implicit none
6 #else
7     implicit logical (a-z)
8 #endif
9     real*8 x_offset , x_gain , x_ymin
10    real*8 y_offset , y_gain , y_ymin
11    real*8 a0 , strain
12    real*8 , dimension(1) :: stress
13    real*8 , dimension(4,1) :: b1 , a1
14    real*8 , dimension(4,1) :: W1
15    real*8 , dimension(3,1) :: b2 , a2
16    real*8 , dimension(3,4) :: W2
17    real*8 , dimension(2,1) :: b3 , a3
18    real*8 , dimension(2,3) :: W3
19    real*8 , dimension(1) :: b4 , a4
20    real*8 , dimension(2) :: W4
21    x_offset = 0
22    x_gain = 2
23    x_ymin = -1
24    y_offset = -1
25    y_gain = 0.018169
26    y_ymin = 646
27    data b1 / -4.23316503118324849453,
    -1.85340053008671823420, 0.37171248405505208368,
    -5.39137284079696055272 /
28    data W1 / 4.34868524755272911619,
    1.78114887810014987934, 0.55348457993771060792,
    -3.86835244323420823775 /
29    data b2 / -2.75053032068799163667,
    -0.31118817730058662141, 2.42637658741737949342
    /
30    data W2 / 0.18059897243039635395,
    0.66561540831778609473, 0.02158375550184942621,
    0.44654030386170651123, -0.29549659468472666557,
    -1.04492739513585530808,
    1.67974003423467377338, -1.15627328337585488782,
    0.06495217021238806898,
    -1.76499893283240516517, 0.05296754291523144559,
    -1.35578243035818757356 /
31    data b3 / 1.33199336413811764146,
    1.66711507608044140483 /
32    data W3 / -1.55084784248552320918,
    1.34409844545814682881, 0.53242300257854990875,
    -1.30718029863429596737, 0.81836946123260534414,
    -1.18844592573205742880 /
33    data b4 / 0.20836408570433107013 /
34    data W4 / -0.88808615067962504153,
    1.65653703633497317860 /
35    a0 = mapminmax_apply(strain , x_offset , x_gain
    , x_ymin)
36    a1 = tansig_apply(b1+W1*a0,4,1)
37    a2 = tansig_apply(b2+W2*a1,3,1)
38    a3 = tansig_apply(b3+W3*a2,2,1)
39    a4 = b4 + matmul(W4,a3)
40    stress = mapminmax_reverse(a4 , y_offset ,
    y_gain , y_ymin , 1)
41    return
42 end

```