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LAMPIRAN**• HASIL KLASIFIKASI GAMBAR****KONDISI POOR**

CATATAN:

Kondisi: **POOR**

Kerusakan pada lapisan/area terdampak korosi: 90.33%

Assesment Scale 90.33%



KONDISI POOR**CATATAN:****Kondisi: POOR**

Kerusakan pada lapisan/area terdampak korosi: 65.5%

Assesment Scale 65.5%



KONDISI POOR

CATATAN:

Kondisi: **POOR**

Kerusakan pada lapisan/area terdampak korosi: 70.83%

Assesment Scale 70.83%



KONDISI POOR



CATATAN:

Kondisi: **POOR**

Kerusakan pada lapisan/area terdampak korosi: 72%

Assesment Scale 72%



KONDISI POOR



CATATAN:

Kondisi: **POOR**

Kerusakan pada lapisan/area terdampak korosi: 56.33%

Assesment Scale 56.33%



KONDISI POOR



CATATAN:

Kondisi: **POOR**

Kerusakan pada lapisan/area terdampak korosi: 50.33%

Assesment Scale 50.33%



KONDISI POOR



CATATAN:

Kondisi: **POOR**

Kerusakan pada lapisan/area terdampak korosi: 74.16%

Assesment Scale 74.16%



KONDISI POOR



CATATAN:

Kondisi: **POOR**

Kerusakan pada lapisan/area terdampak korosi: 78.16%

Assesment Scale 78.16%



KONDISI POOR



CATATAN:

Kondisi: **POOR**

Kerusakan pada lapisan/area terdampak korosi: 56.67%

Assesment Scale 56.67%



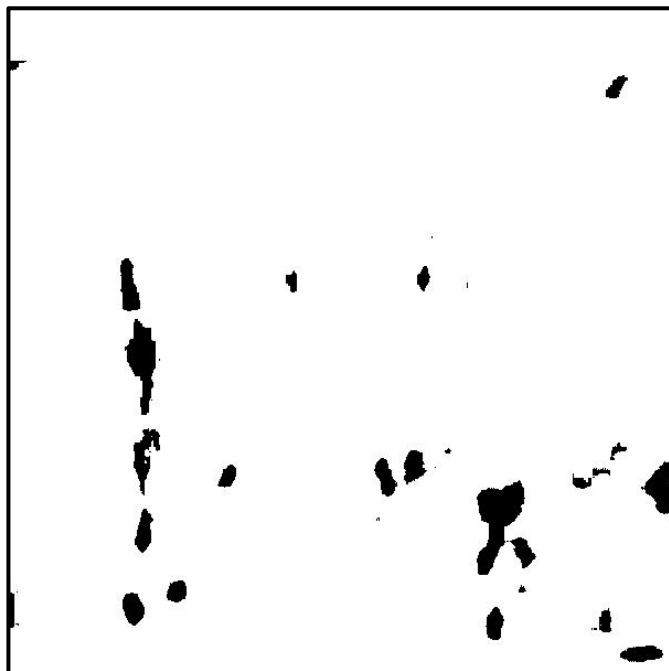
KONDISI FAIR

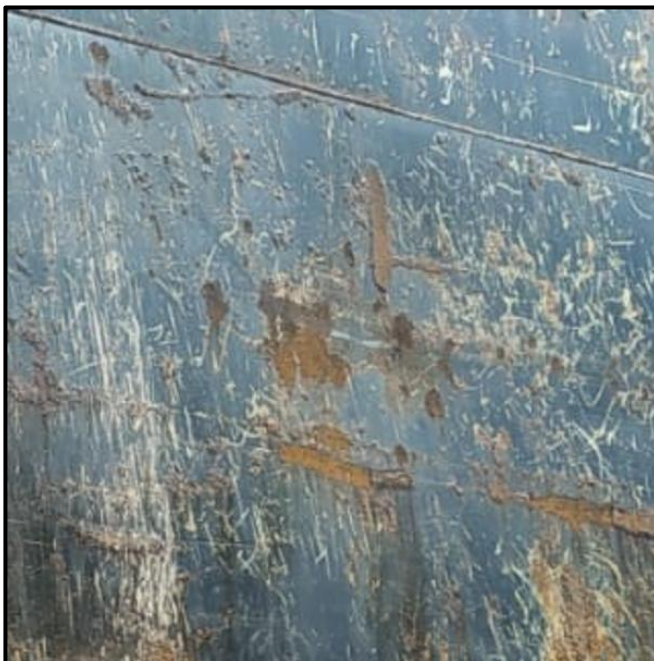
CATATAN:

Kondisi: **FAIR**

Kerusakan pada lapisan/area terdampak korosi: 6.83%

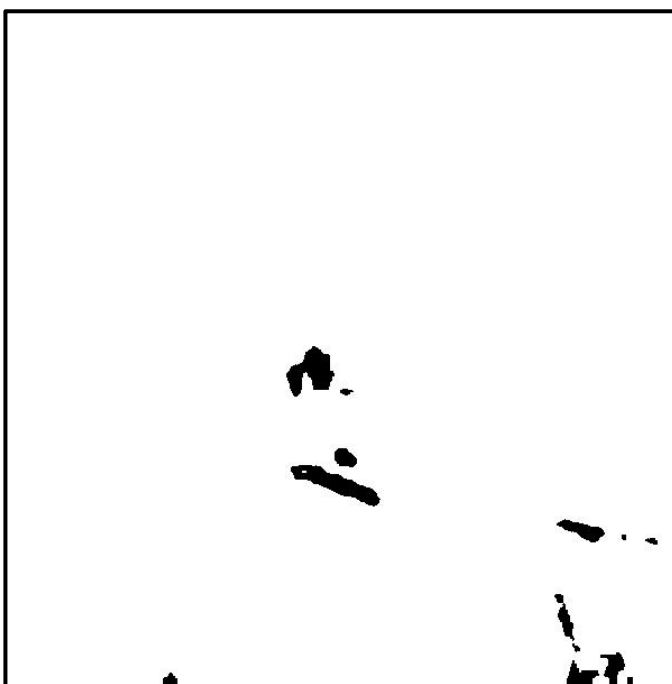
Assesment Scale 6.83%



KONDISI FAIR**CATATAN:**Kondisi: **FAIR**

Kerusakan pada lapisan/area terdampak korosi: 3.27%

Assesment Scale 3.27%



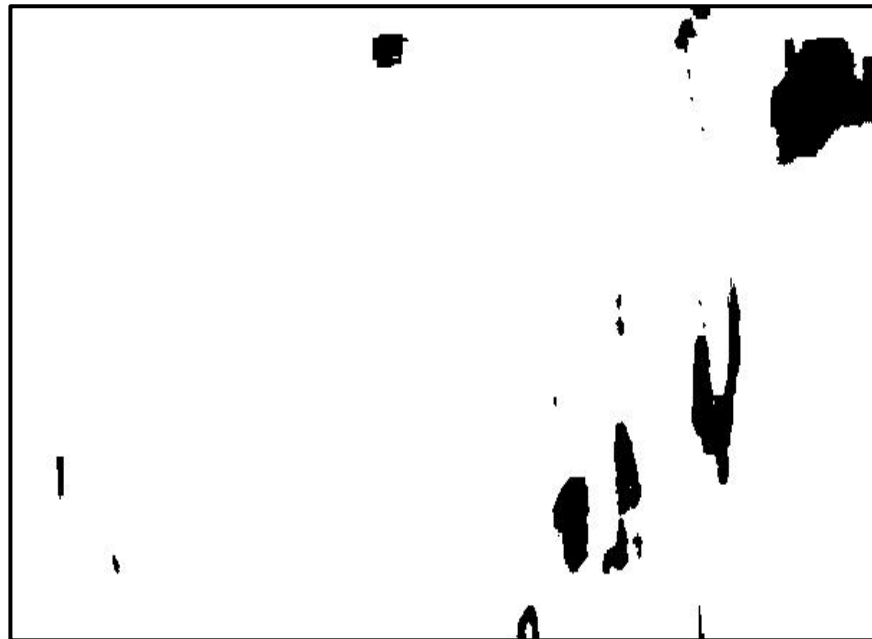
KONDISI FAIR

CATATAN:

Kondisi: **FAIR**

Kerusakan pada lapisan/area terdampak korosi: 3.96%

Assesment Scale 3.96%



KONDISI FAIR



CATATAN:

Kondisi: **FAIR**

Kerusakan pada lapisan/area terdampak korosi: 5.65%

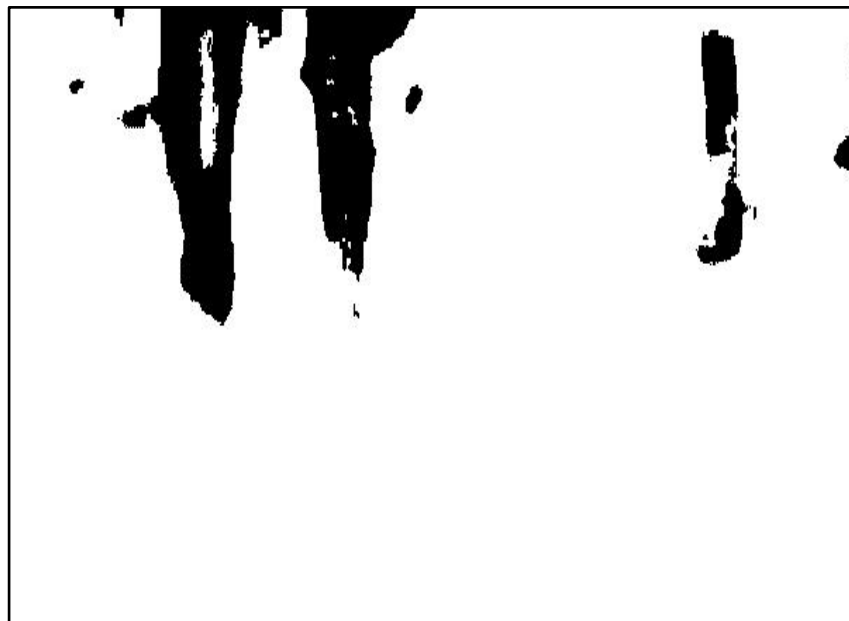
Assesment Scale 5.65%



KONDISI FAIR**CATATAN:****Kondisi: FAIR**

Kerusakan pada lapisan/area terdampak korosi: 8.16%

Assesment Scale 8.16%



KONDISI FAIR**CATATAN:****Kondisi: FAIR**

Kerusakan pada lapisan/area terdampak korosi: 8.16%

Assesment Scale 8.16%



KONDISI FAIR**CATATAN:****Kondisi: FAIR**

Kerusakan pada lapisan/area terdampak korosi: 9.8%

Assesment Scale 9.8%



KONDISI FAIR



CATATAN:

Kondisi: **FAIR**

Kerusakan pada lapisan/area terdampak korosi: 5.75 %

Assesment Scale 5.75%



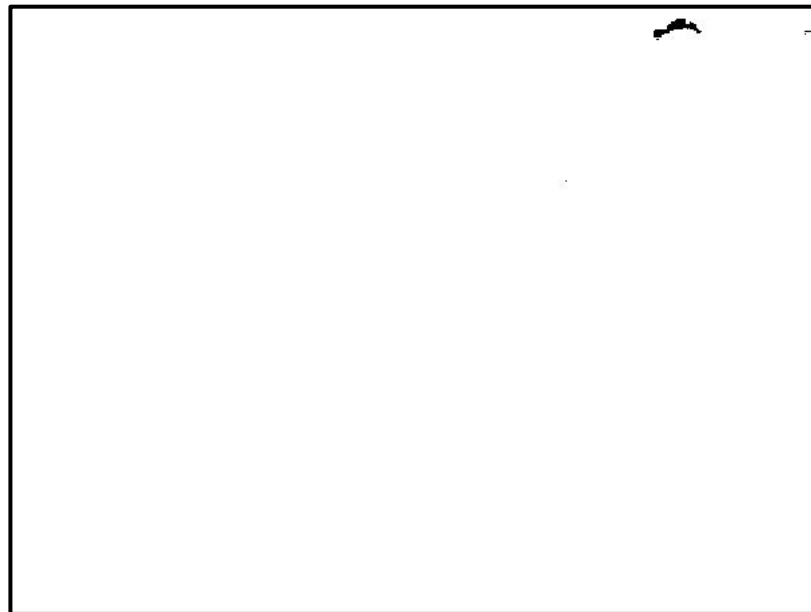
KONDISI GOOD

CATATAN:

Kondisi: GOOD

Kerusakan pada lapisan/area terdampak korosi: 0.075%

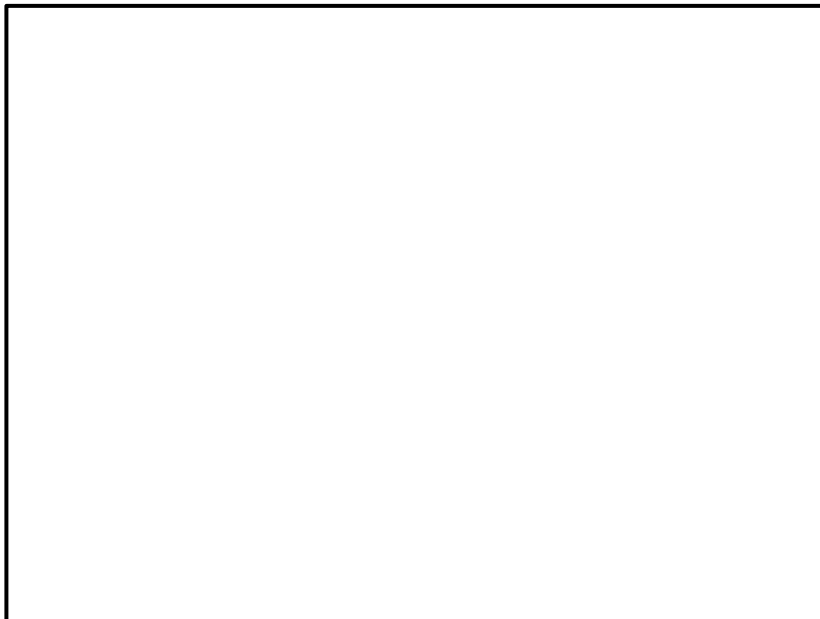
Assesment Scale 0.075%



KONDISI GOOD**CATATAN:****Kondisi: GOOD**

Kerusakan pada lapisan/area terdampak korosi: 0%

Assesment Scale 0%



KONDISI GOOD

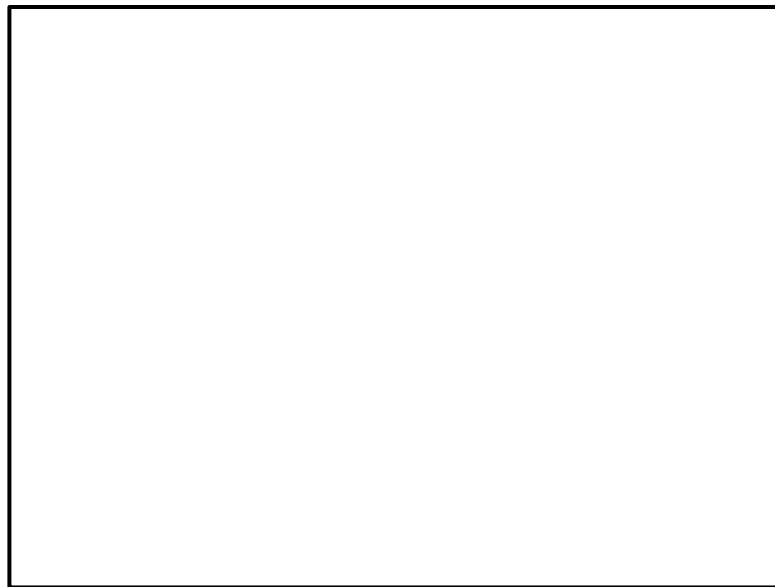


CATATAN:

Kondisi: **GOOD**

Kerusakan pada lapisan/area terdampak korosi: 0%

Assesment Scale 0%



KONDISI GOOD

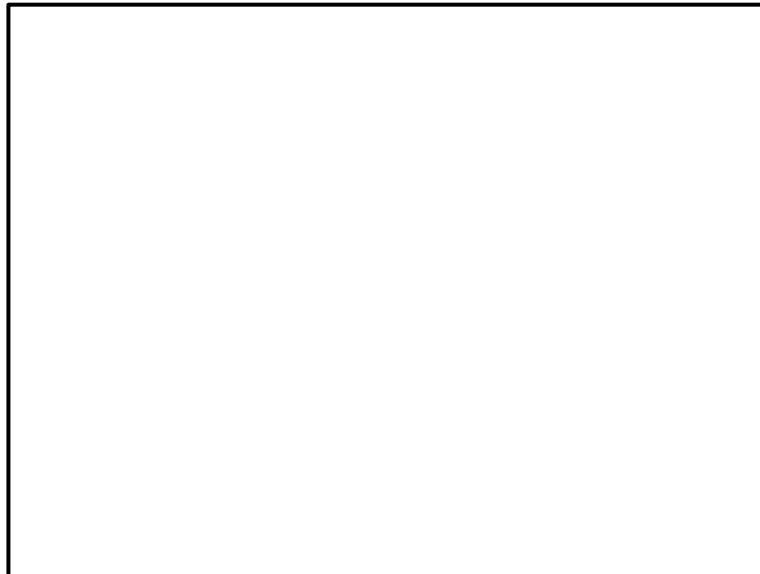


CATATAN:

Kondisi: **GOOD**

Kerusakan pada lapisan/area terdampak korosi: 0%

Assesment Scale 0%



KONDISI GOOD

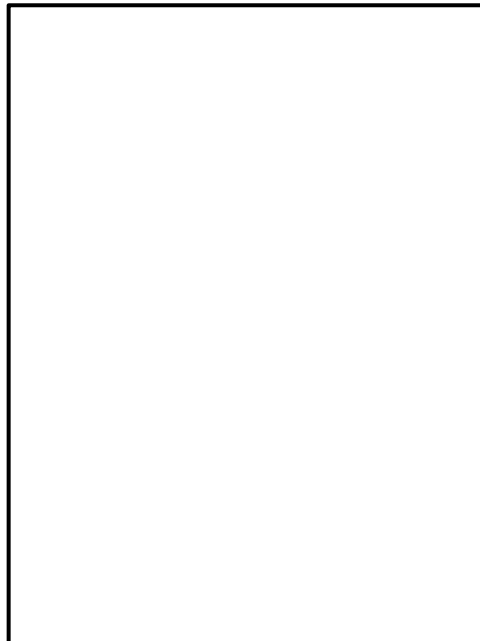


CATATAN:

Kondisi: **GOOD**

Kerusakan pada lapisan/area terdampak korosi: 0%

Assesment Scale 0%



KONDISI GOOD**CATATAN:****Kondisi: GOOD**

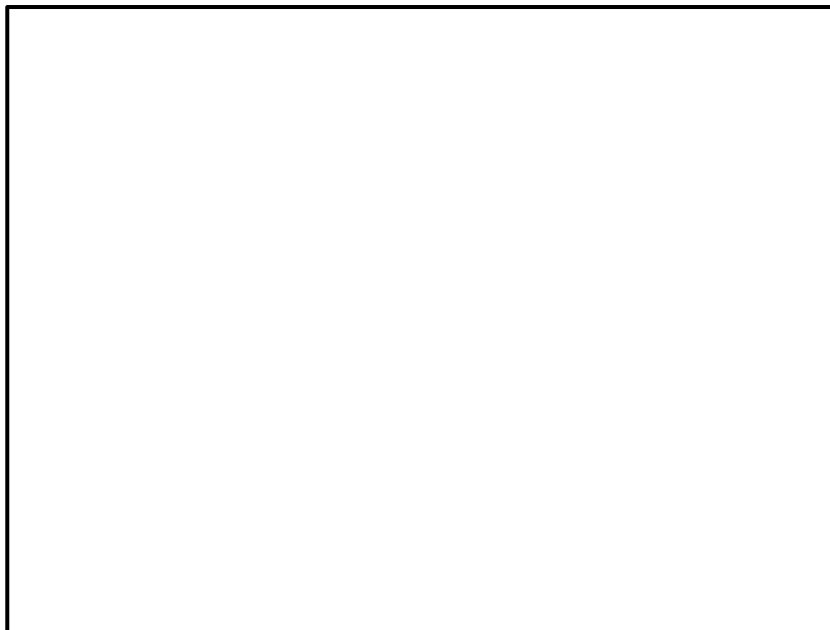
Kerusakan pada lapisan/area terdampak korosi: 0%

Assesment Scale 0%



KONDISI GOOD**CATATAN:****Kondisi: GOOD**

Kerusakan pada lapisan/area terdampak korosi: 0%

Assesment Scale 0%

KONDISI GOOD

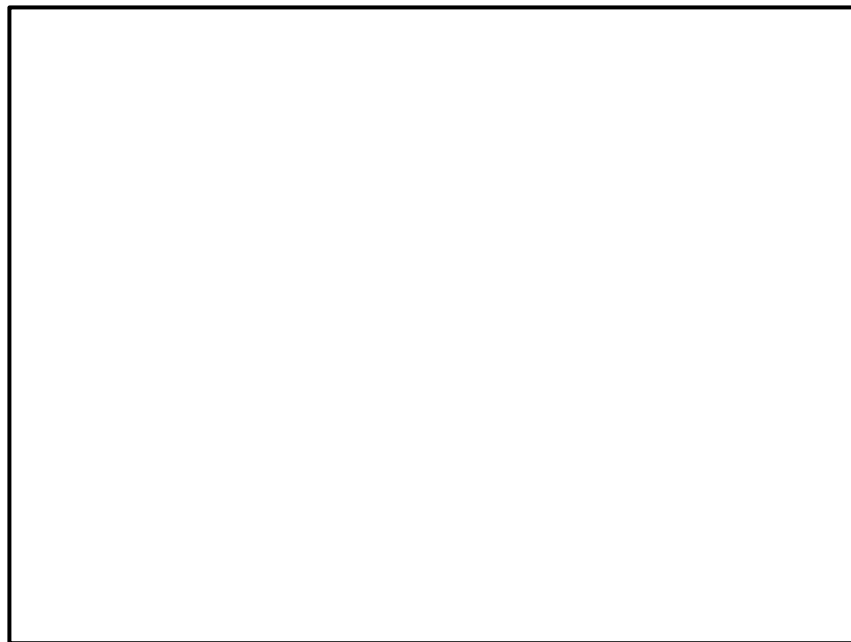


CATATAN:

Kondisi: **GOOD**

Kerusakan pada lapisan/area terdampak korosi: 0%

Assesment Scale 0%



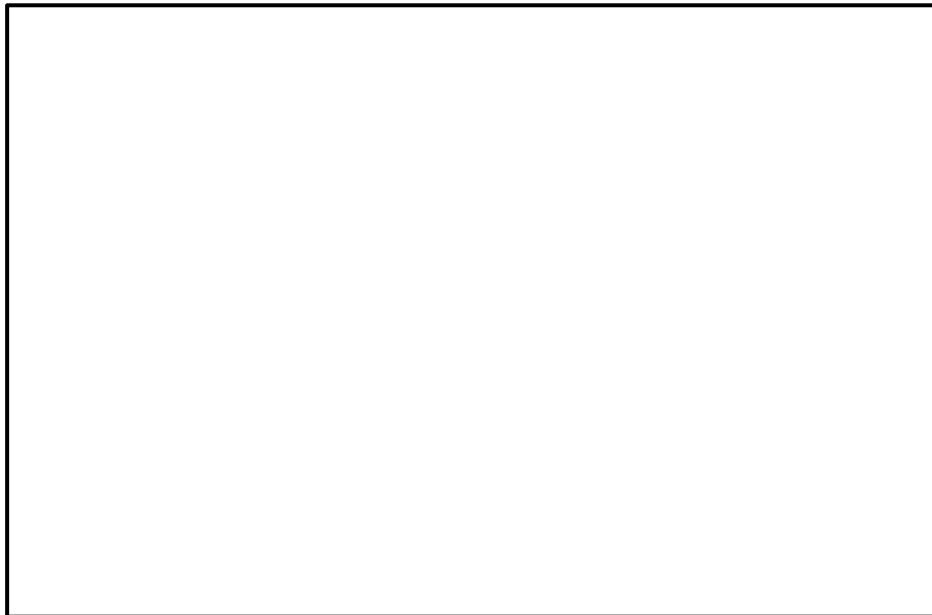
KONDISI GOOD

CATATAN:

Kondisi: GOOD

Kerusakan pada lapisan/area terdampak korosi: 0%

Assesment Scale 0%



KONDISI FAIR**CATATAN:****Kondisi: FAIR**

Kerusakan pada lapisan/area terdampak korosi: 4.9%

Assesment Scale 4.9%



- **Code Program Filter Color Threshold**

```
% Clear memory and command window
clc,clear,close all;
warning off

%% Uploading input image
[filename,pathname] = uigetfile('*.','Choose the input
image');
im = imread([pathname,filename]);      % merubah image
menjadi nilai numerik
%set the image size to suitable value
scale = 600/(max(size(im(:,:,1))));  %Normalization
im = imresize(im,scale*size(im(:,:,1)));
% Image resize
[~,n,~] = size(im);

I = rgb2hsv(im);

% Define thresholds for channel 1 based on histogram
settings
channel1Min = 0.000;
channel1Max = 1.000;

% Define thresholds for channel 2 based on histogram
settings
channel2Min = 0.000;
channel2Max = 1.000;

% Define thresholds for channel 3 based on histogram
settings
channel3Min = 0.000;
```

```

channel3Max = 1.000;

% Create mask based on chosen histogram thresholds
sliderBW = (I(:,:,1) >= channel1Min ) & (I(:,:,1) <=
channel1Max) & ...
           (I(:,:,2) >= channel2Min ) & (I(:,:,2) <=
channel2Max) & ...
           (I(:,:,3) >= channel3Min ) & (I(:,:,3) <=
channel3Max);

% Create mask based on selected regions of interest on
point cloud projection
I = double(I);
[m,n,~] = size(I);
polyBW = false([m,n]);
I = reshape(I, [m*n 3]);

% Convert HSV color space to canonical coordinates
Xcoord = I(:,2).*I(:,3).*cos(2*pi*I(:,1));
Ycoord = I(:,2).*I(:,3).*sin(2*pi*I(:,1));
I(:,1) = Xcoord;
I(:,2) = Ycoord;
%plot(Xcoord,Ycoord)
% clear Xcoord Ycoord

% Project 3D data into 2D projected view from current
camera view point within app
J = rotateColorSpace(I);

% Apply polygons drawn on point cloud in app
polyBW = applyPolygons(J,polyBW);

% Combine both masks
BW = sliderBW & polyBW;

```



```
%membalik nilai warna BW
BW1 = imcomplement(BW);

[a1,b1]=size(BW1);
a = 1;
b=1;
rust = 0;
for b=1:b1
    for a = 1:a1
        if BW1(a,b) == 0
            rust = rust + 1;
        else
            rust = rust;
        end
        a=a+1;
    end
    b=b+1;
end
z = string(rust);

persentase = (rust/(a1*b1))*100

%% BASED ON RULES FROM IACS Recommendation 87
if persentase > 20.00
    Re = "POOR"
elseif persentase > 3.00
    Re = "FAIR"
else
    Re = "GOOD"
end

n_fn = length(filename);
new_fn = filename(1,1:n_fn-4);
```

```

new_fn2 = new_fn + "_" + Re + "_" + persentase +
    "_.jpg";
imwrite(BW1, 'DATA WORD/' + Re + "/" + new_fn2);
T = string(filename);
%delete(T)
%imwrite(BW, 'DATA WORD/GOOD/123.jpg');

string('SELESAI')

```

```

function J = rotateColorSpace(I)

% Translate the data to the mean of the current image
within app
shiftVec = [0.181864 0.096676 0.742684];
I = I - shiftVec;
I = [I ones(size(I,1),1)]';

% Apply transformation matrix
tMat = [0.987537 0.813953 0.000000 -0.673717;
        -0.002873 0.008873 0.770919 -0.501201;
        -0.510184 1.575477 -0.004342 7.842296;
        0.000000 0.000000 0.000000 1.000000];

J = (tMat*I)';
end

```

```

function polyBW = applyPolygons(J, polyBW)

% Define each manually generated ROI
hPoints(1).data = [-0.708173 -0.208816;
                  -0.755325 -0.913486;
                  -0.000907 -0.886140;
                  -0.000907 -0.187781];

```

```
% Iteratively apply each ROI
for ii = 1:length(hPoints)
    if size(hPoints(ii).data,1) > 2
        in =
inpolygon(J(:,1),J(:,2),hPoints(ii).data(:,1),hPoints(i
i).data(:,2));
        in = reshape(in,size(polyBW));
        polyBW = polyBW | in;
    end
end
end
end
```

- **Code Program TrainingNetwork (CNN)**

```

clear
clc
delete net.mat

%INPUT DATASET
imds = imageDatastore('data',...
    'IncludeSubfolders',true,...
    'LabelSource','foldernames');

% PROSES AUGMENTASI DATA
augmenter = imageDataAugmenter( ...
    'RandXReflection',true, ...
    'RandRotation',[-180 180],...
    'RandXScale',[1 4], ...
    'RandYReflection',true, ...
    'RandYScale',[1 4]);

% MEMISAHKAN DATA TRAINING DAN DATA UJI
[imdsTrain,imdsTest] =
splitEachLabel(imds,0.8,'randomize');

% UKURAN CITRA
imageSize = [64 64 3];
datastore = augmentedImageDatastore...

(imageSize,imdsTrain,'DataAugmentation',augmenter);

%LAYER CNN
layers = [ ...
    imageInputLayer(imageSize,'Name','input')
    convolution2dLayer(3,8,'Padding','same')
    batchNormalizationLayer
    reluLayer
    maxPooling2dLayer(2,'Stride',2)

```

```
convolution2dLayer(3,16,'Padding','same')
batchNormalizationLayer
reluLayer
maxPooling2dLayer(2,'Stride',2)
convolution2dLayer(3,32,'Padding','same')
batchNormalizationLayer
reluLayer
maxPooling2dLayer(2,'Stride',2)
convolution2dLayer(3,64,'Padding','same')
batchNormalizationLayer
reluLayer
fullyConnectedLayer(32)
reluLayer
fullyConnectedLayer(16)
reluLayer
fullyConnectedLayer(8)
fullyConnectedLayer(6)
fullyConnectedLayer(3)
softmaxLayer
classificationLayer ];

%PARAMETER-PARAMETER TRAINING DATA
options = trainingOptions('sgdm', ...
    'MaxEpochs',1500,...
    'InitialLearnRate',0.01, ...
    'Verbose',true, ...
    'Plots','training-progress');

%PROSES TRAINING DATA
net = trainNetwork(datastore, layers, options);
%
analyzeNetwork(net)
```

```
%%TESTING DATA, VALIDASI DATA
imdsTest_rsz = augmentedImageDatastore...

(imageSize,imdsTest,'DataAugmentation',augmenter)
YPred = classify(net,imdsTest_rsz);

figure
idx = randperm(length(imdsTest_rsz.Files),25);
for i = 1:25
    subplot(5,5,i);
    I = readimage(imdsTest,idx(i));
    label = YPred(idx(i));
    imshow(I)
    title(char(label))
end
save net
```

- **Code Program Desain GUI**

```

function varargout = class_gui(varargin)
% CLASS_GUI MATLAB code for class_gui.fig
%     CLASS_GUI, by itself, creates a new CLASS_GUI or
raises the existing
%     singleton*.
%
%     H = CLASS_GUI returns the handle to a new
CLASS_GUI or the handle to
%     the existing singleton*.
%
%
CLASS_GUI('CALLBACK', hObject, eventData, handles,...)
calls the local
%     function named CALLBACK in CLASS_GUI.M with the
given input arguments.
%
%     CLASS_GUI('Property','Value',...) creates a new
CLASS_GUI or raises the
%     existing singleton*. Starting from the left,
property value pairs are
%     applied to the GUI before class_gui_OpeningFcn
gets called. An
%     unrecognized property name or invalid value
makes property application
%     stop. All inputs are passed to
class_gui_OpeningFcn via varargin.
%
%     *See GUI Options on GUIDE's Tools menu. Choose
"GUI allows only one
%     instance to run (singleton)".
%
% See also: GUIDE, GUIDATA, GUIHANDLES

```

```

% Edit the above text to modify the response to help
class_gui

% Last Modified by GUIDE v2.5 16-Jan-2022 01:23:54

% Begin initialization code - DO NOT EDIT
gui_Singleton = 1;
gui_State = struct('gui_Name',       mfilename, ...
                  'gui_Singleton',  gui_Singleton, ...
                  'gui_OpeningFcn',
@class_gui_OpeningFcn, ...
                  'gui_OutputFcn',
@class_gui_OutputFcn, ...
                  'gui_LayoutFcn',  [] , ...
                  'gui_Callback',   []);
if nargin && ischar(varargin{1})
    gui_State.gui_Callback = str2func(varargin{1});
end

if nargout
    [varargout{1:nargout}] = gui_mainfcn(gui_State,
varargin{:});
else
    gui_mainfcn(gui_State, varargin{:});
end
% End initialization code - DO NOT EDIT

% --- Executes just before class_gui is made visible.
function class_gui_OpeningFcn(hObject, eventdata,
handles, varargin)
% This function has no output args, see OutputFcn.
% hObject    handle to figure
% eventdata  reserved - to be defined in a future
version of MATLAB

```



```
% handles      structure with handles and user data (see
GUIDATA)
% varargin     command line arguments to class_gui (see
VARARGIN)

% Choose default command line output for class_gui
handles.output = hObject;

% Update handles structure
guidata(hObject, handles);

% UIWAIT makes class_gui wait for user response (see
UIRESUME)
% uiwait(handles.figure1);

% --- Outputs from this function are returned to the
command line.
function varargout = class_gui_OutputFcn(hObject,
eventdata, handles)
% varargout    cell array for returning output args (see
VARARGOUT);
% hObject     handle to figure
% eventdata   reserved - to be defined in a future
version of MATLAB
% handles     structure with handles and user data (see
GUIDATA)

% Get default command line output from handles
structure
varargout{1} = handles.output;

% --- Executes on button press in pushbutton1.
```

```
function pushbutton1_Callback(hObject, eventdata,
handles)
% hObject    handle to pushbutton1 (see GCBO)
% eventdata  reserved - to be defined in a future
version of MATLAB
% handles    structure with handles and user data (see
GUIDATA)
clc,clear,close all;
warning off

%% Uploading input image
[filename,pathname] = uigetfile('*..*','Choose the input
image');
im = imread([pathname,filename]);      % merubah image
menjadi nilai numerik
% set the image size to suitable value
scale = 600/(max(size(im(:,:,1))));
im = imresize(im,scale*size(im(:,:,1)));
% % Image resize
[~,n,~] = size(im);

I = rgb2hsv(im);

% Define thresholds for channel 1 based on histogram
settings
channel1Min = 0.000;
channel1Max = 1.000;

% Define thresholds for channel 2 based on histogram
settings
channel2Min = 0.000;
channel2Max = 1.000;

% Define thresholds for channel 3 based on histogram
settings
```

```

channel3Min = 0.000;
channel3Max = 1.000;

% Create mask based on chosen histogram thresholds
sliderBW = (I(:,:,1) >= channel1Min ) & (I(:,:,1) <=
channel1Max) & ...
    (I(:,:,2) >= channel2Min ) & (I(:,:,2) <=
channel2Max) & ...
    (I(:,:,3) >= channel3Min ) & (I(:,:,3) <=
channel3Max);

% Create mask based on selected regions of interest on
point cloud projection
I = double(I);
[m,n,~] = size(I);
polyBW = false([m,n]);
I = reshape(I, [m*n 3]);

% Convert HSV color space to canonical coordinates
Xcoord = I(:,2).*I(:,3).*cos(2*pi*I(:,1));
Ycoord = I(:,2).*I(:,3).*sin(2*pi*I(:,1));
I(:,1) = Xcoord;
I(:,2) = Ycoord;
clear Xcoord Ycoord

% Project 3D data into 2D projected view from current
camera view point within app
% J = rotateColorSpace(I);
% Translate the data to the mean of the current image
within app

shiftVec = [0.214497 0.110897 0.554577];
I = I - shiftVec;
I = [I ones(size(I,1),1)]';

```

```

% Apply transformation matrix
tMat = [-1.469867 -0.968625 0.000000 0.657518;
        -0.115711 0.406616 1.103826 -0.538958;
        0.625913 -2.199506 0.204061 8.590879;
        0.000000 0.000000 0.000000 1.000000];

J = (tMat*I)';

% Apply polygons drawn on point cloud in app
%polyBW = applyPolygons(J,polyBW);

% Define each manually generated ROI
hPoints(1).data = [0.004425 -0.046427;
                  0.799111 -0.059187;
                  0.793801 -0.972793;
                  0.025664 -0.960033];

% Iteratively apply each ROI
for ii = 1:length(hPoints)
    if size(hPoints(ii).data,1) > 2
        in =
inpolygon(J(:,1),J(:,2),hPoints(ii).data(:,1),hPoints(i
i).data(:,2));
        in = reshape(in,size(polyBW));
        polyBW = polyBW | in;
    end
end

% Combine both masks
BW = sliderBW & polyBW;

a = 1;
rust = 0;
for a=1:length(BW)

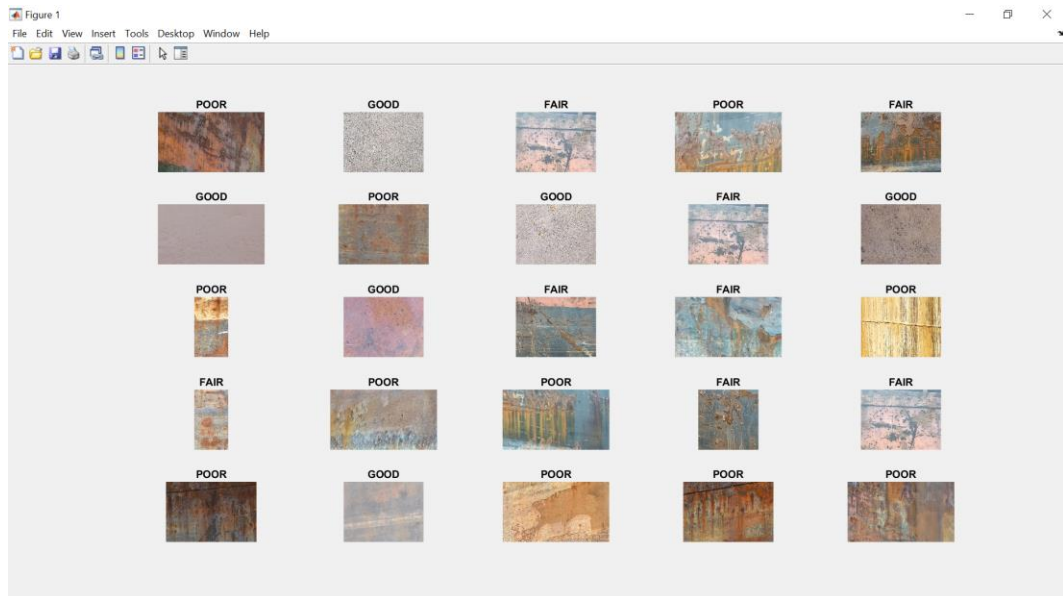
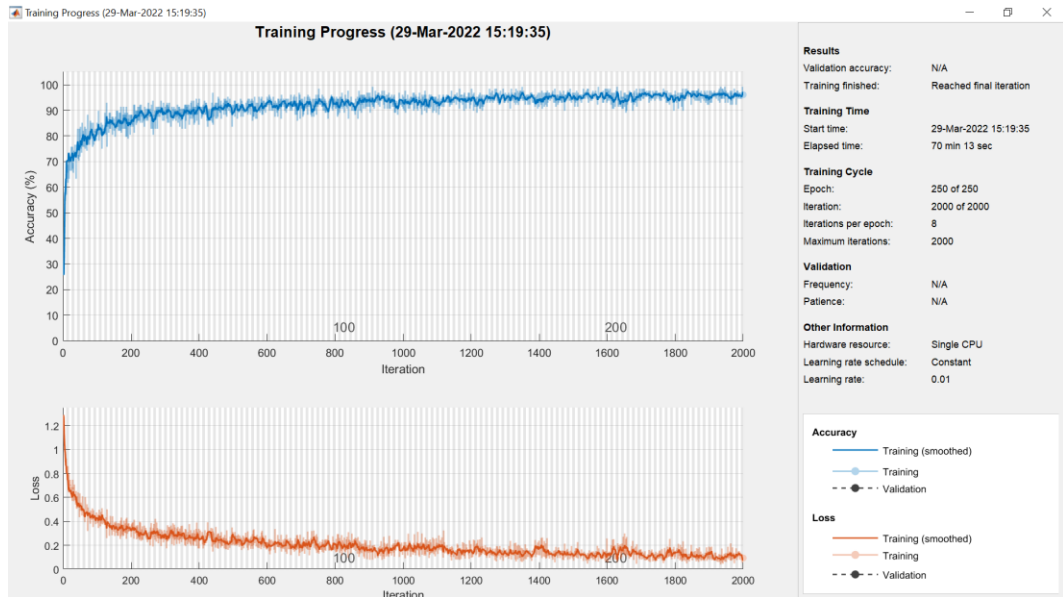
```

```
    if BW(a)==1
        rust = rust + 1;
    else
        rust = rust;
    end
    a=a+1;
end
percentase = rust/length(BW)*100;
%% BASED ON RULES FROM IACS Recomendation 87
if percentase > 20.00
    CLASS = "POOR"
elseif percentase > 3.00
    CLASS = "FAIR"
else
    CLASS = "GOOD"
end
percentase = percentase + "%"
n_fn = length(filename);
new_fn = filename(1,1:n_fn-4);
new_fn2 = i + "_" + CLASS + "_" + percentase + "_.jpg";
imwrite(im, 'data/' + CLASS + "/" + new_fn2);
% imwrite(im, i + "_.jpg");
T = string(filename);
% delete(T)

string('SELESAI')

run AppDETECT.m
```

- Screenshot Proxes Training Network (Epoch 250)



Deep Learning Network Analyzer

net
Analysis date: 10-Dec-2021 21:46:04

25 layers, 0 warnings, 0 errors

Name	Type	Activations	Learnables
1 input	Image Input	32x32x3	-
2 conv_1	Convolution	32x32x8	Weights 3x3x3x8 Bias 1x1x8
3 batchnorm_1	Batch Normalization	32x32x8	Offset 1x1x8 Scale 1x1x8
4 relu_1	ReLU	32x32x8	-
5 maxpool_1	Max Pooling	16x16x8	-
6 conv_2	Convolution	16x16x16	Weights 3x3x8x16 Bias 1x1x16
7 batchnorm_2	Batch Normalization	16x16x16	Offset 1x1x16 Scale 1x1x16
8 relu_2	ReLU	16x16x16	-
9 maxpool_2	Max Pooling	8x8x16	-
10 conv_3	Convolution	8x8x32	Weights 3x3x16x32 Bias 1x1x32
11 batchnorm_3	Batch Normalization	8x8x32	Offset 1x1x32 Scale 1x1x32
12 relu_3	ReLU	8x8x32	-
13 maxpool_3	Max Pooling	4x4x32	-
14 conv_4	Convolution	4x4x64	Weights 3x3x32x64 Bias 1x1x64
15 batchnorm_4	Batch Normalization	4x4x64	Offset 1x1x64 Scale 1x1x64

MATLAB R2019b

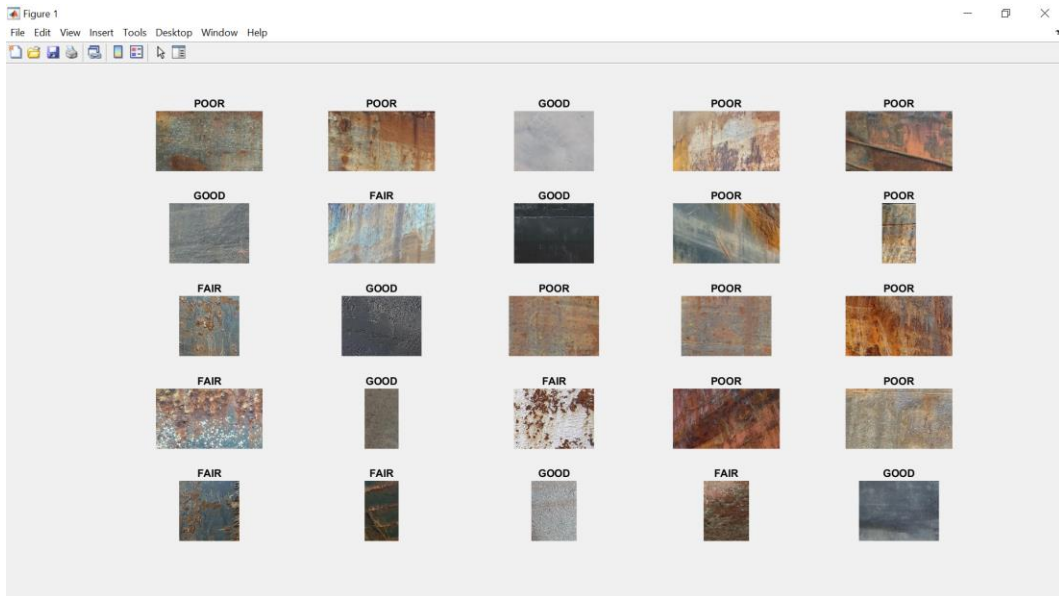
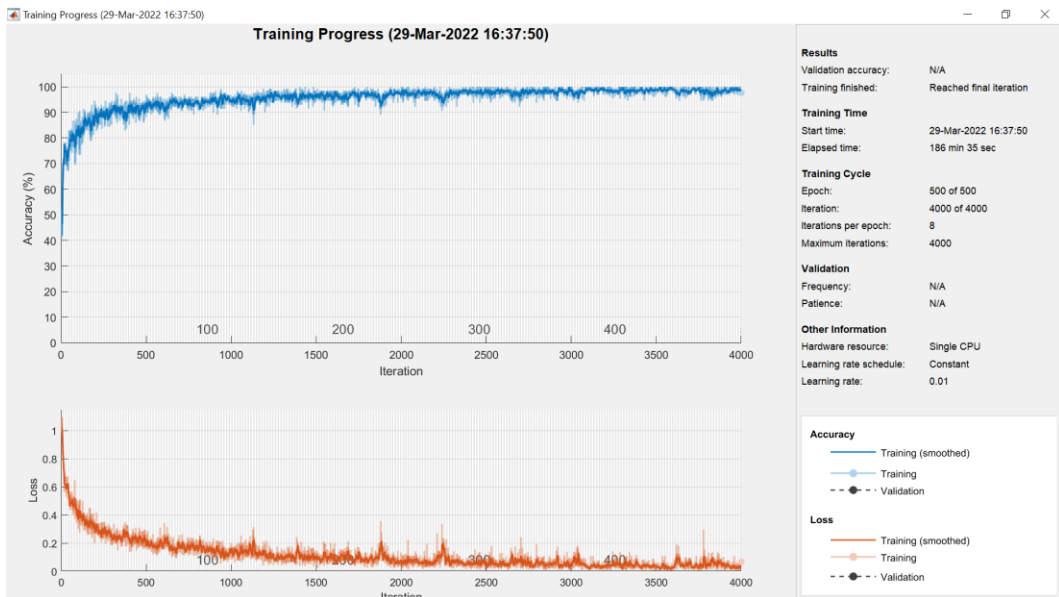
Command Window

```

113 | 900 | 00:34:52 | 95.31% | 0.1936 | 0.0100 |
119 | 950 | 00:36:32 | 96.09% | 0.1063 | 0.0100 |
125 | 1000 | 00:38:16 | 90.63% | 0.1678 | 0.0100 |
132 | 1050 | 00:40:05 | 92.97% | 0.1661 | 0.0100 |
138 | 1100 | 00:41:41 | 91.41% | 0.1904 | 0.0100 |
144 | 1150 | 00:43:16 | 92.97% | 0.1385 | 0.0100 |
150 | 1200 | 00:45:22 | 95.31% | 0.1088 | 0.0100 |
157 | 1250 | 00:46:57 | 94.53% | 0.1562 | 0.0100 |
163 | 1300 | 00:48:48 | 96.88% | 0.1038 | 0.0100 |
169 | 1350 | 00:51:17 | 99.22% | 0.0605 | 0.0100 |
175 | 1400 | 00:52:36 | 92.19% | 0.2194 | 0.0100 |
182 | 1450 | 00:53:48 | 93.75% | 0.1376 | 0.0100 |
188 | 1500 | 00:55:06 | 94.53% | 0.1241 | 0.0100 |
194 | 1550 | 00:56:25 | 96.09% | 0.1242 | 0.0100 |
200 | 1600 | 00:57:40 | 94.53% | 0.1470 | 0.0100 |
207 | 1650 | 00:59:10 | 96.88% | 0.1228 | 0.0100 |
213 | 1700 | 01:00:40 | 96.88% | 0.1131 | 0.0100 |
219 | 1750 | 01:02:11 | 96.88% | 0.0063 | 0.0100 |
225 | 1800 | 01:03:47 | 97.66% | 0.1036 | 0.0100 |
232 | 1850 | 01:05:24 | 95.31% | 0.0831 | 0.0100 |
238 | 1900 | 01:06:57 | 95.31% | 0.1585 | 0.0100 |
244 | 1950 | 01:08:37 | 96.88% | 0.0605 | 0.0100 |
250 | 2000 | 01:10:13 | 96.09% | 0.0935 | 0.0100 |
    
```

augmentedImage datastore with properties:

• Screenshot Proxes Training Network (Epoch 500)



Deep Learning Network Analyzer

net
Analysis date: 10-Dec-2021 23:26:18

25 layers 0 warnings 0 errors

Name	Type	Activations	Learnables
1 input	Image Input	32x32x3	-
2 conv_1	Convolution	32x32x8	Weights 3x3x3x8 Bias 1x1x8
3 batchnorm_1	Batch Normalization	32x32x8	Offset 1x1x8 Scale 1x1x8
4 relu_1	ReLU	32x32x8	-
5 maxpool_1	Max Pooling	16x16x8	-
6 conv_2	Convolution	16x16x16	Weights 3x3x8x16 Bias 1x1x16
7 batchnorm_2	Batch Normalization	16x16x16	Offset 1x1x16 Scale 1x1x16
8 relu_2	ReLU	16x16x16	-
9 maxpool_2	Max Pooling	8x8x16	-
10 conv_3	Convolution	8x8x32	Weights 3x3x16x32 Bias 1x1x32
11 batchnorm_3	Batch Normalization	8x8x32	Offset 1x1x32 Scale 1x1x32
12 relu_3	ReLU	8x8x32	-
13 maxpool_3	Max Pooling	4x4x32	-
14 conv_4	Convolution	4x4x64	Weights 3x3x32x64 Bias 1x1x64
15 batchnorm_4	Batch Normalization	4x4x64	Offset 1x1x64 Scale 1x1x64

MATLAB R2019b

train_classifier.m

```

1 clear
2 clc
3 delete net.mat
4
5 %INPUT DATASET
6 imds = imageDatastore('
7     'includeSubfolders'
8     'LabelSource','fol
9
10 % PROSES AUGMENTASI DA
11 augmenter = imageDataA
12     'RandXReflection',t
13     'RandRotation',[-1f
14     'RandScale',[1 4],
15     'RandXReflection',t
16     'RandScale',[1 4])
17
18 % MEMISAHKAN DATA TRAI
19 [imdsTrain,imdsTest] =
20
21 % UKURAN CITRA
22 imageSize = [32 32 3];
23 datastore = augmentedI
24     (imageSize,
25     (imageSize,
26 %LAYER CNN
    
```

Command Window

```

432 | 3400 | 02:14:20 | 97.22% | 0.0259 | 0.0100 |
433 | 3500 | 02:44:11 | 96.88% | 0.0498 | 0.0100 |
444 | 3550 | 02:46:12 | 100.00% | 0.0145 | 0.0100 |
450 | 3600 | 02:48:29 | 98.44% | 0.0423 | 0.0100 |
457 | 3650 | 02:50:48 | 99.22% | 0.0530 | 0.0100 |
463 | 3700 | 02:53:05 | 97.66% | 0.0842 | 0.0100 |
469 | 3750 | 02:55:22 | 98.44% | 0.0434 | 0.0100 |
475 | 3800 | 02:57:39 | 99.22% | 0.0277 | 0.0100 |
482 | 3850 | 03:00:00 | 100.00% | 0.0139 | 0.0100 |
488 | 3900 | 03:02:01 | 98.44% | 0.0348 | 0.0100 |
494 | 3950 | 03:04:19 | 100.00% | 0.0158 | 0.0100 |
500 | 4000 | 03:06:35 | 97.66% | 0.0658 | 0.0100 |
    
```

imdsTest_rsiz =
augmentedImageDatastore with properties:
NumObservations: 257
Files: (257x1 cell)
AlternateFileSystemRoots: {}
MiniBatchSize: 128
DataAugmentation: [1x1 imageDataAugmenter]
ColorPreprocessing: 'none'
OutputSize: [32 32]
OutputSizeMode: 'resize'
DispatchInBackground: 0

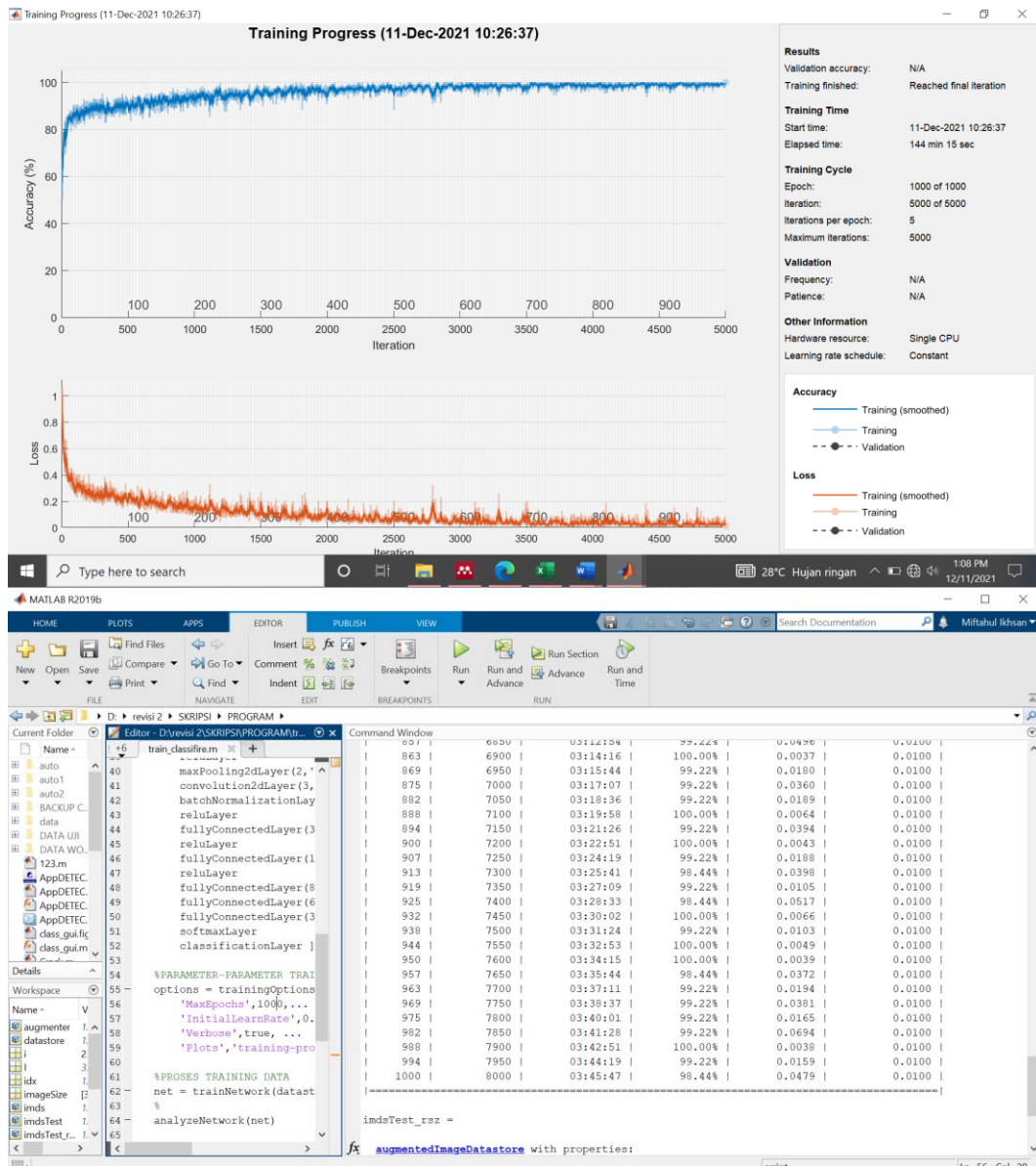
• Screenshot Proses Training Network (Epoch 1000)

The screenshot displays a training process at Epoch 1000. The top portion shows a grid of 20 images, each labeled with its predicted class: 'BERAT' or 'NON KOROSI'. The labels are as follows:

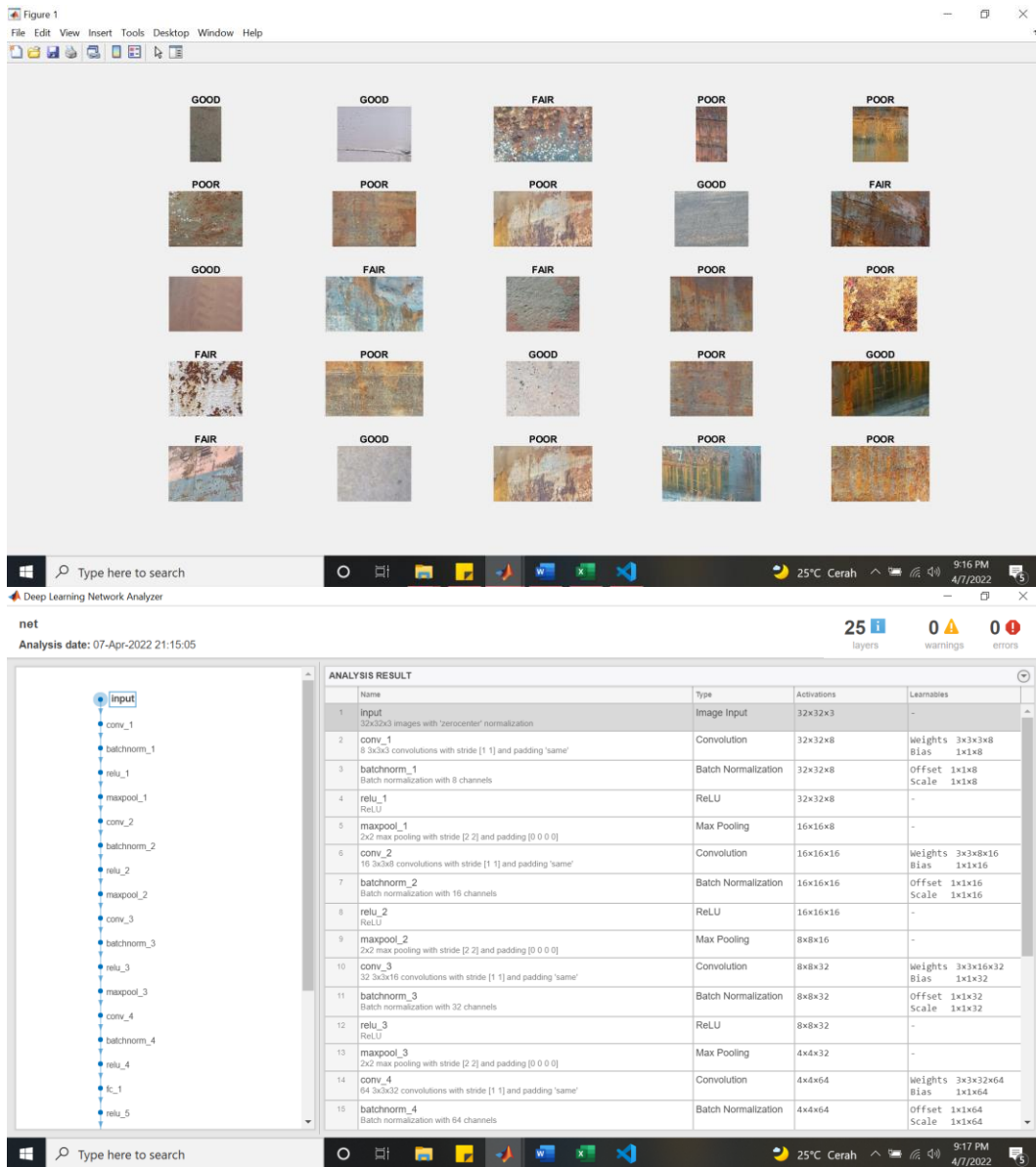
- Row 1: BERAT, NON KOROSI, BERAT, NON KOROSI, NON KOROSI
- Row 2: SEDANG, BERAT, BERAT, NON KOROSI, SEDANG
- Row 3: NON KOROSI, SEDANG, NON KOROSI, NON KOROSI, NON KOROSI
- Row 4: BERAT, NON KOROSI, NON KOROSI, NON KOROSI, BERAT
- Row 5: NON KOROSI, NON KOROSI, NON KOROSI, NON KOROSI, BERAT

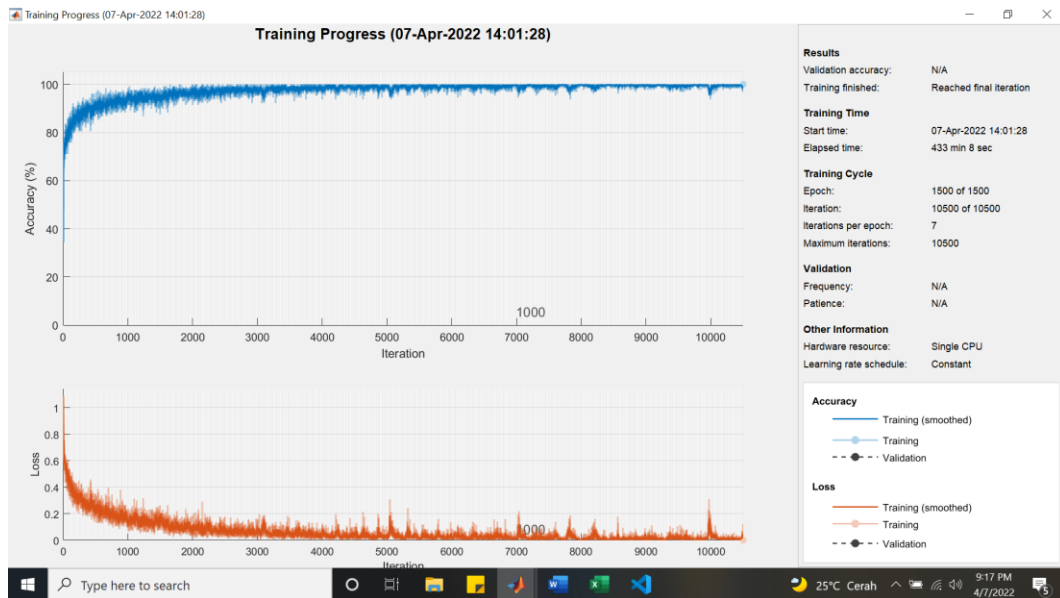
The bottom portion shows the 'Deep Learning Network Analyzer' window. The analysis date is 11-Dec-2021 12:51:06. The network architecture is visualized on the left, and the 'ANALYSIS RESULT' table is shown on the right:

Name	Type	Activations	Learnables
1 input 32x32x3 images with 'zerocenter' normalization	Image Input	32x32x3	-
2 conv_1 8 3x3x3 convolutions with stride [1 1] and padding 'same'	Convolution	32x32x8	Weights 3x3x3x8 Bias 1x1x8
3 batchnorm_1 Batch normalization with 8 channels	Batch Normalization	32x32x8	Offset 1x1x8 Scale 1x1x8
4 relu_1 ReLU	ReLU	32x32x8	-
5 maxpool_1 2x2 max pooling with stride [2 2] and padding [0 0 0 0]	Max Pooling	16x16x8	-
6 conv_2 16 3x3x8 convolutions with stride [1 1] and padding 'same'	Convolution	16x16x16	Weights 3x3x8x16 Bias 1x1x16
7 batchnorm_2 Batch normalization with 16 channels	Batch Normalization	16x16x16	Offset 1x1x16 Scale 1x1x16
8 relu_2 ReLU	ReLU	16x16x16	-
9 maxpool_2 2x2 max pooling with stride [2 2] and padding [0 0 0 0]	Max Pooling	8x8x16	-
10 conv_3 32 3x3x16 convolutions with stride [1 1] and padding 'same'	Convolution	8x8x32	Weights 3x3x16x32 Bias 1x1x32
11 batchnorm_3 Batch normalization with 32 channels	Batch Normalization	8x8x32	Offset 1x1x32 Scale 1x1x32
12 relu_3 ReLU	ReLU	8x8x32	-
13 maxpool_3 2x2 max pooling with stride [2 2] and padding [0 0 0 0]	Max Pooling	4x4x32	-
14 conv_4 64 3x3x32 convolutions with stride [1 1] and padding 'same'	Convolution	4x4x64	Weights 3x3x32x64 Bias 1x1x64
15 batchnorm_4 Batch normalization with 64 channels	Batch Normalization	4x4x64	Offset 1x1x64 Scale 1x1x64



• Screenshot Proxes Training Network (Epoch 1500)





MATLAB R2019b

HOME PLOTS APPS EDITOR PUBLISH VIEW

Find Files Find Compare Go To Comment Indent Breakpoints Run Run and Advance Run Section Run and Time

FILE PRINT NAVIGATE EDIT BREAKPOINTS RUN

Current Folder: D:\revisi 2\SKRIPSI\PROGRAM

Editor: D:\revisi 2\SKRIPSI\PROGRAM\train_classfire.m

```

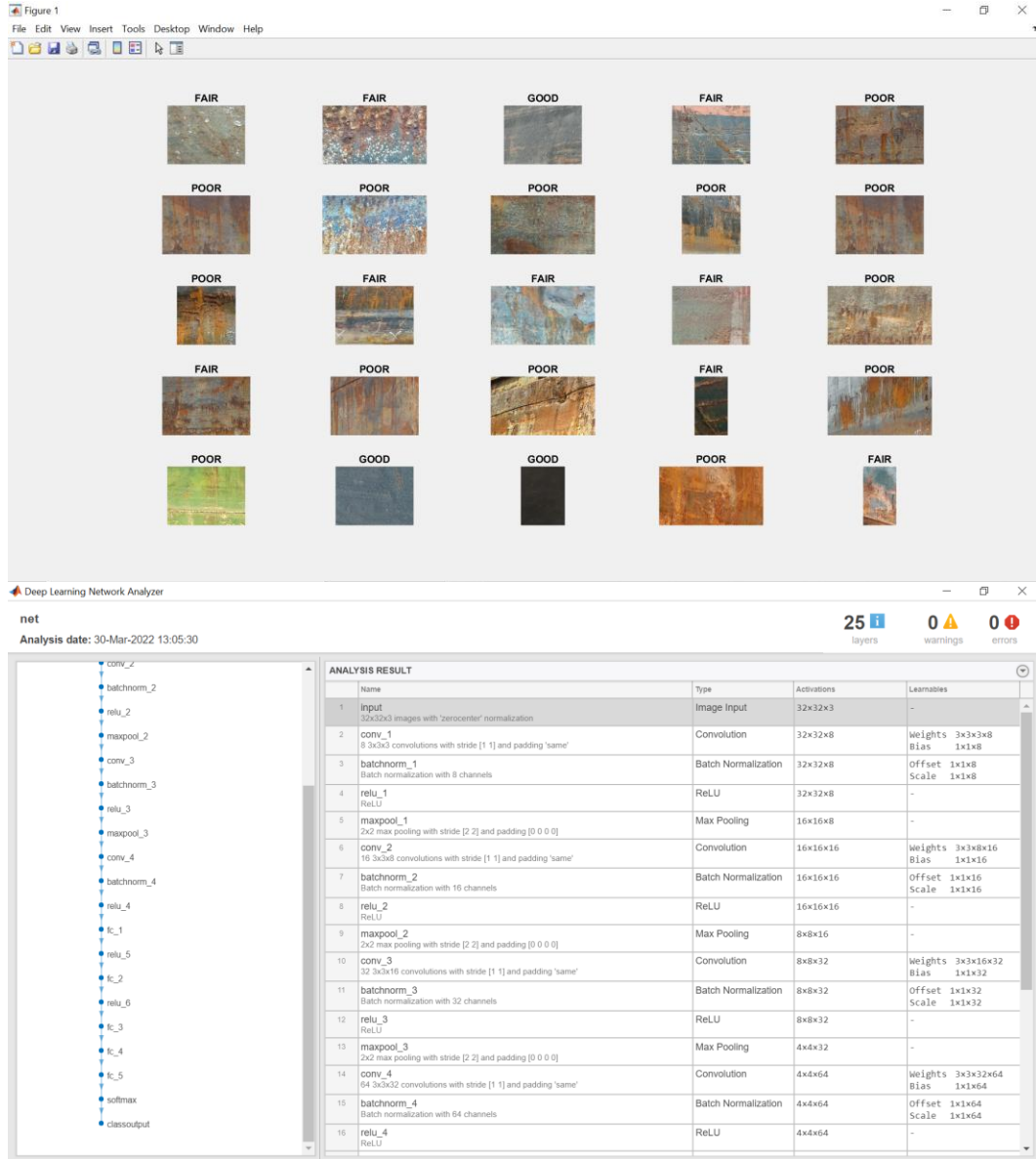
1 clear
2 clc
3 delete net.mat
4
5 %INPUT DATASET
6 imds = imageDatastore('data',...
7 'IncludeSubfolders',true,...
8 'LabelSource','foldernames');
9
10 % PROSES AUGMENTASI DATA
11 augmenter = imageDataAugmenter( ...
12 'RandXReflection',true, ...
13 'RandXScale',[1 4], ...
14 'RandYReflection',true, ...
15 'RandYScale',[1 4]);
16
17 % MEMISAHKAN DATA TRAINING DAN DATA UJI
18 [imdsTrain,imdsTest] = splitEachLabel(imds,0.1);
19
20 % UKURAN CITRA
21 imageSize = [32 32 3];
22
23 datastore = augmentedImageDatastore...
24 (imageSize,imdsTrain,'DataBinamant
    
```

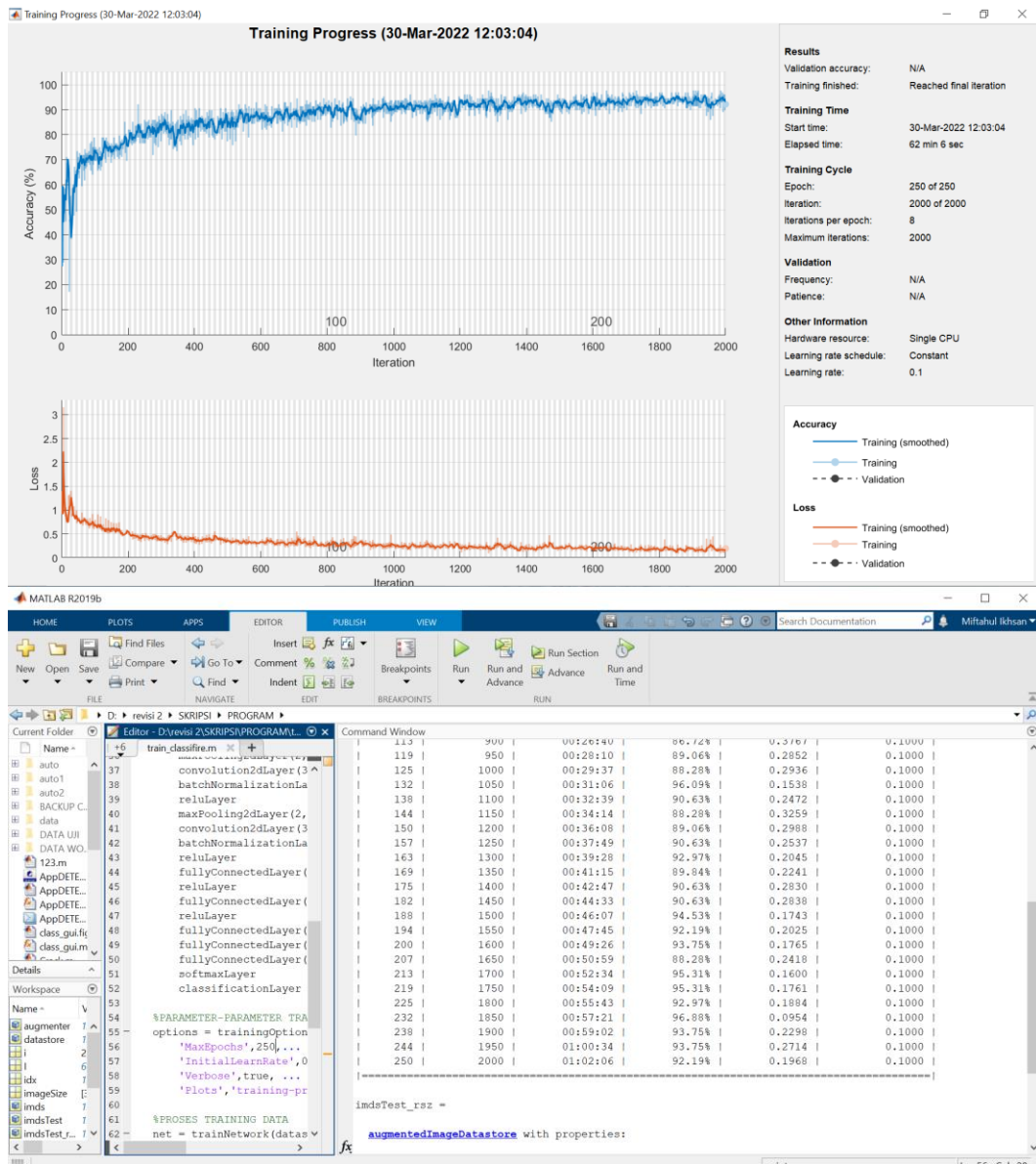
Command Window

1343	9400	06:30:52	99.22%	0.0123
1350	9450	06:32:45	100.00%	0.0013
1358	9500	06:34:43	98.44%	0.0202
1365	9550	06:36:40	99.22%	0.0141
1372	9600	06:38:38	100.00%	0.0100
1379	9650	06:40:25	100.00%	0.0057
1386	9700	06:42:24	100.00%	0.0043
1393	9750	06:44:20	100.00%	0.0016
1400	9800	06:46:13	100.00%	0.0007
1408	9850	06:48:13	99.22%	0.0158
1415	9900	06:50:06	100.00%	0.0007
1422	9950	06:52:04	98.44%	0.0420
1429	10000	06:53:59	98.44%	0.0186
1436	10050	06:55:30	100.00%	0.0149
1443	10100	06:57:12	99.22%	0.0134
1450	10150	06:58:56	100.00%	0.0068
1458	10200	07:00:37	100.00%	0.0053
1465	10250	07:02:24	100.00%	0.0035
1472	10300	07:04:16	100.00%	0.0089
1479	10350	07:06:25	99.22%	0.0175
1486	10400	07:08:35	99.22%	0.0121
1493	10450	07:10:37	100.00%	0.0084
1500	10500	07:13:08	100.00%	0.0008

script |Ln 56 Col 21

• Screenshot Proxes Training Network (Learning Rate 0.1)





- Screenshot Proses Training Network (Learning Rate 0.01)

Figure 1

The screenshot displays a training network process. The top part shows a grid of 25 images, each labeled with a classification: POOR, FAIR, or GOOD. The bottom part shows the Deep Learning Network Analyzer window, which provides details about the network architecture and the analysis results.

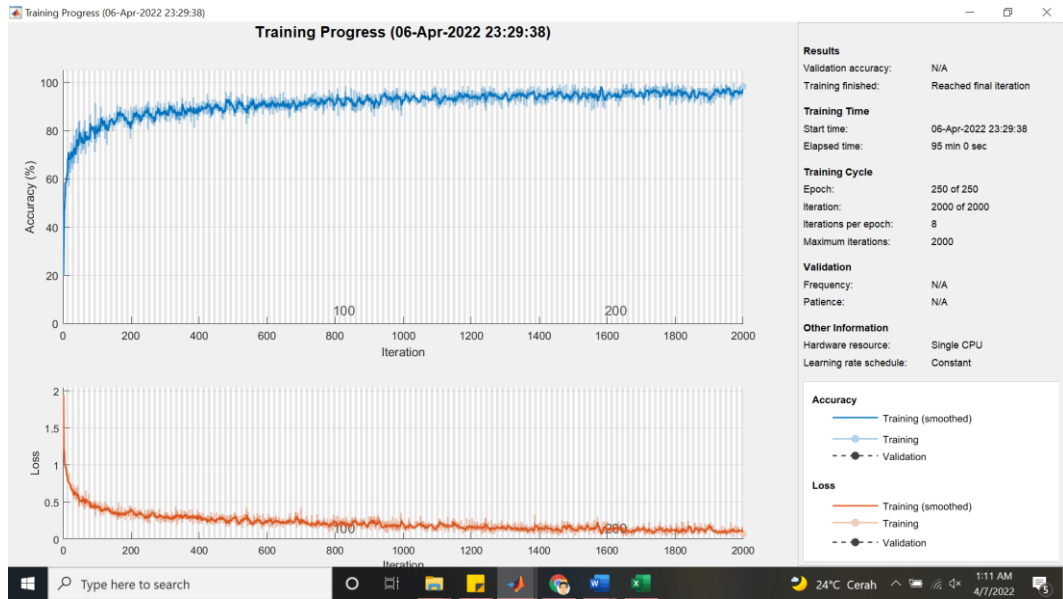
Deep Learning Network Analyzer

net
Analysis date: 07-Apr-2022 01:05:07

25 layers, 0 warnings, 0 errors

ANALYSIS RESULT

Name	Type	Activations	Learnables
1 input 32x32x3 images with 'zerocenter' normalization	Image Input	32x32x3	-
2 conv_1 8 3x3x3 convolutions with stride [1 1] and padding 'same'	Convolution	32x32x8	Weights 3x3x3x8 Bias 1x1x8
3 batchnorm_1 Batch normalization with 8 channels	Batch Normalization	32x32x8	Offset 1x1x8 Scale 1x1x8
4 relu_1 ReLU	ReLU	32x32x8	-
5 maxpool_1 2x2 max pooling with stride [2 2] and padding [0 0 0 0]	Max Pooling	16x16x8	-
6 conv_2 16 3x3x8 convolutions with stride [1 1] and padding 'same'	Convolution	16x16x16	Weights 3x3x8x16 Bias 1x1x16
7 batchnorm_2 Batch normalization with 16 channels	Batch Normalization	16x16x16	Offset 1x1x16 Scale 1x1x16
8 relu_2 ReLU	ReLU	16x16x16	-
9 maxpool_2 2x2 max pooling with stride [2 2] and padding [0 0 0 0]	Max Pooling	8x8x16	-
10 conv_3 32 3x3x16 convolutions with stride [1 1] and padding 'same'	Convolution	8x8x32	Weights 3x3x16x32 Bias 1x1x32
11 batchnorm_3 Batch normalization with 32 channels	Batch Normalization	8x8x32	Offset 1x1x32 Scale 1x1x32
12 relu_3 ReLU	ReLU	8x8x32	-
13 maxpool_3 2x2 max pooling with stride [2 2] and padding [0 0 0 0]	Max Pooling	4x4x32	-
14 conv_4 64 3x3x32 convolutions with stride [1 1] and padding 'same'	Convolution	4x4x64	Weights 3x3x32x64 Bias 1x1x64
15 batchnorm_4 Batch normalization with 64 channels	Batch Normalization	4x4x64	Offset 1x1x64 Scale 1x1x64



MATLAB R2019b

HOME PLOTS APPS EDITOR PUBLISH VIEW

File Edit View Command Window

Current Folder: D:\revisi 2\SKRIPSI\PROGRAM

Workspace:

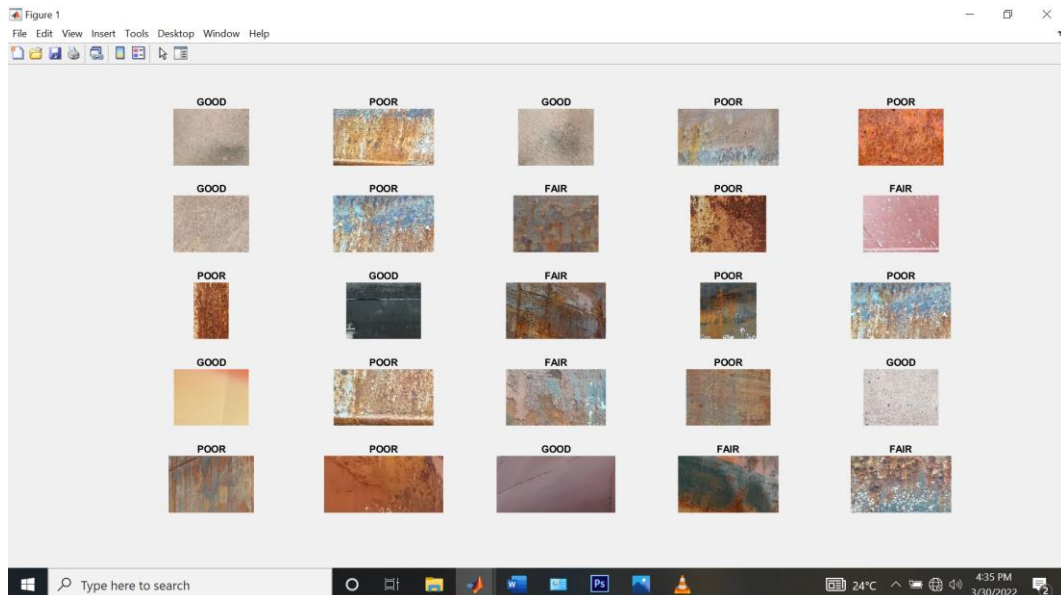
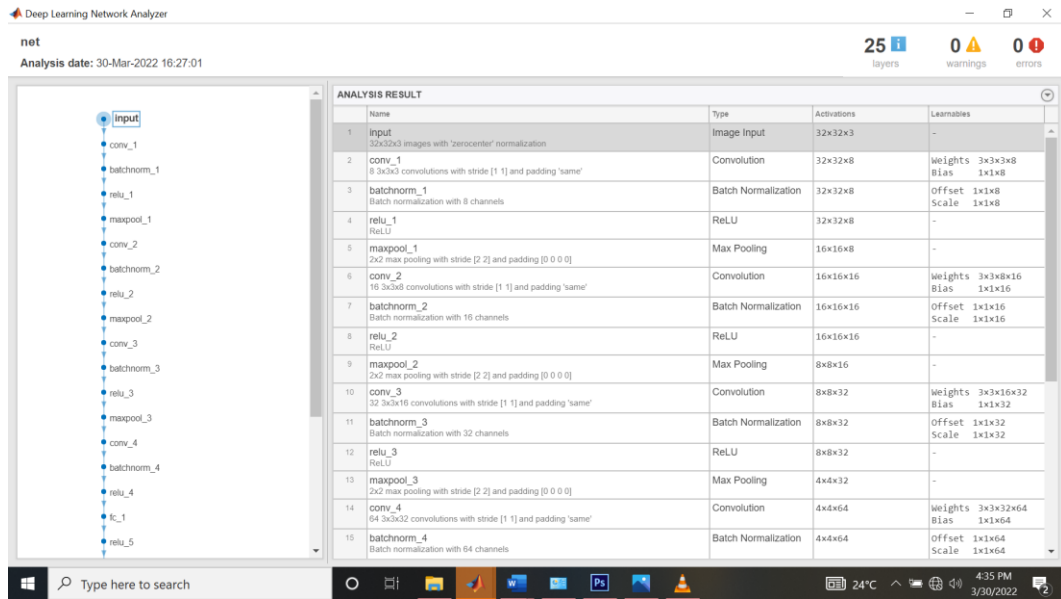
- augmenter 7x1
- datastore 7x1
- i 25
- idx 7x2
- imageSize [32, 32]
- imds 7x1
- imdsTest 7x1

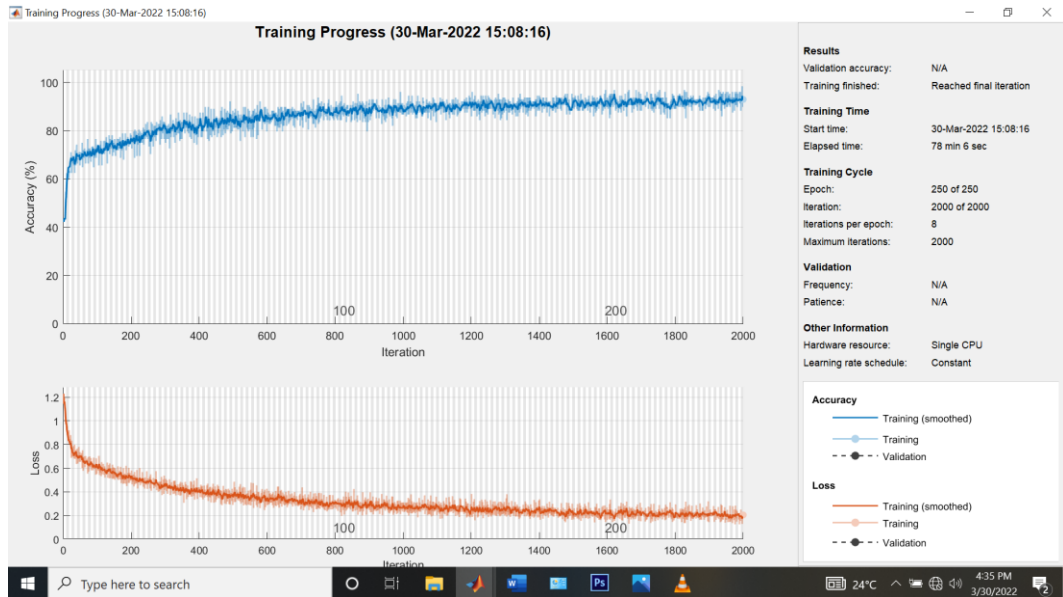
Command Window:

125	1000	00:35:15	92.97%	0.1657	0.0100
132	1050	00:39:59	93.75%	0.1979	0.0100
138	1100	00:43:37	93.75%	0.1589	0.0100
144	1150	00:46:41	96.88%	0.0874	0.0100
150	1200	00:50:29	92.97%	0.1909	0.0100
157	1250	00:53:58	96.09%	0.1433	0.0100
163	1300	00:57:15	96.88%	0.1145	0.0100
169	1350	01:00:36	94.53%	0.1201	0.0100
175	1400	01:04:03	96.09%	0.1487	0.0100
182	1450	01:07:30	96.09%	0.1509	0.0100
188	1500	01:11:13	95.31%	0.1287	0.0100
194	1550	01:14:46	94.53%	0.1615	0.0100
200	1600	01:17:59	92.19%	0.2383	0.0100
207	1650	01:20:07	94.53%	0.1034	0.0100
213	1700	01:21:46	94.53%	0.1554	0.0100
219	1750	01:23:43	93.75%	0.1824	0.0100
225	1800	01:25:43	96.09%	0.0953	0.0100
232	1850	01:28:52	97.66%	0.0885	0.0100
238	1900	01:31:13	92.97%	0.1952	0.0100
244	1950	01:33:18	96.88%	0.0891	0.0100
250	2000	01:35:00	98.44%	0.0695	0.0100

augmenter = augmentedImageDatastore with properties:

- Screenshot Proxes Training Network (Learning Rate 0.001)





MATLAB R2019b

The screenshot shows the MATLAB R2019b interface. The Editor window displays the following neural network architecture:

```

convolution2dLayer(3,32,'Padding','same')
batchNormalizationLayer
reluLayer
maxPooling2dLayer(2,'Stride',2)
convolution2dLayer(3,64,'Padding','same')
batchNormalizationLayer
reluLayer
fullyConnectedLayer(32)
reluLayer
fullyConnectedLayer(16)
reluLayer
fullyConnectedLayer(8)
fullyConnectedLayer(6)
fullyConnectedLayer(3)
softmaxLayer
classificationLayer ]
    
```

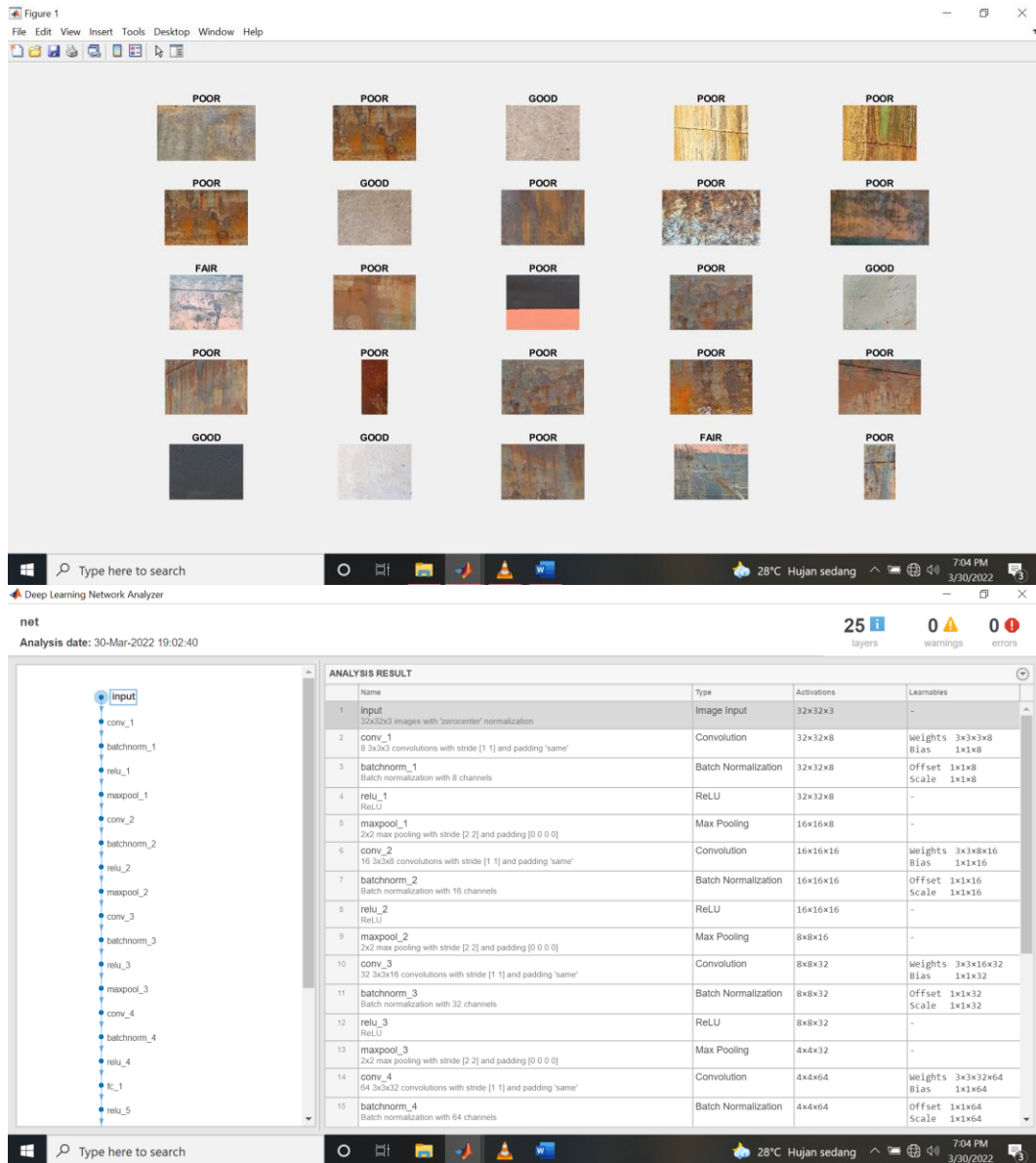
The Command Window displays the following table of training metrics:

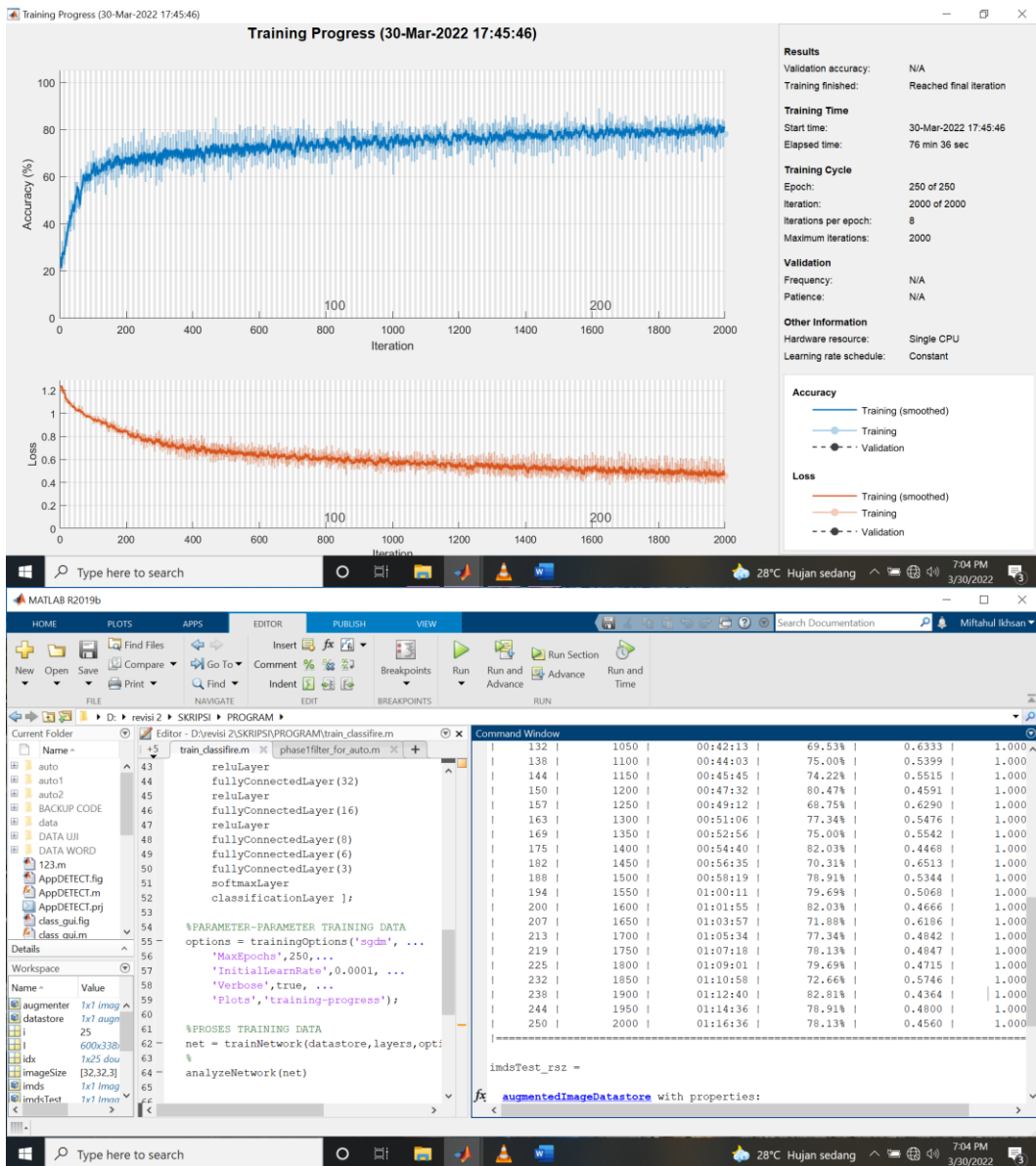
132	1050	00:33:43	89.06%	0.2798	0
138	1100	00:35:19	89.06%	0.2769	0
144	1150	00:37:03	86.72%	0.3234	0
150	1200	00:38:48	89.84%	0.2036	0
157	1250	00:40:31	94.53%	0.1935	0
163	1300	00:42:11	91.41%	0.2499	0
169	1350	00:43:41	92.19%	0.2437	0
175	1400	00:46:04	94.53%	0.1843	0
182	1450	00:48:21	94.53%	0.2259	0
188	1500	00:51:29	91.41%	0.2243	0
194	1550	00:53:46	87.50%	0.2516	0
200	1600	00:56:17	93.75%	0.1614	0
207	1650	00:58:52	92.97%	0.2390	0
213	1700	01:01:11	91.41%	0.2285	0
219	1750	01:03:47	92.97%	0.1863	0
225	1800	01:06:21	92.19%	0.1796	0
232	1850	01:09:08	93.75%	0.1800	0
238	1900	01:12:13	92.97%	0.2104	0
244	1950	01:15:09	96.09%	0.1638	0
250	2000	01:18:06	92.97%	0.2049	0

The Workspace window shows the following variables:

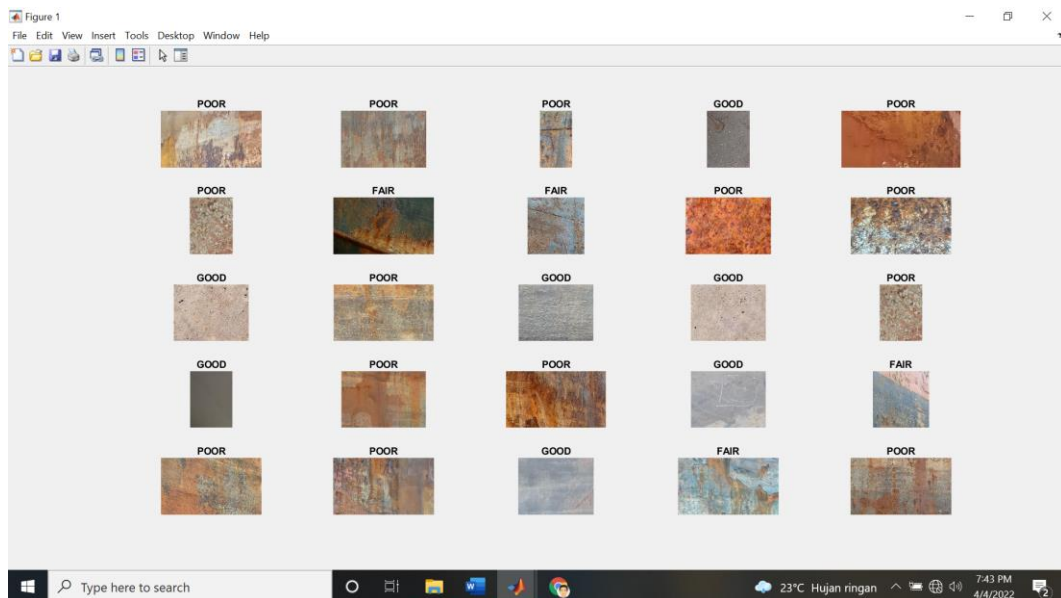
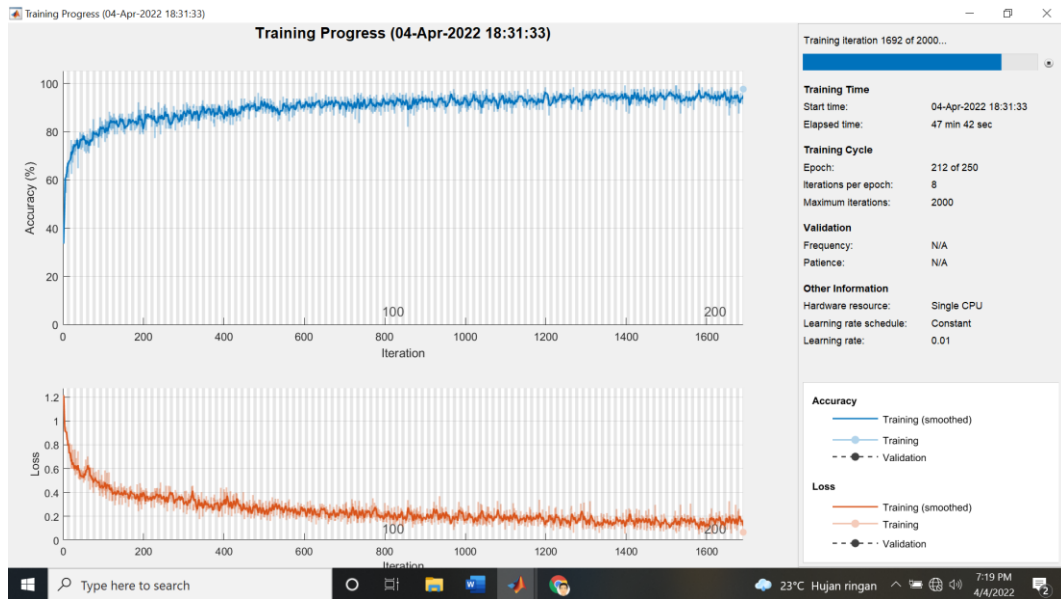
- augmenter: 1x1 imag
- datastore: 1x1 augm
- i: 25
- i: 339x600
- idx: 1x25 dou
- imageSize: [32,32,3]
- imds: 1x1 imag
- imdsTest: 1x1 imag

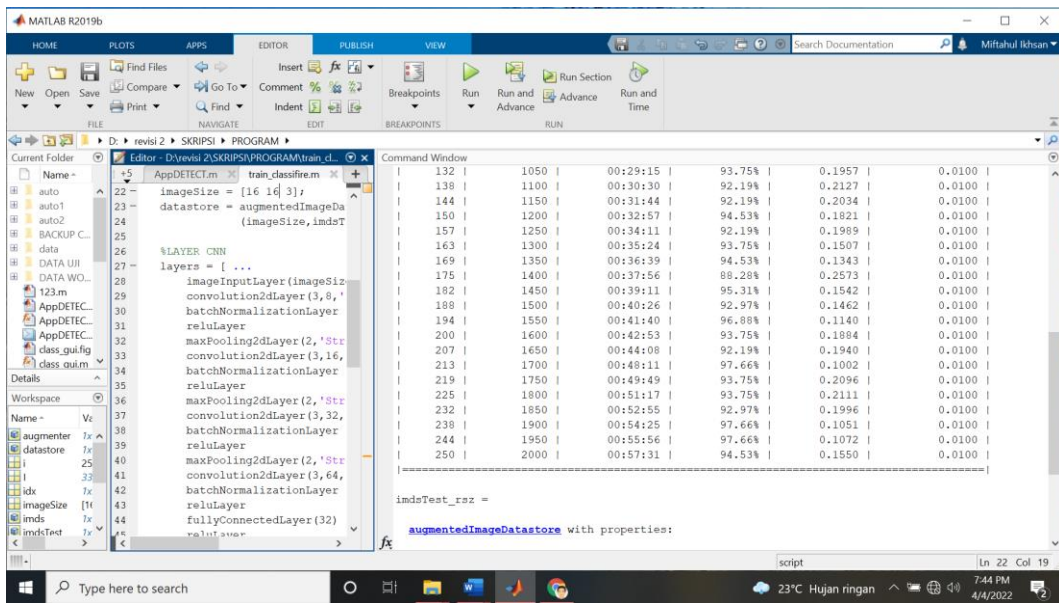
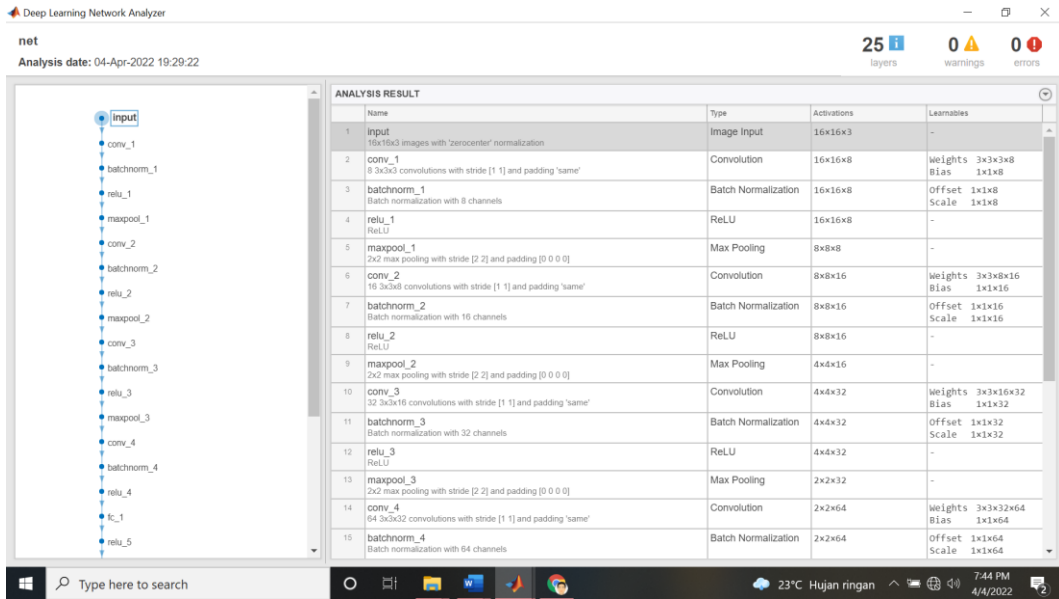
- Screenshot Proses Training Network (Learning Rate 0.0001)



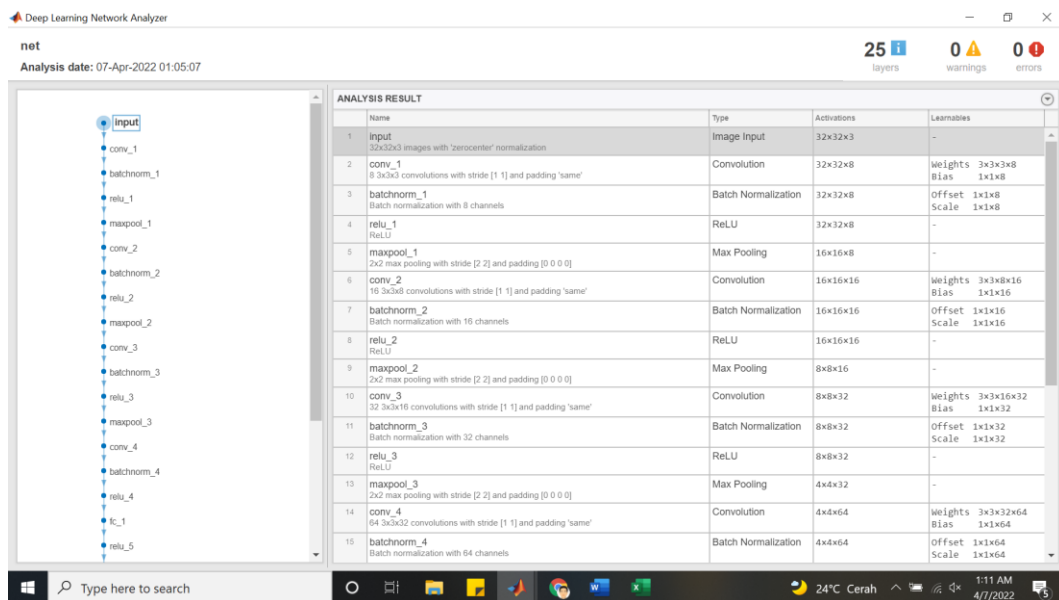
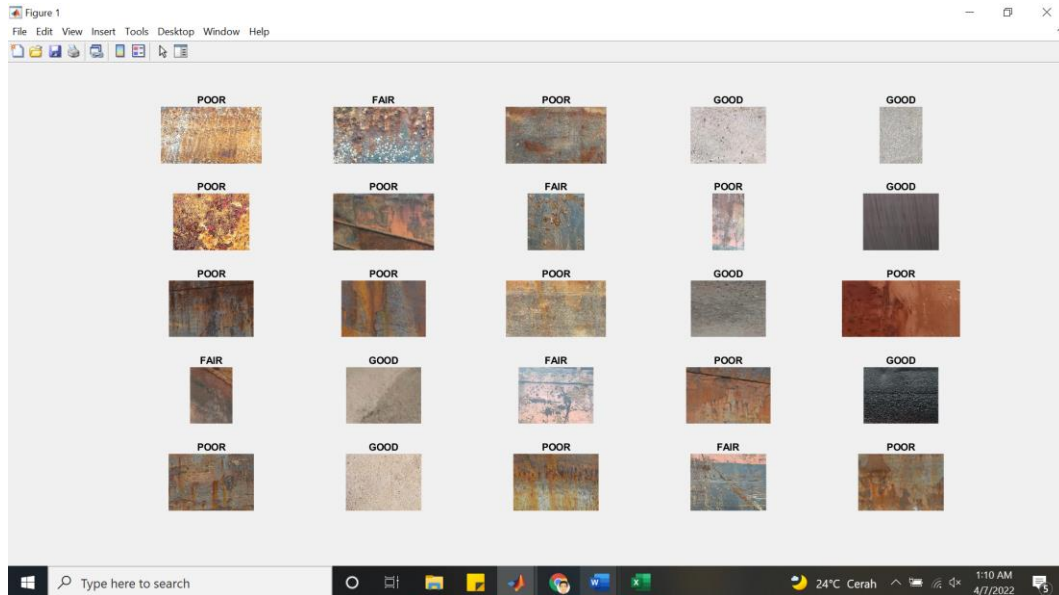


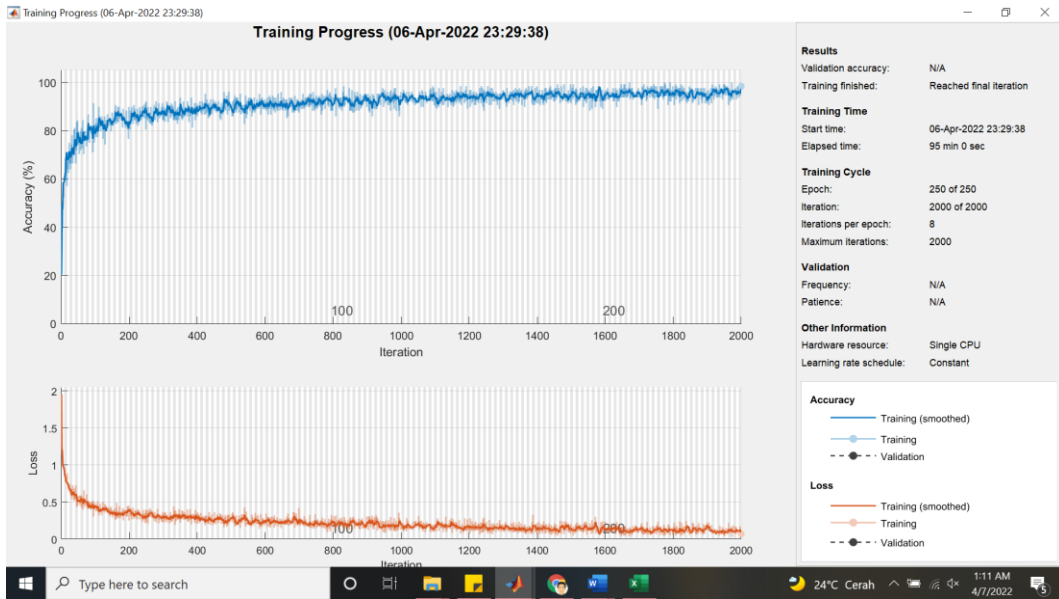
- Screenshot Proses Training Network (Image Size 16x16x3)





- Screenshot Proxes Training Network (Image Size 32x32x3)





MATLAB R2019b

Current Folder: D:\revisi 2\SKRIPSI\PROGRAM

Editor: D:\revisi 2\SKRIPSI\PROGRAM\train_class.m

```

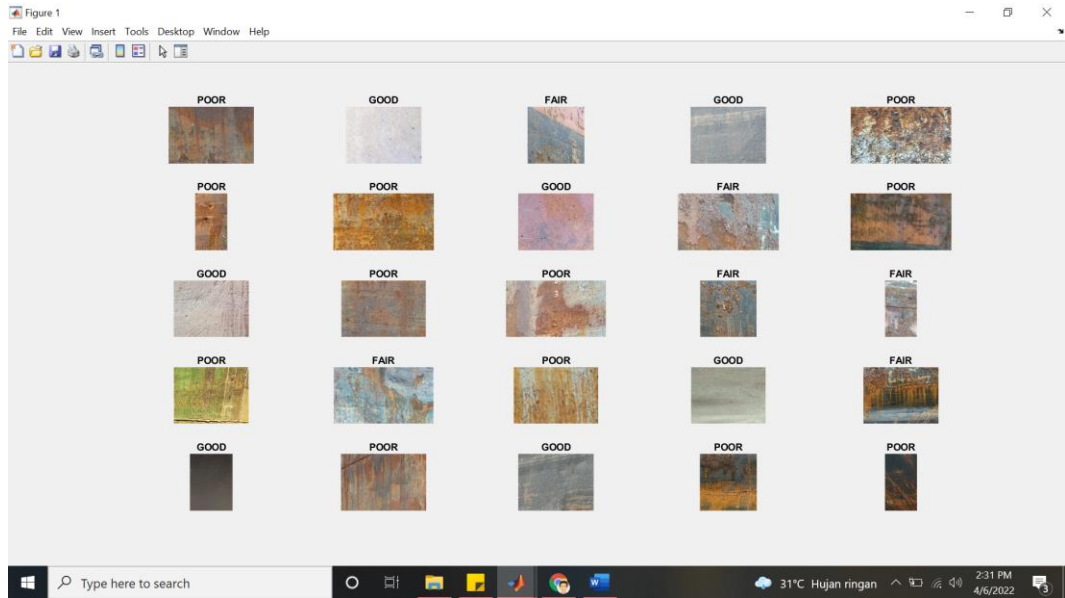
10 % FROSES AUGMENTASI DATA
11 augmenter = imageDataAugmenter
12     'RandXReflection',true, ...
13     'RandRotation',[-180 180],
14     'RandXScale',[1 4], ...
15     'RandYReflection',true, ...
16     'RandYScale',[1 4]);
17
18 % MEMISAHKAN DATA TRAINING DAN
19 [imdsTrain,imdsTest] = splitEa
20
21 % UKURAN CITRA
22 imageSize = [32 32 3];
23 datastore = augmentedImageData
24     (imageSize,imdsTra
25
26 %LAYER CNN
27 layers = [ ...
28     imageInputLayer(imageSize,
29     convolution2dLayer(3,8,'Pa
30     batchNormalizationLayer
31     reluLayer
32     maxPooling2dLayer(2,'Strid
33     convolution2dLayer(3,16,'Pa
    
```

Command Window:

125	1000	00:35:15	92.97%	0.1657	0.0100
132	1050	00:39:59	93.75%	0.1979	0.0100
138	1100	00:43:37	93.75%	0.1589	0.0100
144	1150	00:46:41	96.88%	0.0874	0.0100
150	1200	00:50:29	92.97%	0.1909	0.0100
157	1250	00:53:58	96.09%	0.1433	0.0100
163	1300	00:57:15	96.88%	0.1145	0.0100
169	1350	01:00:36	94.53%	0.1201	0.0100
175	1400	01:04:03	96.09%	0.1487	0.0100
182	1450	01:07:30	96.09%	0.1509	0.0100
188	1500	01:11:13	95.31%	0.1287	0.0100
194	1550	01:14:48	94.53%	0.1615	0.0100
200	1600	01:17:59	92.19%	0.2383	0.0100
207	1650	01:20:07	94.53%	0.1034	0.0100
213	1700	01:21:46	94.53%	0.1554	0.0100
219	1750	01:23:43	93.75%	0.1824	0.0100
225	1800	01:25:43	96.09%	0.0953	0.0100
232	1850	01:28:52	97.66%	0.0885	0.0100
238	1900	01:31:13	92.97%	0.1952	0.0100
244	1950	01:33:18	96.88%	0.0891	0.0100
250	2000	01:35:00	98.44%	0.0695	0.0100

imdsTest_rsiz =
 augmentedImageDatastore with properties:

• Screenshot Proses Training Network (Image Size 64x64x3)

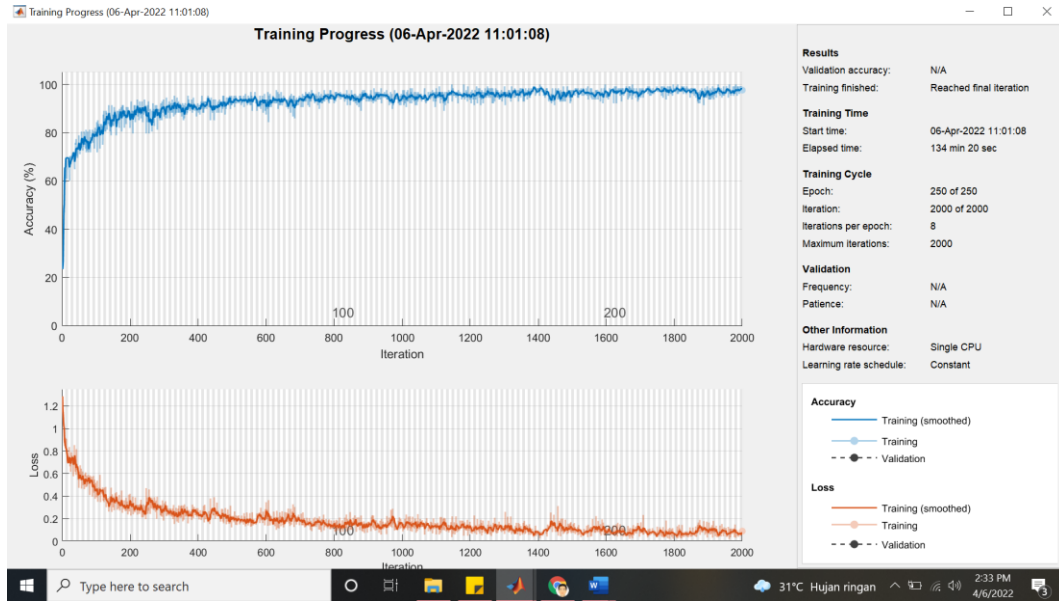


Deep Learning Network Analyzer

net
Analysis date: 06-Apr-2022 13:16:17

25 layers 0 warnings 0 errors

ANALYSIS RESULT			
Name	Type	Activations	Learnables
1 input 64x64x3 images with 'zerocenter' normalization	Image Input	64x64x3	-
2 conv_1 8 3x3x3 convolutions with stride [1 1] and padding 'same'	Convolution	64x64x8	weights 3x3x3x8 Bias 1x1x8
3 batchnorm_1 Batch normalization with 8 channels	Batch Normalization	64x64x8	Offset 1x1x8 Scale 1x1x8
4 relu_1 ReLU	ReLU	64x64x8	-
5 maxpool_1 2x2 max pooling with stride [2 2] and padding [0 0 0 0]	Max Pooling	32x32x8	-
6 conv_2 16 3x3x8 convolutions with stride [1 1] and padding 'same'	Convolution	32x32x16	weights 3x3x8x16 Bias 1x1x16
7 batchnorm_2 Batch normalization with 16 channels	Batch Normalization	32x32x16	Offset 1x1x16 Scale 1x1x16
8 relu_2 ReLU	ReLU	32x32x16	-
9 maxpool_2 2x2 max pooling with stride [2 2] and padding [0 0 0 0]	Max Pooling	16x16x16	-
10 conv_3 32 3x3x16 convolutions with stride [1 1] and padding 'same'	Convolution	16x16x32	weights 3x3x16x32 Bias 1x1x32
11 batchnorm_3 Batch normalization with 32 channels	Batch Normalization	16x16x32	Offset 1x1x32 Scale 1x1x32
12 relu_3 ReLU	ReLU	16x16x32	-
13 maxpool_3 2x2 max pooling with stride [2 2] and padding [0 0 0 0]	Max Pooling	8x8x32	-
14 conv_4 64 3x3x32 convolutions with stride [1 1] and padding 'same'	Convolution	8x8x64	weights 3x3x32x64 Bias 1x1x64
15 batchnorm_4 Batch normalization with 64 channels	Batch Normalization	8x8x64	Offset 1x1x64 Scale 1x1x64



Editor - D:\revisi 2\SKRIPSI\PROGRAM\train_classifire.m

```

10 % FROSES AUGMENTASI DATA
11 augmenter = imageDataAugmenter(
12     'RandXReflection',true,...
13     'RandRotation',[-180 180],...
14     'RandXScale',[1 4],...
15     'RandYReflection',true,...
16     'RandYScale',[1 4]);
17
18 % MEMISAHKAN DATA TRAINING DAN DA
19 [imdsTrain,imdsTest] = splitEachI
20
21 % UKURAN CITRA
22 imageSize = [64 64 3];
23 datastore = augmentedImageDataac
24 (imageSize,imdsTrain,
25
26 %LAYER CNN
27 layers = [ ...
28     imageInputLayer(imageSize,'Na
29     convolution2dLayer(3,8,'Paddi
30     batchNormalizationLayer
31     reluLayer
32     maxPooling2dLayer(2,'Stride',
33     convolution2dLayer(3,16,'Padd
    
```

Command Window

132	1050	01:07:11	94.53%	0.1167	0.0100
138	1100	01:10:08	98.44%	0.0607	0.0100
144	1150	01:15:04	97.66%	0.0645	0.0100
150	1200	01:17:56	93.75%	0.1502	0.0100
157	1250	01:27:46	97.66%	0.0639	0.0100
163	1300	01:31:35	96.09%	0.1369	0.0100
169	1350	01:35:51	98.44%	0.1003	0.0100
175	1400	01:39:00	96.88%	0.0881	0.0100
182	1450	01:41:53	96.88%	0.0880	0.0100
188	1500	01:44:32	96.88%	0.0989	0.0100
194	1550	01:47:13	96.09%	0.0836	0.0100
200	1600	01:50:00	95.31%	0.1026	0.0100
207	1650	01:52:56	98.44%	0.0398	0.0100
213	1700	01:55:41	96.09%	0.0929	0.0100
219	1750	01:58:38	98.44%	0.1289	0.0100
225	1800	02:01:34	98.44%	0.0475	0.0100
232	1850	02:04:50	97.66%	0.0689	0.0100
238	1900	02:07:39	97.66%	0.0707	0.0100
244	1950	02:11:21	96.88%	0.0709	0.0100
250	2000	02:14:20	97.66%	0.0894	0.0100

imdsTest_resz =
 augmentedImageDatastore with properties:

• Screenshot Proxes Training Network (Image Size 128x128x3)

Deep Learning Network Analyzer

net
Analysis date: 06-Apr-2022 21:40:21

25 layers 0 warnings 0 errors

```

graph TD
    input --> conv_1
    conv_1 --> batchnorm_1
    batchnorm_1 --> relu_1
    relu_1 --> maxpool_1
    maxpool_1 --> conv_2
    conv_2 --> batchnorm_2
    batchnorm_2 --> relu_2
    relu_2 --> maxpool_2
    maxpool_2 --> conv_3
    conv_3 --> batchnorm_3
    batchnorm_3 --> relu_3
    relu_3 --> maxpool_3
    maxpool_3 --> conv_4
    conv_4 --> fc_1
    fc_1 --> relu_5
            
```

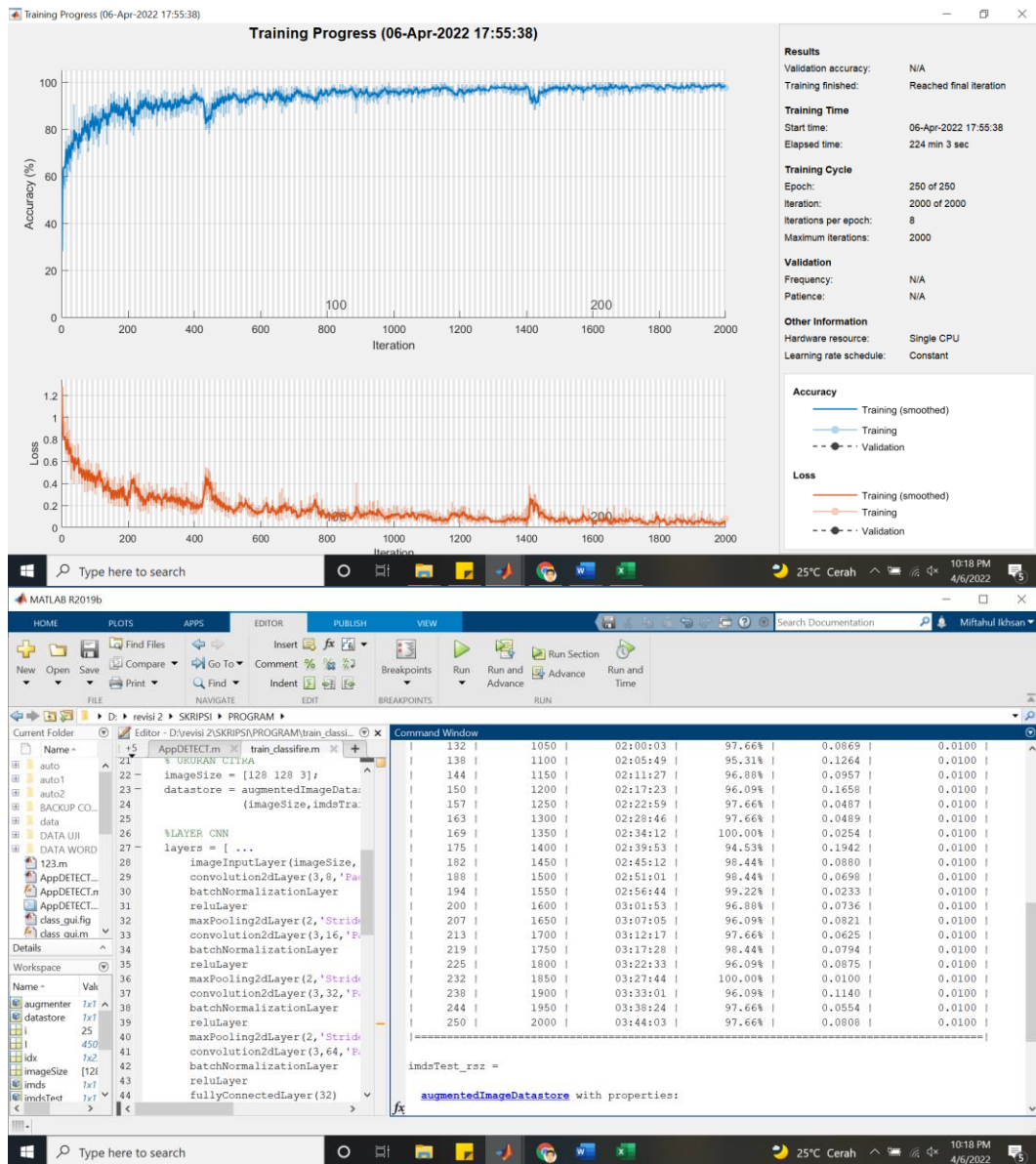
Name	Type	Activations	Learnables
1 input 128x128x3 images with 'zerocenter' normalization	Image Input	128x128x3	-
2 conv_1 8 3x3x3 convolutions with stride [1 1] and padding 'same'	Convolution	128x128x8	Weights 3x3x3x8 Bias 1x1x8
3 batchnorm_1 Batch normalization with 8 channels	Batch Normalization	128x128x8	Offset 1x1x8 Scale 1x1x8
4 relu_1 ReLU	ReLU	128x128x8	-
5 maxpool_1 2x2 max pooling with stride [2 2] and padding [0 0 0 0]	Max Pooling	64x64x8	-
6 conv_2 16 3x3x3 convolutions with stride [1 1] and padding 'same'	Convolution	64x64x16	Weights 3x3x3x16 Bias 1x1x16
7 batchnorm_2 Batch normalization with 16 channels	Batch Normalization	64x64x16	Offset 1x1x16 Scale 1x1x16
8 relu_2 ReLU	ReLU	64x64x16	-
9 maxpool_2 2x2 max pooling with stride [2 2] and padding [0 0 0 0]	Max Pooling	32x32x16	-
10 conv_3 32 3x3x3 convolutions with stride [1 1] and padding 'same'	Convolution	32x32x32	Weights 3x3x3x32 Bias 1x1x32
11 batchnorm_3 Batch normalization with 32 channels	Batch Normalization	32x32x32	Offset 1x1x32 Scale 1x1x32
12 relu_3 ReLU	ReLU	32x32x32	-
13 maxpool_3 2x2 max pooling with stride [2 2] and padding [0 0 0 0]	Max Pooling	16x16x32	-
14 conv_4 64 3x3x3 convolutions with stride [1 1] and padding 'same'	Convolution	16x16x64	Weights 3x3x3x64 Bias 1x1x64
15 batchnorm_4 Batch normalization with 64 channels	Batch Normalization	16x16x64	Offset 1x1x64 Scale 1x1x64

Figure 1

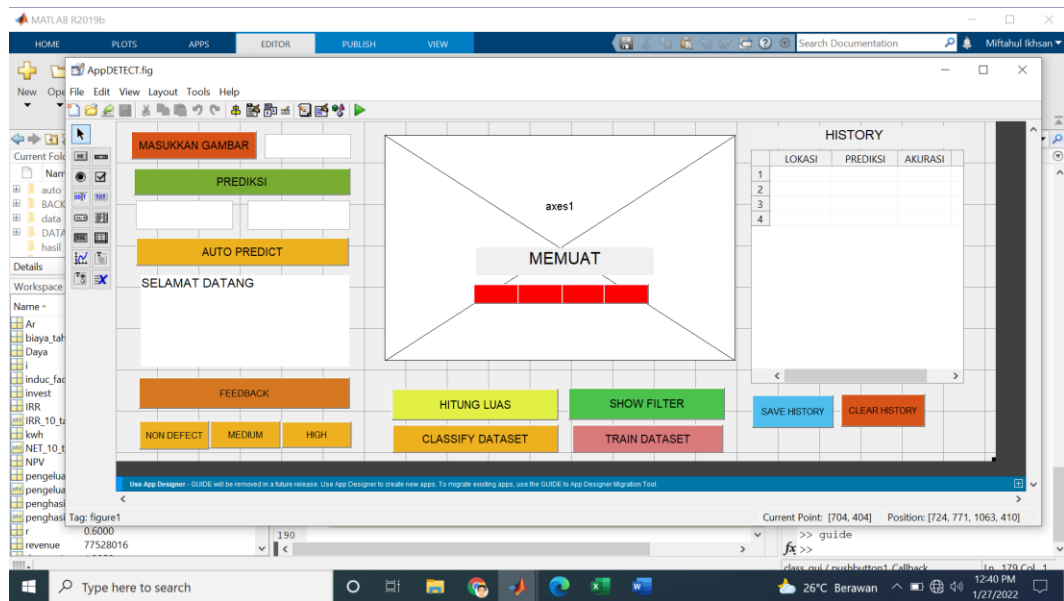
File Edit View Insert Tools Desktop Window Help

POOR GOOD POOR GOOD POOR
GOOD FAIR POOR FAIR POOR
POOR GOOD POOR POOR FAIR
FAIR GOOD POOR POOR POOR
POOR FAIR FAIR POOR FAIR

Type here to search 25°C Cerah 10:17 PM 4/6/2022



- Screenshot Proses Desain GUI



- Foto uji validasi luas karat







**KEMENTERIAN PENDIDIKAN DAN KEBUDAYAAN
UNIVERSITAS HASANUDDIN
FAKULTAS TEKNIK**

Jl. Poros Malino Km.06 Bontomarannu (92172) Gowa, Sulawesi Selatan, 92172
Telp. (0411) 586015,586262 Fax. (0411) 586015
<http://eng.unhas.ac.id> Email : teknik@unhas.ac.id

SURAT PENUGASAN
No. 2168/UN4.7.1/TD.06/2022

Dari : Dekan Fakultas Teknik Universitas Hasanuddin
Kepada : 1. **Haryanti Rivai, S.T., M.T.,Ph.D.** **Pemb. I**
2. **Rahimuddin, S.T., M.T., Ph.D.** **Pemb. II**

Isi : 1. Berdasarkan Surat Ketua Departemen Teknik Sistem Perkapalan Fakultas Teknik Nomor: 2167/UN4.7.7/TD.06/2021 tanggal 4 Januari 2022, tentang usul DOSEN PEMBIMBING MAHASISWA, maka dengan ini kami menugaskan Saudara untuk membimbing penulisan Skripsi/Tugas Akhir mahasiswa Teknik Sistem Perkapalan Fakultas Teknik Universitas Hasanuddin di bawah ini :

Nama : **Miftahul Ikhsan Tahir** No. Stambuk : **D33116307**

Judul Skripsi/Tugas Akhir:

Aplikasi Pendeteksi Korosi pada Pelat kapal dengan Metode Convolutional Neural Network

2. Surat penugasan pembimbing ini mulai berlaku sejak tanggal ditetapkannya dan berakhir sampai selesainya penulisan Skripsi/Tugas Akhir Mahasiswa tersebut.
3. Agar surat penugasan ini dilaksanakan sebaik - baiknya dengan penuh rasa tanggung jawab.

Ditetapkan di Gowa,
Pada tanggal 4 Februari 2022
a.n Dekan,
Wakil Dekan I Bidang Akademik, Riset dan
Inovasi Fakultas Teknik UH

Prof. Baharuddin Hamzah, ST.,M.Arch.,Ph.D
Nip. 19690308 199512 1 001

Tembusan:

1. Dekan FT-UH.
2. Ketua Departemen Teknik Sistem Perkapalan FT-UH.
3. Mahasiswa yang bersangkutan



CERTIFICATE NO. JKT 36788



KEMENTERIAN PENDIDIKAN DAN KEBUDAYAAN
DEPARTEMEN TEKNIK SISTEM PERKAPALAN
FAKULTAS TEKNIK UNIVERSITAS HASANUDDIN
Jalan Poros Malino Km. 6 Bontomarannu, Gowa, Sulawesi Selatan, 92172
Telp/Fax: +62-411-588400, Email: kapal9uh@indosat.net.id

No. : 2167/UN4.7.7/TD.06/2022
Lamp : -
Hal : Penugasan Bimbingan Tugas Akhir

Kepada Yth : **Wakil Dekan I**
Bidang Akademik, Riset dan Inovasi
Fakultas Teknik UNHAS
di-
Gowa

Dengan hormat,
Kiranya dosen pembimbing tugas akhir (skripsi) dari mahasiswa :

Nama : Miftahul Ikhsan Tahir
Stambuk : D33116307
Program Studi : Teknik Sistem Perkapalan

Dengan judul Tugas Akhir:
Aplikasi Pendeteksi Korosi pada Pelat kapal dengan Metode Convolutional Neural Network

Dosen Pembimbing :

1. Haryanti Rivai, S.T., M.T., Ph.D.
2. Rahimuddin, S.T., M.T., Ph.D.

Dapat dibuatkan Surat Penugasan Bimbingan Tugas Akhir
Demikian penyampaian kami, atas perhatian dan kerjasamanya diucapkan terima kasih.

Gowa, 4 Februari 2022

Ketua,



Dr. Eng. Faisal Mahmuddin, S.T., M.Inf.Tech., M.Eng.
Nip. 19810211 200501 1 003



KEMENTERIAN PENDIDIKAN, KEBUDAYAAN,
RISET, DAN TEKNOLOGI
UNIVERSITAS HASANUDDIN
KAMPUS TAMALANREA

JALAN PERINTIS KEMERDEKAAN KM.10 MAKASSAR 90245
TELEPON : 0411-586200 (6 SALURAN), 584002, FAX: 585188

SURAT PERSETUJUAN

Nomor : 10563/UN4.1.1.2.1.1/PK.02.03/2022


Berdasarkan Peraturan Rektor Universitas Hasanuddin tentang Penyelenggaraan Program Sarjana Nomor : 2781/UN4.1/KEP/2018 tanggal 16 Juli 2018, dengan ini menerangkan bahwa :

NIK : 7315060808980008 ✓
N a m a : MIFTAHUL IKHSAN TAHIR ✓
Tempat/Tanggal Lahir : SIDOMULYO, 14 AGUSTUS 1998 ✓
NIM : D33116307 ✓
Fakultas : TEKNIK ✓
Program Studi : TEK. SISTEM PERKAPALAN ✓

Telah memenuhi syarat untuk Ujian Skripsi Strata I (S1) **PERIODE IV MEI 2022 TAHAP I**. Demikian Surat Persetujuan ini dibuat untuk digunakan dalam proses pelaksanaan ujian skripsi, dengan ketentuan dapat mengikuti wisuda **PERIODE IV MEI 2022 TAHAP I**, jika persyaratan kelulusan/wisuda telah **dipenuhi**. Terima Kasih.

Makassar, 13 APRIL 2022

Kepala Biro Administrasi Akademik
u.b Kepala Sub Bagian Pendidikan dan Evaluasi
Universitas Hasanuddin,


MURSALIM, S.Sos.
NIP. 19730216 199601 1001

Keterangan :

Nomor User : D33116307 ✓
Nomor password/pin : 2163511 ✓
Alamat Website : <http://unhas.ac.id/akad/wisuda/>
Catatan :

1. Bagi Mahasiswa yang telah melaksanakan ujian Sarjana dan dinyatakan lulus, segera menyerahkan lembar pengesahan Skripsi dan Berita Acara Ujian Sarjana ke Sub Bagian Akademik Fakultas, untuk memperoleh nomor Alumni dan didaftar sebagai Wisudawan pada periode berjalan.
2. Jika terjadi perubahan Judul Skripsi agar melaporkan ke Kasubag. Pendidikan Fakultas sebelum didaftar sebagai Wisudawan pada Periode berjalan
3. Pada saat ON-LINE Mahasiswa diharapkan mengisi identitas diri sesuai surat izin ujian ini
4. Surat izin ini hanya berlaku untuk Wisuda periode berjalan (WISUDA PERIODE IV MEI 2022 TAHAP I)





KEMENTERIAN PENDIDIKAN, KEBUDAYAAN, RISET, DAN TEKNOLOGI

DEPARTEMEN TEKNIK SISTEM PERKAPALAN

FAKULTAS TEKNIK UNIVERSITAS HASANUDDIN

Jalan Poros Malino KM 6. Bontomarannu (92171) Gowa, Sulawesi Selatan

Telp. (0411) 588400 Fax. (0411) 2006

No. : 8236/UN4.7.7/TD.06/2022
Lamp : -
Hal : Penerbitan Surat Penugasan Panitia
Ujian Sarjana Strata Satu (S1)

Kepada Yth : **Wakil Dekan I**
Bidang Akademik, Riset dan Inovasi
Fakultas Teknik UNHAS
di-
Gowa

Dengan hormat,

Berdasarkan Persetujuan Pembimbing Mahasiswa, Bersama ini diusulkan susunan Panitia Ujian Sarjana Strata Satu (S1) bagi mahasiswa Departemen Teknik Sistem Perkapalan Fakultas Teknik Universitas Hasanuddin atas nama :

Nama : Miftahul Ikhsan Tahir
Stambuk : D33116307

Maka dengan ini kami sampaikan Susunan Panitia Ujian Sarjana Strata Satu (S1) sebagai berikut :

Ketua : Haryanti Rivai, S.T., M.T., Ph.D.
Sekretaris : Rahimuddin, S.T., M.T., Ph.D.
Anggota : 1. Andi Haris Muhammad, S.T., M.T., Ph.D.
2. Dr. Ir. Ganding Sitepu, Dipl.-Ing.

Judul Tugas Akhir mahasiswa yang bersangkutan adalah :

Aplikasi Pendeteksi Korosi pada Pelat kapal dengan Metode Convolutional Neural Network

Untuk dapat diterbitkan surat penugasannya.

Demikian penyampaian kami, atas perhatian dan kerjasamanya diucapkan terima kasih.

G o w a, 22 April 2022

Ketua Departemen Teknik Sistem Perkapalan



Faisal
Dr. Eng. Faisal Mahmuddin, S.T., M.Inf.Tech., M.Eng
Nip. 19810211 200501 1 003



KEMENTERIAN PENDIDIKAN, KEBUDAYAAN, RISET, DAN TEKNOLOGI
UNIVERSITAS HASANUDDIN
FAKULTAS TEKNIK

Jalan Poros Malino Km. 6. Bontomarannu Gowa, 92171, Sulawesi Selatan
Telp. (0411) 586015, 586262 Fax (0411) 586015.
<http://eng.unhas.ac.id> E-mail:teknik@unhas.ac.id

SURAT PENUGASAN

No.8237/UN4.7.7/TD.06/2022

Dari : Dekan Fakultas Teknik Universitas Hasanuddin
Kepada : Mereka yang tercantum namanya dibawah ini.

Isi : 1. Bahwa berdasarkan peraturan Akademik Universitas Hasanuddin Tahun 2018 pasal 19 (SK. Rektor Unhas nomor : 2781/UN4.1/KEP/2018), dengan ini menugaskan Saudara sebagai PANITIA UJIAN SARJANA Program Strata Satu (S1) Teknik Sistem Perkapalan Fakultas Teknik Universitas Hasanuddin dengan susunan sebagai berikut :

Ketua : Haryanti Rivai, S.T., M.T..Ph.D.
Sekretaris : Rahimuddin, S.T., M.T., Ph.D.
Anggota : 1. Andi Haris Muhammad, S.T., M.T., Ph.D
2. Dr. Ir. Ganding Sitepu, Dipl.-Ing.

Untuk menguji bagi mahasiswa tersebut dibawah ini :

Nama/Nim : Miftahul Ikhsan Tahir / D33116307
Departemen : Teknik Sistem Perkapalan

Judul Thesis/Skripsi :

Aplikasi Pendeteksi Korosi pada Pelat kapal dengan Metode Convolutional Neural Network

2. Waktu ujian ditetapkan oleh Panitia Ujian Akhir Program Strata Satu (S1).
3. Agar surat penugasan ini dilaksanakan sebaik-baiknya dengan penuh rasa tanggung jawab.
4. Surat penugasan ini berlaku sejak tanggal ditetapkan sampai dengan berakhirnya Ujian Sarjana tersebut, dengan ketentuan bahwa segala sesuatunya akan ditinjau dan diperbaiki sebagaimana mestinya apabila dikemudian hari ternyata terdapat kekeliruan dalam keputusan ini.

Ditetapkan di Gowa,
Pada tanggal , 22 April 2022
a.n Dekan,
Wakil Dekan I Bidang Akademik, Riset dan
Inovasi Fakultas Teknik UH

Prof. Baharuddin Hamzah, ST.,M.Arch.,Ph.D
Nip. 19690308 199512 1 001

Tembusan :

1. Dekan FT-UH.
2. Ketua Departemen Teknik Sistem Perkapalan FT-UH.
3. Kasubag Umum dan Perlengkapan FT-UH



CERTIFICATE NO. JKT 36788



KEMENTERIAN PENDIDIKAN, KEBUDAYAAN, RISET, DAN TEKNOLOGI

DEPARTEMEN TEKNIK SISTEM PERKAPALAN

FAKULTAS TEKNIK UNIVERSITAS HASANUDDIN

Jalan Poros Malino KM 6. Bontomarannu (92171) Gowa, Sulawesi Selatan

Telp. (0411) 588400 Fax. (0411) 2006

Nomor : 8236/UN4.7.7/TD.06/2022
Lamp : -
Hal : Undangan Ujian Akhir

22 April 2022

Kepada
Yth. : 1. **Andi Haris Muhammad, S.T., M.T., Ph.D**
2. **Dr. Ir. Ganding Sitepu, Dipl.-Ing.**

Dengan hormat,

Kami mengundang Saudara/saudari kiranya berkenan hadir untuk menyaksikan/bertindak selaku penguji Ujian Akhir Strata Satu Fakultas Teknik Universitas Hasanuddin yang akan diselenggarakan pada :

Hari / Tanggal : **Selasa, 26 April 2022**
Jam : **13.00 Wita - selesai**
Tempat : **Ruang Sidang Teknik Sistem Perkapalan (Daring/Luring)**

Dibawakan oleh :

Nama / Stambuk : Miftahul Ikhsan Tahir / D33116307

Atas kesedian dan kehadiran Saudara/Saudari diucapkan terima kasih.



Dr. Eng. Faisal Mahmuddin, S.T., M.Inf.Tech., M.Eng.
Nip. 19810211 200501 1 003



KEMENTERIAN PENDIDIKAN, KEBUDAYAAN, RISET, DAN TEKNOLOGI
UNIVERSITAS HASANUDDIN
FAKULTAS TEKNIK

Jalan Poros Malino Km. 6, Bontomarannu Gowa, 92171, Sulawesi Selatan
☎ Telp. (0411) 586015, 586262 Fax (0411) 586015.
<http://eng.unhas.ac.id> E-mail teknik@unhas.ac.id

BERITA ACARA UJIAN SARJANA

Terhadap Mahasiswa

Nama : Miftahul Ikhsan Tahir
Stambuk : D33116307
Judul : *Aplikasi Pendeteksi Korosi Pada Pelat Kapal Dengan Metode Convolutional Neural Network*
Hari/Tanggal : Selasa, 26 April 2022
Waktu : 13.00 Wita - selesai
Tempat : Ruang Sidang Teknik Sistem Perkapalan (Daring/Luring)
Keputusan Sidang / Catatan : lulus dengan nilai $A = 87$

PANITIA UJIAN

No.	Susunan Panitia	Nama	Tanda Tangan
1	Ketua/Anggota	Haryanti Rivai, S.T., M.T., Ph.D.	1.
2	Sekretaris/Anggota	Rahimuddin, S.T., M.T., Ph.D.	2.
3	Anggota	Andi Haris Muhammad, S.T., M.T., Ph.D.	3.
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