

## CHAPTER 5

### IMPLEMENTATION AND TESTING

#### 5.1 Implementation

This project requires training data and testing data. Of the two processes, there are similarities in the processes used, namely determining the image to be processed, converting the image to grayscale, changing image from grayscale to binary and calculating the average of R,G,B.

##### 5.1.1 Function determines image to be processes.

```
1. public void setImage(BufferedImage input) {  
2.     realImage = input;  
3.     setSize();  
4.     imageToGray();  
5.     grayToBinary();  
6.     hitungMeanR_G_B();  
7. }
```

On this function, used to call several functions. In the second line used for explain that `input` is defined as `realImage`, it will be used for next function. Lines three until four contain the commands to call functions `setSize` and `imageToGray`. Then on the fifth line used for call function `grayToBinary`. On the sixth line used for call function `hitungMeanR_G_B`.

##### 5.1.2 Function finding original image size.

```
1. private void setSize(){  
2.     tinggiCitra = realImage.getHeight();  
3.     lebarCitra = realImage.getWidth();  
4. }
```

Based on the function above, used for get the original image size. On the second line contains commands to return height dimensions with double precision and stored in high variable images. On the third line contain commands to return width dimensions with double precision and stored in high variable image.

### 5.1.3 Original image to grayscale conversion function.

```

1. private void imageToGray() {
2.     double red, green, blue;
3.     int gray;
4.     Color before, after;
5.
6.     BufferedImage output = new BufferedImage(lebarCitra,
        tinggiCitra, BufferedImage.TYPE_BYTE_GRAY);
7.
8.     for (int y = 0; y < tinggiCitra; y++) {
9.         for (int x = 0; x < lebarCitra; x++) {
10.
11.             before = new Color(realImage.getRGB(x, y));
12.
13.             // Calculate RGB to gray
14.             // with lumonisity algorithm
15.             red = (double) (before.getRed() * 0.2989);
16.             green = (double) (before.getGreen() * 0.5870);
17.             blue = (double) (before.getBlue() * 0.1140);
18.             gray = (int) (red + green + blue);
19.
20.             after = new Color(gray, gray, gray);
21.
22.             output.setRGB(x, y, after.getRGB());
23.         }
24.     }
25.     grayImage = output;
26. }

```

Based on function 5.1.3, used for change the original image. On the second until four line used to declare a variable to be used. Sixth line contains commands to take lebarCitra, tinggiCitra data stored in the output variable. On the eight until twenty-four lines used for processing image where each R,G,B is multiplied by a certain value then the results are stored in the grayImage variable so that it can be used in other function.

#### 5.1.4 Grayscale image to binary conversion function.

```

1. private void grayToBinary() {
2.   Color before, after;
3.   ArrayObjek.clear();
4.
5.   BufferedImage output = new BufferedImage(lebarCitra,
        tinggiCitra, BufferedImage.TYPE_BYTE_GRAY);
6.
7.   for (int y = 0; y < tinggiCitra; y++) {
8.     for (int x = 0; x < lebarCitra; x++) {
9.       before = new Color(this.grayImage.getRGB(x, y));
10.
11.      if (before.getBlue() < 251) {
12.        after = new Color(255, 255, 255);
13.        ArrayObjek.add(String.valueOf(x) + "," +
            String.valueOf(y));
14.      } else {
15.        after = new Color(0, 0, 0);
16.      }
17.      output.setRGB(x, y, after.getRGB());
18.    }
19.  }
20.  binaryImage = output;
21. }

```

Pada fungsi ini, digunakan untuk mengubah grayscale menjadi binary. Pada baris 2 digunakan untuk deklarasi variabel, baris 3 berisi perintah untuk mengosongkan ArrayObjek. Kemudian pada baris selanjutnya sama dengan mengubah gambar asli ke grayscale, lalu pada baris ke 11-16 berisi perintah jika `before.getBlue() < 251` lalu menambahkan data x dan y kedalam arrayObjek (pada baris 13), dan hasil dari perintah tersebut disimpan di variabel binaryImage.

On this function, used to change grayscale to binary. On the second line used for declare variable, third line contains command for clear the ArrayObjek. Then on the next line is the same as changing the original image to grayscale, then on the 11-16 lines contains command if `before.getBlue() < 251` then add data x and y into arrayObjek(on the 13 line), and the result is stored in the binary image variable.

### 5.1.5 Calculate the mean R,G,B function

```

1. private void hitungMeanR_G_B() {
2.     int length = ArrayObjek.size();
3.     double red = 0;
4.     double green = 0;
5.     double blue = 0;
6.     int x, y;
7.     Color before;
8.     String xy[];
9.
10.    for (int i = 0; i < length; i++) {
11.        xy = ArrayObjek.get(i).toString().split(",");
12.
13.        //System.out.println(xy[0] + "-" + xy[1]);
14.        x = Integer.parseInt(xy[0]);
15.        y = Integer.parseInt(xy[1]);
16.
17.        before = new Color(this.realImage.getRGB(x, y));
18.
19.        red += before.getRed();
20.        blue += before.getBlue();
21.        green += before.getGreen();
22.    }
23.    red /= length;
24.    blue /= length;
25.    green /= length;
26.
27.    meanR = red;
28.    meanG = green;
29.    meanB = blue;
30.    }

```

Based on function `hitungMeanR_G_B` used for calculate mean R,G,B on each image. On the second until eight lines used for the variable declaration to be used. On the 19-21 line contains command to find the R,G,B value on each pixel. 23-25 line contains command to find the mean R,G,B from each image, then on the 27-29 line mean value of R,G,B stored on variable `meanR`, `meanG`, `meanB`.

### 5.1.6 Button Training Data

```

542 // initialization
543 String[] fileName = {" "};
544 String[] classFile = {" "};
545
546 if (j <= total_data_train_jn) {
547     fileName[0] = (path + "/" + segar + Integer.toString(j) + ".jpg");
548     imageInput = ImageIO.read(new File(path + "/" + segar + Integer.toString(j) + ".jpg"));
549     classFile[0] = "Segar";
550 } else {
551     fileName[0] = (path + "/" + tidaksegar + Integer.toString(j - total_data_train_jn) + ".jpg");
552     imageInput = ImageIO.read(new File(path + "/" + tidaksegar + Integer.toString(j - total_data_train_jn) + ".jpg"));
553     classFile[0] = "Tidak Segar";
554 }
555
556

```

Illustration 5.1: Button training data

On this training data button, 546-549 lines contains command if the file path is fresh, and using the jpg extension will be classified as Fresh in the class column. And 550-553 contains command if the file path is not fresh and using the jpg extension will be classified as Not Fresh in the class column.

### 5.1.7 Function Training Data

```

1. public ProsesGambar dataTraining(BufferedImage input, String
   className) {
2.     ProsesGambar image = new ProsesGambar();
3.
4.     image.setImage(input);
5.     double[] fitur = {image.getR(), image.getG(), image.getB()};
6.     valueFeature = new ArrayDataTiapFitur(fitur[0], fitur[1],
   fitur[2]);
7.     for (int i = 0; i < class_.length; i++) {
8.         if(class_[i].toLowerCase().equals(className.toLowerCase()))
   {
9.             this.dataset.get(i).addData(valueFeature);
10.            break;
11.        }
12.    }
13.    return image;
14. }

```

On this function, used for processing training data. Second line serves to create an ProsesGambar object. Third line is used to input the image into the setImage function in the ProsesGambar class. Then on the fifth-sixth line contains command for inputting the mean R,G,B results into the feature array, and those feature being inserted into valueFeature variable. Then lines 7-13 are used to populate the dataset array according to the valueFeature variable.

### 5.1.8 Function Testing Data

```

102 public String dataTesting(BufferedImage input) {
103
104     //Take data sample test
105     ArrayDataLikelihood dataTesting;
106     ProsesGambar imageTesting = new ProsesGambar();
107     imageTesting.setImage(input);
108     System.out.println(input);
109     double[] fitur = {imageTesting.getR(), imageTesting.getG(), imageTesting.getB()};
110     dataTesting = new ArrayDataLikelihood(fitur[0], fitur[1], fitur[2]);
111     double[] sampel = dataTesting.getMatrix();
112
113     //Create ArrayDataMeanVar new ArrayList
114     ArrayList<ArrayDataMeanVar> cl = new ArrayList<>();
115
116     System.out.println("class length: " + cl.length());
117     System.out.println("class class: " + cl.get(0));
118
119     for(int i = 0; i < cl.length(); i++) {
120         System.out.println("class: " + cl.get(i));
121         System.out.println("data: " + dataTesting.get(i).toString());
122         cl.add(new ArrayDataMeanVar(cl.get(i).getMatrix()));
123     }
124
125     //Create ArrayDataLikelihood
126     ArrayList<ArrayDataLikelihood> lk = new ArrayList<>();
127     for(int i = 0; i < cl.length(); i++) {
128         lk.add(new ArrayDataLikelihood(cl.get(i).getMatrix(), sampel));
129     }
130
131     //Create ArrayDataPosterior
132     ArrayList<ArrayDataPosterior> pt = new ArrayList<>();
133     for(int k = 0; k < lk.length(); k++) {
134         pt.add(new ArrayDataPosterior(lk.get(k), lk.get(k).getProbClass()));
135     }
136
137     //Final result
138     ArrayDataPosterior chosen = pt.get(0);
139     for(int m = 1; m < pt.length(); m++) {
140         if(chosen.getProbability() < pt.get(m).getProbability()) {
141             chosen = pt.get(m);
142         }
143     }
144
145     this.resultPosterior = pt;
146
147     System.out.println("class: " + chosen.getProbClass());
148     return chosen.getProbClass();
149 }
150
151

```

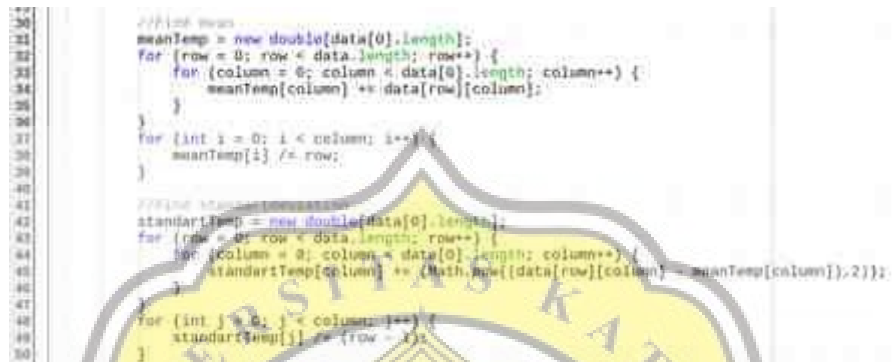
Illustration 5.2: Fungsi Data Testing

Based on function above, used to processing testing data. Line 106 creates an *ProsesGambar* object. Line 107 is used to input the image into the *setImage* function in the *ProsesGambar* class. Then lines 109-110 contain commands for inputting the mean R,G,B results into the feature array, and those feature being entered into the *valueFeature* variable. Then line 114 is used to create an arraylist which will be used to calculate the mean and variance of each class in the training data, then on lines 119-123 is used to add data to the *ArrayDataMeanVar* arraylist according to the class and data. On lines 126-129 contains commands for adding data to the *ArrayDataLikelihood* array according to class, mean and variance. Lines 132-135 are used to add data to the *ArrayDataPosterior* arraylist according



to the class and likelihood calculation results. Then on lines 138-143 contains commands to display the final classification result.

### 5.1.9 Function looks for Mean and Varian



```

30 //Find mean
31 meanTemp = new double[data[0].length];
32 for (row = 0; row < data.length; row++) {
33     for (column = 0; column < data[0].length; column++) {
34         meanTemp[column] += data[row][column];
35     }
36 }
37 for (int i = 0; i < column; i++) {
38     meanTemp[i] /= row;
39 }
40
41 //Find standard deviation
42 standartTemp = new double[data[0].length];
43 for (row = 0; row < data.length; row++) {
44     for (column = 0; column < data[0].length; column++) {
45         standartTemp[column] += (Math.pow((data[row][column] - meanTemp[column]), 2));
46     }
47 }
48 for (int j = 0; j < column; j++) {
49     standartTemp[j] /= (row - 1);
50 }

```

Illustration 5.3: Function mencari mean dan varian

The formula for finding the mean and variance for each class on training data can be seen in illustration 5.3. lines 31-39 contain commands for calculating the mean, and lines 42-50 contain commands for calculating the variance.

### 5.1.10 Function looks for probability



```

27 double[] probabilitas = new double[column];
28 for (int i = 0; i < column; i++) {
29     probabilitas[i] = (double) (standartTemp[i] / Math.sqrt(standartTemp[i]));
30 }
31
32 double a = Math.sqrt(standartTemp);
33
34 probabilitas[i] = 1 / (Math.sqrt(2 * Math.PI) * Math.pow(a, 0.5) * Math.pow((value - meanTemp[i]), 2));
35
36 this.probabilitas = probabilitas;
37
38 this.calculateProbabilitas();
39
40 }

```

Illustration 5.4: Fungsi Mencari Probabilitas

Lines 27-28 used for to take the variance value, the mean and store it in the variance and mean variables. On the line 29 take testing data value and store it in value\_ variable. Line 31 used for change variance value to standard deviation by using formula  $\sqrt{\text{variance}}$ . Then, on the 34 line is a function to calculate the probability of each class.

### 5.1.11 Function looks for final probability value.

```

21 public ArrayDataPosterior(String class_, ArrayDataLikelihood likelihood, double probClass) {
22     double tempProbPosterior = probClass;
23
24     //Kalikan semua probabilitas filter untuk mendapatkan posterior
25     for(int i = 0; i < likelihood.getPFeatureClass().length; i++) {
26         tempProbPosterior *= likelihood.getPFeatureClass()[i];
27     }
28     this.nameClass = class_;
29     this.ProbPosterior = tempProbPosterior;
30     //data.add(tempProbPosterior);
31
32     //Untuk input
33     this.printProbability();
34 }

```

Illustration 5.5: Mencari probabilitas akhir

Illustration 5.5 is a function to find the final probability value.

## 5.2 Testing

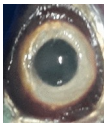

In the testing process, conducted three times with 48 testing data with a composition of eight fresh and non-fresh fish eye image, eight fresh and non-fresh gills, eight fresh and not fresh meat color image. Then used 50,100 and 150 training data which aim to implement the results of the program using Naive Bayes algorithm.

### 5.2.1 Testing 1


#### a Testing 1 Parameter




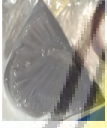
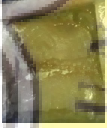

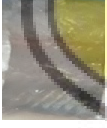



This is the first test table using one parameter:

Table 5.1: One parameter testing table

No	Input	Mean			Probability Value		Image	Output
		Red	Green	Blue	F	NF		
1		122,58	117,25	110,33	3.26367703 12488506E-6	4.64127808 00884197E-7	F	F
2		143,75	148,64	148,64	3.77513504 12877664E-8	6.19994994 9946974E-10	F	F



3		119,24	112,91	106,80	3.55110515 74198456E- 6	7.24655922 899192E-7	F	F
4		118,97	110,42	103,06	3.45549937 2638653E-6	9.12312466 431341E-7	F	F
5		141,38	139,97	134,80	2.60232239 5777415E-7	7.36418163 1297155E-9	F	F
6		135,44	138,49	136,70	3.58065603 92011237E- 7	9.76389765 4556822E-9	F	F
7		126,13	130,00	129,97	1.20668826 64329287E- 6	5.34630774 4117931E-8	F	F
8		143,98	148,94	148,81	3.55670096 8566803E-8	5.80108229 3569649E- 10	F	F
9		103,63	101,08	107,03	2.67540258 6923913E-6	1.38166263 5405866E-6	F	F
10		93,72	94,88	98,54	1.28251888 0104012E-6	1.98209688 0101004E-6	F	NF
11		123,21	113,01	105,26	3.31991387 19593836E- 6	6.67528096 5383367E-7	F	F
12		103,66	101,08	107,03	2.67875733 8485086E-6	1.38086224 98024211E- 6	F	F

13		112,16	110,63	112,69	3.50233043 68694363E- 6	7.26733721 3763175E-7	F	F
14		107,47	103,73	101,15	2.93416454 4024801E-6	1.51755899 8522632E-6	F	F
15		117,54	112,36	106,22	3.59194378 2252404E-6	7.95904476 0700094E-7	F	F
16		117,29	112,09	105,95	3.59279745 49520096E- 6	8.18324175 2448024E-7	F	F
17		99,72	93,12	92,15	1.30670094 52351722E- 6	2.23422580 1331316E-6	F	NF
18		135,76	131,23	124,05	1.01728975 38706175E- 6	5.0025715 91413183E- 8	F	F
19		111,70	106,76	101,10	3.26447101 50763505E- 6	1.30484407 13388687E- 6	F	F
20		96,92	91,24	86,35	8.13633222 3868095E-7	2.28882622 57515526E- 6	F	NF
21		95,49	90,21	85,53	6.90127473 99972E-7	2.27912607 46285893E- 6	F	NF
22		96,16	89,78	87,91	7.91337401 2608586E-7	2.30305610 0082046E-6	F	NF

23		97,62	91,19	88,95	9.54672617 5676622E-7	2.29818167 50289757E- 6	F	NF
24		91,34	82,43	74,96	1.7003103 118890426 E-7	1.80219844 10624583E- 6	F	NF
25		98,97	55,25	65,93	7.75498782 3351549E-9	5.5602841 69782326E- 7	NF	NF
26		107,20	104,52	94,43	2.48989750 51968583E- 6	1.70848859 63520355E- 6	NF	F
27		93,21	89,90	91,53	8.00214206 4077176E-7	2.24590617 00143747E- 6	NF	NF
28		99,96	94,28	100,70	1.76968823 16964376E- 6	1.91145477 37663463E- 6	NF	NF
29		76,47	21,67	63,69	4.37240401 2525958E- 12	2.36746226 74658517E- 8	NF	NF
30		78,55	37,52	75,52	3.19912438 63550855E- 10	1.80146154 02589325E- 7	NF	NF
31		36,19	10,28	22,03	1.09785246 93802138E- 18	1.04308072 28002911E- 11	NF	NF
32		78,08	36,89	75	2.63422541 22225427E- 10	1.66310452 18433931E- 7	NF	NF

33		89,44	79,12	90,36	3.09147905 59609796E- 7	2.02511179 10749455E- 6	NF	NF
34		74,54	72,44	75,98	2.06824819 08348465E- 8	1.07030773 48166412E- 6	NF	NF
35		69,67	71,69	80,79	1.35676071 06164755E- 6	8.15274380 0549914E-8	NF	F
36		94,83	81,42	92,88	5.57946266 0651963E-7	2.14795426 74374034E- 6	NF	NF
37		77,61	82,15	90,10	2.96186040 10874055E- 6	1.50252263 52019168E- 6	NF	F
38		99,72	99,33	105,34	2.20645727 29643677E- 6	1.55287564 23789271E- 6	NF	F
39		83,29	81,11	85,43	1.76482414 12298587E- 7	1.87081917 10138786E- 6	NF	NF
40		57,46	61,60	61,21	1.89021920 9713321E- 10	1.80474112 89255654E- 7	NF	NF
41		110,56	101,64	100,35	2.91983196 66041887E- 6	1.54409131 3638465E-6	NF	F
42		95,48	82,30	76,03	2.31984691 99225332E- 7	1.90234388 533413E-6	NF	NF


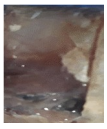

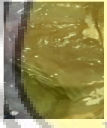
43		68,81	57,92	56,49	3.20610364 34519237E- 10	2.02249741 483476E-7	NF	NF
44		116,95	103,47	97,47	2.92080289 53502673E- 6	1.36423069 9695198E-6	NF	F
45		95,49	90,21	85,53	4.88906032 8226918E-8	1.21598941 2877089E-6	NF	NF
46		91,63	74,57	67,48	1.8725436 702903774 E-8	8.95540770 7576777E-7	NF	NF
47		71,25	61,06	55,18	5.25172941 8035252E- 10	2.36720921 74349508E- 7	NF	NF
48		90,95	82,46	77,79	2.06751843 28986435E- 7	1.92651837 1459945E-6	NF	NF

Table Information :

F = Fresh

NF = Not Fresh

Based on table 5.1 the test result using 50 training data on one parameter showed that 35 data were successfully classified correctly and 13 data were incorrect. And calculated using the precision-recall formula to get an accuracy value of 72,9% with the following calculation:

$$Accuracy = \frac{(TP+TN)}{(TP+FP+FN+TN)}$$

Illustration 5.6: Presicion-recall  
function

$$Accuracy = \frac{(17+18)}{(17+7+6+18)} = \frac{35}{48} = 72,91 \%$$

Information :

True Positive (TP) = The prediction of fresh results is indeed fresh

True Negative (NF) = The predictions are not fresh are really not fresh

False Positive (FP) = Fresh predictions are not fresh

False Negative (FN) = Not Fresh predictions turn out to be fresh

b Testing 3 Parameter

This is the first test table using three parameter:





Table 5.2: Three parameter testing table

No	Input	Probability Value		Probability Value		Image	Output
		F	NF	F	NF		
1		3.26367703 12488506E -6	4.64127808 00884197E -7			F	F
		2.67540258 6923913E- 6	1.38166263 5405866E- 6	2.41526018 7802645E- 6	1.36000541 49153413E -6		
		1.30670094 52351722E -6	2.23422580 1331316E- 6				
2		3.77513504 12877664E -8	6.19994994 9946974E- 10			F	F
		1.28251888 0104012E- 6	1.98209688 0101004E- 6	7.79186661 4625023E- 7	6.77580863 6700435E- 7		
		1.01728975 38706175E -6	5.0025715 91413183E -8				



3		3.55110515 74198456E -6	7.24655922 899192E-7	3.37849668 14851932E -6	8.99009363 5921324E- 7	F	F
		3.31991387 19593836E -6	6.67528096 5383367E- 7				
		3.26447101 50763505E -6	1.30484407 13388687E -6				
4		3.45549937 2638653E- 6	9.12312466 431341E-7	2.31596331 1170183E- 6	1.52733364 73284382E -6	F	F
		2.67875733 8485086E- 6	1.38086224 98024211E -6				
		8.13633222 3868095E- 7	2.28882622 57515526E -6				
5		2.60232239 5777415E- 7	7.36418163 1297155E- 9	1.48423005 01489663E -6	1.00440799 25454012E -6	F	F
		3.50233043 68694363E -6	7.26733721 3763175E- 7				
		6.90127473 99972E-7	2.27912607 46285893E -6				
6		3.58065603 92011237E -7	9.76389765 4556822E- 9	1.36118918 30685909E -6	1.27679299 87530782E -6	F	F

		2.93416454 4024801E- 6	1.51755899 8522632E- 6				
		7.91337401 2608586E- 7	2.30305610 0082046E- 6				
7		1.20668826 64329287E- 6	5.34630774 4117931E- 8			F	F
		3.59194378 2252404E- 6	7.95904476 0700094E- 7				
		9.54672617 5676622E- 7	2.29818167 50289757E- 6				
8		3.55670096 8566803E- 8	5.80108229 3569649E- 10			F	F
		3.59279745 49520096E- 6	8.18324175 2448024E- 7				
		1.7003103 118890426 E-7	1.80219844 10624583E- 6				
9		7.75498782 3351549E- 9	5.5602841 69782326E- 7	1.07891162 00078795E- 6	1.37507717 38972144E- 6	NF	NF
		3.09147905 59609796E- 7	2.02511179 10749455E- 6				

		2.91983196 66041887E -6	1.54409131 3638465E- 6				
10		2.48989750 51968583E -6	1.70848859 63520355E -6	9.14188226 36582E-7	1.56038007 21676022E -6	NF	NF
		2.06824819 08348465E -8	1.07030773 48166412E -6				
		2.31984691 99225332E -7	1.90234388 533413E-6				
11		8.00214206 4077176E- 7	2.24590617 00143747E -6	7.19098509 1295127E- 7	8.43227783 1677833E- 7	NF	NF
		1.35676071 06164755E -6	8.15274380 0549914E- 8				
		3.20610364 34519237E -10	2.02249741 483476E-7				
12		1.76968823 16964376E -6	1.91145477 37663463E -6	1.74947913 10373003E -6	1.80787991 36329828E -6	NF	F
		5.57946266 0651963E- 7	2.14795426 74374034E -6				
		2.92080289 53502673E -6	1.36423069 9695198E- 6				

13		4.37240401 2525958E- 12	2.36746226 74658517E -8	1.00358512 5591229E- 6	9.14062223 5845549E- 7	NF	NF
		2.96186040 10874055E -6	1.50252263 52019168E -6				
		4.88906032 8226918E- 8	1.21598941 2877089E- 6				
14		3.19912438 63550855E -10	1.80146154 02589325E -7	7.41834207 3686358E- 7	8.76187522 3874993E- 7	NF	NF
		2.20645727 29643677E -6	1.55287564 23789271E -6				
		1.8725436 702903774 E-8	8.95540770 7576777E- 7				
15		1.09785246 93802138E -18	1.04308072 28002911E -11	5.90025290 2196242E- 8	7.02516841 1882005E- 7	NF	NF
		1.76482414 12298587E -7	1.87081917 10138786E -6				
		5.25172941 8035252E- 10	2.36720921 74349508E -7				
16		2.63422541 22225427E -10	1.66310452 18433931E -7	6.90680959 1735265E- 8	7.57767645 5122804E- 7	NF	NF


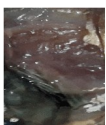
	1.89021920 9713321E- 10	1.80474112 89255654E -7				
	2.06751843 28986435E -7	1.92651837 1459945E- 6				

Table Information :

F = Fresh

NF = Not Fresh

Based on the test results using 50 data training with three parameters (Table 5.2) shows the results that 15 data successfully classified correctly and one data is wrong. And calculated using the precision-recall formula to get an accuracy value of 93,75% with the following calculation:

$$Accuracy = \frac{(TP+TN)}{(TP+FP+FN+TN)}$$

$$Accuracy = \frac{(8+7)}{(8+0+1+7)} = \frac{15}{16} = 93,75\%$$

Information :

True Positive (TP) = The prediction of fresh results is indeed fresh

True Negative (NF) = The predictions are not fresh are really not fresh

False Positive (FP) = Fresh predictions are not fresh

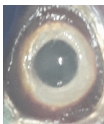



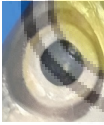
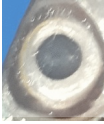


False Negative (FN) = Not Fresh predictions turn out to be fresh

## 5.2.2 Testing 2



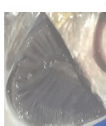
### a. Testing 1 Parameter

This is the second test table using one parameter:

Table 5.3: One parameter testing table

No	Input	Mean			Nilai Probabilitas		Image	Output
		Red	Green	Blue	F	NF		
1		122,58	117,25	110,33	3.37933594 0430306E-6	2.27489707 1668028E-7	F	F
2		143,75	148,64	148,64	4.28627709 364102E-8	1.21305658 56058015E-10	F	F
3		119,24	112,91	106,80	3.66945077 82633893E-6	3.92903857 48976725E-7	F	F
4		118,97	110,42	103,06	3.56055135 3281302E-6	5.23413546 2221825E-7	F	F
5		141,38	139,97	134,80	2.86334891 9661883E-7	1.90161869 12623885E-9	F	F
6		135,44	138,49	136,70	3.88608498 66623116E-7	2.64773276 24563457E-9	F	F
7		126,13	130,00	129,97	1.27903535 7225818E-6	1.85312024 95095344E-8	F	F
8		143,98	148,94	148,81	4.04347718 2445632E-8	1.12617400 47787663E-10	F	F



9		103,63	101,08	107,03	2.85500808 2946851E-6	9.51527726 1840812E-7	F	F
10		93,72	94,88	98,54	1.42141496 286225E-6	1.66362158 68516995E-6	F	NF
11		123,21	113,01	105,26	3.42067770 7838823E-6	3.54263192 01389736E-7	F	F
12		103,66	101,08	107,03	2.85822551 06701903E-6	9.50626517 9736463E-7	F	F
13		112,16	110,63	112,69	3.66978848 1888311E-6	4.06944194 32193757E-7	F	F
14		107,47	103,73	101,15	3.08173188 5729386E-6	1.03436284 47496624E-6	F	F
15		117,54	112,36	106,22	3.71569689 58554822E-6	4.41974105 1096645E-7	F	F
16		117,29	112,09	105,95	3.71670747 66717504E-6	4.57620297 2351579E-7	F	F
17		99,72	93,12	92,15	1.40353371 31530735E-6	1.90171184 53121962E-6	F	NF
18		135,76	131,23	124,05	1.08521918 0402322E-6	1.66934544 11368752E-8	F	F

19		111,70	106,76	101,10	3.39392900 31435476E-6	8.35326186 9651393E-7	F	F
20		96,92	91,24	86,35	8.82746228 7769126E-7	2.10115217 58180095E-6	F	NF
21		95,49	90,21	85,53	7.53767712 8132886E-7	2.14534107 22778695E-6	F	NF
22		96,16	89,78	87,91	8.62380511 179671E-7	2.13270023 3658448E-6	F	NF
23		97,62	91,19	88,95	1.03357172 1724134E-6	2.06512938 12638576E-6	F	NF
24		91,34	82,43	74,96	1.89538546 3113386E-7	1.98545936 83322443E-6	F	NF
25		98,97	55,25	65,93	8.43737946 2659222E-9	7.32435689 9048205E-7	NF	NF
26		107,20	104,52	94,43	2.60735919 16935685E-6	1.21529120 139388E-6	NF	F
27		93,21	89,90	91,53	8.86528728 132773E-7	2.07545382 754718E-6	NF	NF
28		99,96	94,28	100,70	1.90759553 1238902E-6	1.51328900 8760915E-6	NF	NF

29		76,47	21,67	63,69	5.67917497 8513428E- 12	4.39451794 6900631E-8	NF	NF
30		78,55	37,52	75,52	4.01304221 8977528E- 10	2.90688456 9971146E-7	NF	NF
31		36,19	10,28	22,03	2.34246459 0146485E- 18	2.97508924 2071751E- 11	NF	NF
32		78,08	36,89	75	3.31843790 85476253E- 10	2.70852344 5447614E-7	NF	NF
33		89,44	79,12	90,36	3.50653078 24965584E- 7	2.12062630 95825458E- 6	NF	NF
34		74,54	72,44	75,98	2.53107832 4333787E-7	2.02124159 09069954E- 6	NF	NF
35		69,67	71,69	80,79	2.09765541 33326017E- 8	1.31163154 86800596E- 6	NF	F
36		94,83	81,42	92,88	6.14710282 2968982E-7	2.08471006 16401105E- 6	NF	NF
37		77,61	82,15	90,10	1.76390213 73195795E- 7	1.84126238 63936435E- 6	NF	NF
38		99,72	99,33	105,34	2.38724173 0623541E-6	1.13546726 2833257E-6	NF	F

39		83,29	81,11	85,43	2.07728571 62241603E- 7	2.08381918 56831934E- 6	NF	NF
40		57,46	61,60	61,21	2.83766658 6024975E- 10	3.16277969 39754985E- 7	NF	NF
41		110,56	101,64	100,35	3.04056817 41232383E- 6	1.05389561 69352673E- 6	NF	F
42		95,48	82,30	76,03	2.53107832 4333787E-7	2.02124159 09069954E- 6	NF	NF
43		68,81	57,92	56,49	4.28497793 84694566E- 10	3.41348637 7621566E-7	NF	NF
44		116,95	103,47	97,47	3.00293466 03406796E- 6	8.87812813 8388672E-7	NF	F
45		95,49	90,21	85,53	5.45024277 0310348E-8	1.47531694 0984608E-6	NF	NF
46		91,63	74,57	67,48	2.16174177 88073328E- 8	1.17143313 13181722E- 6	NF	NF
47		71,25	61,06	55,18	6.84668283 1222191E- 10	3.86684511 07777994E- 7	NF	NF
48		90,95	82,46	77,79	2.31141542 97255555E- 7	2.09575759 92256605E- 6	NF	NF

Table Information :

F = Fresh

NF = Not Fresh

In the test table above, using 100 training data on one parameter the results are 36 data classified correctly and 12 data incorrectly. Therefore calculated using the precision-recall formula to get an accuracy value of 75% with the following calculation:

$$Accuracy = \frac{(TP+TN)}{(TP+FP+FN+TN)}$$

$$Accuracy = \frac{(17+19)}{(17+7+5+19)} = \frac{36}{48} = 75\%$$

Information :

True Positive (TP) = The prediction of fresh results is indeed fresh

True Negative (NF) = The predictions are not fresh are really not fresh

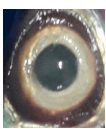

False Positive (FP) = Fresh predictions are not fresh



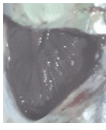
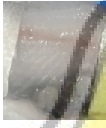

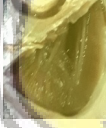
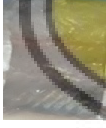
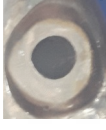


False Negative (FN) = Not Fresh predictions turn out to be fresh

#### b. Testing 3 Parameter

This is the second test table using one parameter:

Table 5.4: Three parameter testing table

No	Input	Probability Value		Probability Value		Image	Output
		F	NF	F	NF		
1		3.37933594 0430306E- 6	2.27489707 1668028E- 7	2.54595924 5510077E- 6	1.02690975 95543602E- 6	F	F
		2.85500808 2946851E- 6	9.51527726 1840812E- 7				

		1.40353371 31530735E -6	1.90171184 53121962E -6				
2		4.28627709 364102E-8	1.21305658 56058015E -10	1.25331707 1632286E- 6	8.40157520 6315341E- 7	F	F
		1.42141496 286225E-6	1.66362158 68516995E -6				
		1.08521918 0402322E- 6	1.66934544 11368752E -8				
3		3.66945077 82633893E -6	3.92903857 48976725E -7	3.49468582 9748586E- 6	5.27497745 4896014E- 7	F	F
		3.42067770 7838823E- 6	3.54263192 01389736E -7				
		3.39392900 31435476E -6	8.35326186 9651393E- 7				
4		3.56055135 3281302E- 6	5.23413546 2221825E- 7	2.43384103 09094685E -6	1.19173074 66712794E -6	F	F
		2.85822551 06701903E -6	9.50626517 9736463E- 7				
		8.82746228 7769126E- 7	2.10115217 58180095E -6				



5		2.86334891 9661883E- 7	1.90161869 12623885E -9	1.56996369 55559292E -6	8.51395628 4303565E- 7	F	F
		3.66978848 1888311E- 6	4.06944194 32193757E -7				
		7.53767712 8132886E- 7	2.14534107 22778695E -6				
6		3.88608498 66623116E -7	2.64773276 24563457E -9	1.44424029 8525096E- 6	1.05657027 0390189E- 6	F	F
		3.08173188 5729386E- 6	1.03436284 47496624E -6				
		8.62380511 179671E-7	2.13270023 3658448E- 6				
7		1.27903535 7225818E- 6	1.85312024 95095344E -8	2.00943465 8268478E- 6	8.41878229 6228725E- 7	F	F
		3.71569689 58554822E -6	4.41974105 1096645E- 7				
		1.03357172 1724134E- 6	2.06512938 12638576E -6				
8		4.04347718 2445632E- 8	1.12617400 47787663E -10	1.31556026 49358484E -6	8.14397427 65596E-7	F	F

		3.71670747 66717504E -6	4.57620297 2351579E- 7				
		1.89538546 3113386E- 7	1.98545936 83322443E -6				
9		8.43737946 2659222E- 9	7.32435689 9048205E- 7			NF	NF
		3.50653078 24965584E -7	2.12062630 95825458E -6				
		3.04056817 41232383E -6	1.05389561 69352673E -6				
10		2.60735919 16935685E -6	1.21529120 139388E-6			NF	NF
		2.53107832 4333787E- 7	2.02124159 09069954E -6				
		2.53107832 4333787E- 7	2.02124159 09069954E -6				
11		8.86528728 132773E-7	2.07545382 754718E-6	3.02644593 35331534E -7	1.24281133 79964656E -6	NF	NF
		2.09765541 33326017E -8	1.31163154 86800596E -6				

		4.28497793 84694566E-10	3.41348637 7621566E-7				
12		1.90759553 1238902E-6	1.51328900 8760915E-6				
		6.14710282 2968982E-7	2.08471006 16401105E-6	1.84174682 46254933E-6	1.49527062 80799644E-6	NF	F
		3.00293466 03406796E-6	8.87812813 8388672E-7				
13		5.67917497 8513428E-12	4.39451794 6900631E-8				
		1.76390213 73195795E-7	1.84126238 63936435E-6	7.69661068 7001331E-8	1.12017483 56157527E-6	NF	NF
		5.45024277 0310348E-8	1.47531694 0984608E-6				
14		4.01304221 8977528E-10	2.90688456 9971146E-7				
		2.38724173 0623541E-6	1.13546726 2833257E-6	8.03086817 544504E-7	8.65862950 382848E-7	NF	NF
		2.16174177 88073328E-8	1.17143313 13181722E-6				

15		2.34246459 0146485E- 18	2.97508924 2071751E- 11	6.94710799 6929357E- 8	8.23511149 217798E-7	NF	NF
		2.07728571 62241603E -7	2.08381918 56831934E -6				
		6.84668283 1222191E- 10	3.86684511 07777994E -7				
16		3.31843790 85476253E -10	2.70852344 5447614E- 7	7.72523844 7400427E- 8	8.94295971 0559905E- 7	NF	NF
		2.83766658 6024975E- 10	3.16277969 39754985E -7				
		2.31141542 97255555E -7	2.09575759 92256605E -6				

Table Information :

F = Fresh

NF = Not Fresh

When using 100 training data with three parameters, there are results that 15 data are classified as correct and only one data is incorrect. Then calculated using the precision-recall formula to get an accuracy value of 72,9% with the following calculation:

$$Accuracy = \frac{(TP+TN)}{(TP+FP+FN+TN)}$$

$$Accuracy = \frac{(8+7)}{(8+0+1+7)} = \frac{15}{16} = 93,75 \%$$

Information :

True Positive (TP) = The prediction of fresh results is indeed fresh

True Negative (NF) = The predictions are not fresh are really not fresh

False Positive (FP) = Fresh predictions are not fresh

False Negative (FN) = Not Fresh predictions turn out to be fresh




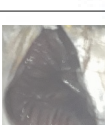
### 5.2.3 Testing 3

#### a. Testing 1 Parameter

This is the third test table using one parameter:

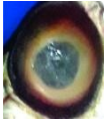
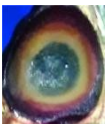

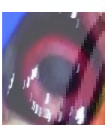
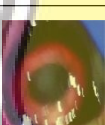

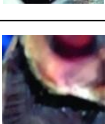
Table 5.5: One parameter testing table

No	Input	Mean			Nilai Probabilitas		Image	Output
		Red	Green	Blue	F	NF		
1		122,58	117,25	110,33	2,26614600 78120767E-6	1,29196180 47431114E-7	F	F
2		143,75	148,64	148,64	3,66405920 2158702E-8	1,03006382 77967571E-10	F	F
3		119,24	112,91	106,80	2,58877533 61064907E-6	2,25669279 44051961E-7	F	F
4		118,97	110,42	103,06	2,62258286 778355E-6	3,01503419 4122523E-7	F	F
5		141,38	139,97	134,80	1,99248276 4251493E-7	1,26101845 4590758E-9	F	F
6		135,44	138,49	136,70	2,69568025 88442557E-7	1,82621441 55191489E-9	F	F

7		126,13	130,00	129,97	8,34897172 1390833E-7	1,19148053 56843132E-8	F	F
8		143,98	148,94	148,81	3,47637662 9786519E-8	9,60748105 7144936E-11	F	F
9		103,63	101,08	107,03	2,42161188 57154485E-6	6,08700216 2434428E-7	F	F
10		93,72	94,88	98,54	1,53604773 31504138E-6	1,17730761 58176364E-6	F	F
11		123,21	113,01	105,26	2,40344089 93484096E-6	1,98878357 46300125E-7	F	F
12		103,66	101,08	107,03	2,42336590 22520835E-6	6,07964230 2760974E-7	F	F
13		112,16	110,63	112,69	2,63419758 9170889E-6	2,45505862 76304365E-7	F	F
14		107,47	103,73	101,15	2,59470846 6343718E-6	6,45555896 5583374E-7	F	F
15		117,54	112,36	106,22	2,65937168 9828407E-6	2,56372473 77009857E-7	F	F
16		117,29	112,09	105,95	2,67225738 90391966E-6	2,65828945 0965785E-7	F	F



17		99,72	93,12	92,15	1,56945407 33604508E-6	1,33430186 24243987E-6	F	F
18		135,76	131,23	124,05	6,93195107 1928658E-7	9,80453434 856808E-9	F	F
19		111,70	106,76	101,10	2,70995455 0657485E-6	5,05554576 587918E-7	F	F
20		96,92	91,24	86,35	1,13697976 47179923E-6	1,57537406 89033246E-6	F	NF
21		95,49	90,21	85,53	1,01393429 06688073E-6	1,64376415 92919894E-6	F	NF
22		96,16	89,78	87,91	1,11375237 44354133E-6	1,60062418 83988588E-6	F	NF
23		97,62	91,19	88,95	1,26848123 5751247E-6	1,51248570 3891169E-6	F	NF
24		91,34	82,43	74,96	3,69959387 22642526E-7	1,82485239 01282566E-6	F	NF
25		98,97	55,25	65,93	3,39177561 08352396E-8	9,53506415 1723339E-7	NF	NF
26		107,20	104,52	94,43	2,35607530 67404213E-6	7,76446262 1507177E-7	NF	F

27		93,21	89,90	91,53	1,12005489 19446743E- 6	1,55386037 27179556E- 6	NF	NF
28		99,96	94,28	100,70	1,89687150 04716493E- 6	1,02047854 5521369E-6	NF	F
29		76,47	21,67	63,69	1,01452311 69413389E- 10	1,28707713 0038304E-7	NF	NF
30		78,55	37,52	75,52	2,88337478 94103767E- 9	5,14774253 7074897E-7	NF	NF
31		36,19	10,28	22,03	8,04836841 882609E-16	4,52864360 7971705E- 10	NF	NF
32		78,08	36,89	75	2,48307553 1989153E-9	4,90110734 4788515E-7	NF	NF
33		89,44	79,12	90,36	5,59506512 0211251E-7	1,76075517 39070495E- 6	NF	NF
34		74,54	72,44	75,98	7,70014373 3581849E-8	1,61393869 00881787E- 6	NF	NF
35		69,67	71,69	80,79	6,23181878 7656558E-8	1,47032954 8109342E-6	NF	NF
36		94,83	81,42	92,88	8,51264779 4881545E-7	1,61168373 59936744E- 6	NF	NF

37		77,61	82,15	90,10	3,19591577 3972919E-7	1,64357838 76781743E-6	NF	NF
38		99,72	99,33	105,34	2,16262753 28577727E-6	7,49589096 4446559E-7	NF	F
39		83,29	81,11	85,43	3,77351517 52309327E-7	1,85139169 68682921E-6	NF	NF
40		57,46	61,60	61,21	2,17915839 32870472E-9	5,74441091 329135E-7	NF	NF
41		110,56	101,64	100,35	2,56848863 84450185E-6	6,48144667 6314135E-7	NF	F
42		95,48	82,30	76,03	4,60477587 0950364E-7	1,78122559 55314242E-6	NF	NF
43		68,81	57,92	56,49	3,27764660 33725315E-9	6,38919422 4344442E-7	NF	NF
44		116,95	103,47	97,47	2,47894234 51907386E-6	5,27321241 6552401E-7	NF	F
45		95,49	90,21	85,53	1,01393429 06688073E-6	1,64376415 92919894E-6	NF	NF
46		91,63	74,57	67,48	1,45922340 5136991E-7	1,59103919 8748477E-6	NF	NF



47		71,25	61,06	55,18	4,81244652 98827644E- 9	6,98651901 04038E-7	NF	NF
48		90,95	82,46	77,79	4,26610362 5320162E-7	1,86983513 87499258E- 6	NF	NF

Table Information :

F = Fresh

NF = Not Fresh

Based on table 5.5 test results using 150 training data on one parameter 38 data are successfully classified correctly and 10 data are incorrect. Therefore calculated using the precision-recall formula to get an accuracy value of 79,16% with the following calculation:

$$Accuracy = \frac{(TP+TN)}{(TP+FP+FN+TN)}$$

$$Accuracy = \frac{(19+19)}{(19+5+5+19)} = \frac{38}{48} = 79,16\% \quad \text{Information :}$$

True Positive (TP) = The prediction of fresh results is indeed fresh

True Negative (NF) = The predictions are not fresh are really not fresh

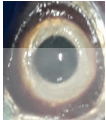

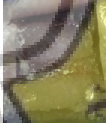

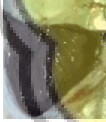

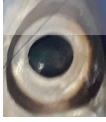


False Positive (FP) = Fresh predictions are not fresh

False Negative (FN) = Not Fresh predictions turn out to be fresh

## b. Testing 3 Parameter


This is the third test table using three parameter:

Table 5.6: Three parameter testing table

No	Input	Nilai Probabilitas		Nilai Probabilitas		Image	Output
		F	NF	F	NF		
1		2,26614600 78120767E -6	1,29196180 47431114E -7			F	F
		2,42161188 57154485E -6	6,08700216 2434428E- 7	2,08573732 22959917E -6	6,90732753 0473843E- 7		
		1,56945407 33604508E -6	1,33430186 24243987E -6				
2		3,66405920 2158702E- 8	1,03006382 77967571E -10			F	F
		1,53604773 31504138E -6	1,17730761 58176364E -6	7,55294477 4549557E- 7	3,95738385 51632803E -7		
		6,93195107 1928658E- 7	9,80453434 856808E-9				
3		2,58877533 61064907E -6	2,25669279 44051961E -7			F	F
		2,40344089 93484096E -6	1,98878357 46300125E -7	2,56739026 2037462E- 6	3,10034071 16381297E -7		
		2,70995455 0657485E- 6	5,05554576 587918E-7				

4		2,62258286 778355E-6	3,01503419 4122523E- 7	2,06097617 82512084E -6	8,28280572 8638915E- 7	F	F
		2,42336590 22520835E -6	6,07964230 2760974E- 7				
		1,13697976 47179923E -6	1,57537406 89033246E -6				
5		1,99248276 4251493E- 7	1,26101845 4590758E- 9	1,28246005 20882817E -6	6,30177013 5032078E- 7	F	F
		2,63419758 9170889E- 6	2,45505862 76304365E -7				
		1,01393429 06688073E -6	1,64376415 92919894E -6				
6		2,69568025 88442557E -7	1,82621441 55191489E -9	1,32600962 22211856E -6	7,49335433 1242385E- 7	F	F
		2,59470846 6343718E- 6	6,45555896 5583374E- 7				
		1,11375237 44354133E -6	1,60062418 83988588E -6				
7		8,34897172 1390833E- 7	1,19148053 56843132E -8	1,58758336 5906246E- 6	5,93590994 3393703E- 7	F	F



		2,65937168 9828407E- 6	2,56372473 77009857E -7				
		1,26848123 5751247E- 6	1,51248570 3891169E- 6				
8		3,47637662 9786519E- 8	9,60748105 7144936E- 11			F	F
		2,67225738 90391966E -6	2,65828945 0965785E- 7				
		3,69959387 22642526E -7	1,82485239 01282566E -6				
9		3,39177561 08352396E -8	9,53506415 1723339E- 7			NF	NF
		5,59506512 0211251E- 7	1,76075517 39070495E -6				
		2,56848863 84450185E -6	6,48144667 6314135E- 7				
10		2,35607530 67404213E -6	7,76446262 1507177E- 7	9,64518110 3904253E- 7	1,39053684 92567736E -6	NF	NF
		7,70014373 3581849E- 8	1,61393869 00881787E -6				

		4,60477587 0950364E- 7	1,78122559 55314242E -6				
11		1,12005489 19446743E -6	1,55386037 27179556E -6				
		6,23181878 7656558E- 8	1,47032954 8109342E- 6	3,95216908 80820413E -7	1,22103644 7753914E- 6	NF	NF
		3,27764660 33725315E -9	6,38919422 4344442E- 7				
12		1,89687150 04716493E -6	1,02047854 5521369E- 6				
		8,51264779 4881545E- 7	1,61168373 59936744E -6	1,74235954 17168475E -6	1,05316117 43900944E -6	NF	F
		2,47894234 51907386E -6	5,27321241 6552401E- 7				
13		1,01452311 69413389E -10	1,28707713 0038304E- 7				
		3,19591577 3972919E- 7	1,64357838 76781743E -6	1,55205123 4075617E- 7	1,12110843 31434939E -6	NF	NF
		1,01393429 06688073E -6	1,64376415 92919894E -6				

14		2,88337478 94103767E -9	5,14774253 7074897E- 7			NF	NF
		2,16262753 28577727E -6	7,49589096 4446559E- 7				
		1,45922340 5136991E- 7	1,59103919 8748477E- 6				
15		8,04836841 882609E- 16	4,52864360 7971705E- 10			NF	NF
		3,77351517 52309327E -7	1,85139169 68682921E -6				
		4,81244652 98827644E -9	6,98651901 04038E-7				
16		2,48307553 1989153E- 9	4,90110734 4788515E- 7			NF	NF
		2,17915839 32870472E -9	5,74441091 329135E-7				
		4,26610362 5320162E- 7	1,86983513 87499258E -6				

Information Table

F = Fresh

NF = Not Fresh

Based on the test results using 150 training data with three parameter (Table 5.6) shows the results showed that 15 data successfully classified correctly and one data is wrong. Therefore calculated using the precision-recall formula to get an accuracy value of 93,75% with the following calculation:

$$Accuracy = \frac{(TP+TN)}{(TP+FP+FN+TN)}$$

$$Accuracy = \frac{(8+7)}{(8+0+1+7)} = \frac{15}{16} = 93,75\%$$

Information :

True Positive (TP) = The prediction of fresh results is indeed fresh

True Negative (NF) = The predictions are not fresh are really not fresh

False Positive (FP) = Fresh predictions are not fresh

False Negative (FN) = Not Fresh predictions turn out to be fresh

#### 5.2.4 Diagram of analysis

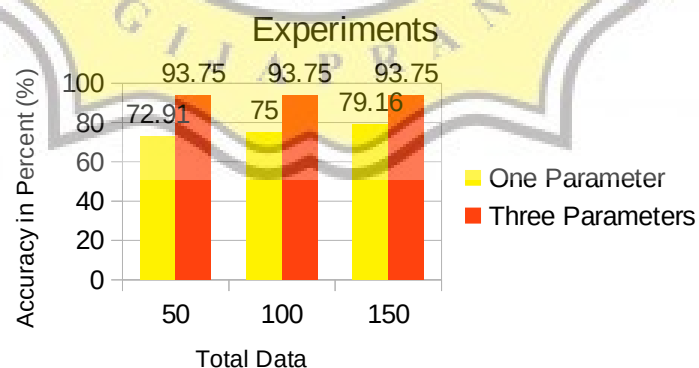


Illustration 5.7: Chart of Result

After doing three experiments using one parameter, illustration 5.6 gets an accuracy value of 72,9% with 50 training data. Then in the second experiment, the accuracy value of 100 training data is 75%. In the third experiment of 150 training

data, the results obtained 79,16% accuracy. Moreover, the three experiments using three parameters obtained the same value amounting to 93,75%.

Based on three experiments that use one parameter, several factors influence the algorithm's decision that the fish included in the either fresh or not fresh category, namely the mean value of R,G,B on each image, the standard deviation value, and the amount of in the training data. Furthermore, experiments using three parameters obtained the same accuracy value because each parameter's average probability value is sought in the experiment.

