

1 Matlab correction

1.1 Feature extraction

We make a loop on the whole database to extract some features of each image. The 9 features used here are: area, convex area, eccentricity, equivalent diameter, extent, major axis length, minor axis length, perimeter and solidity. All these parameters are defined in the documentation of the Matlab function `regionprops`.



```
% 18 classes of 12 images
2 folderImages = './images_Kimia216/';
   classes = {'bird','bone','brick','camel','car','children',...
4           'classic','elephant','face','fork','fountain',...
           'glass','hammer','heart','key','misk','ray','turtle'};
6 nbClasses = length(classes);
   nbImages = 12;

8
   features = [];
10 targets = zeros(nbClasses,nbClasses*nbImages);

12 for i=1:nbClasses
    for k=1:nbImages
14         nameImage = strcat(folderImages,classes{i},'-',num2str(k),'.bmp')
           ↳ ;
           currentImage = imread(nameImage);
16         currentImage = currentImage==0;
           s = regionprops(currentImage,'Area','ConvexArea',...
18                 'Eccentricity','EquivDiameter','Extent',...
                 'MajorAxisLength','MinorAxisLength',...
20                 'Perimeter','Solidity');
           sArray = [s.Area;s.ConvexArea;s.Eccentricity;...
22                 s.EquivDiameter;s.Extent;...
                 s.MajorAxisLength;s.MinorAxisLength;...
24                 s.Perimeter;s.Solidity];
           features = [features,sArray(:,1)];
26     end
           targets(i,(i-1)*nbImages+[1:nbImages]) = 1;
28 end
```

Note that in the same time, the target array (required in the following) is built within this loop.

1.2 Classification

The network is created with 10 hidden layers. We used a training set of 75% of the database and 25% for the test set.



```
% create the network
2 hiddenLayerSize = 10;
  net = patternnet(hiddenLayerSize);
4
% set up the division of data
6 net.divideParam.trainRatio = 75/100;
  %net.divideParam.valRatio   = 20/100;
8 net.divideParam.testRatio  = 25/100;
```

Now the network is trained and tested:



```
% train the network
2 [net,tr] = train(net,features,targets);
4
% test the network
  outputs = net(features);
```

The overall performance as well as the confusion matrix is here computed:



```
1 % overall performance
  [c,cm,ind,per] = confusion(targets,outputs);
3 perf = 1-c
5
% confusion matrix
  figure; plotconfusion(targets,outputs);
```

and the result is:

Command window

```
perf =
2      0.9444
```

Note that if we run the same code, the result can be different since the initialization of the optimization process (used for the training task) is different. For example, by running again, we get the following result:

Command window

```
perf =
2      0.9167
```

Confusion Matrix

Output Class	Target Class																	
	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15	16	17	18
1	12 5.6%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	1 0.5%	92.3% 7.7%
2	0 0.0%	12 5.6%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	100% 0.0%
3	0 0.0%	0 0.0%	11 5.1%	0 0.0%	0 0.0%	0 0.0%	1 0.5%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	91.7% 8.3%
4	0 0.0%	0 0.0%	0 0.0%	10 4.6%	0 0.0%	0 0.0%	2 0.9%	0 0.0%	0 0.0%	0 0.0%	1 0.5%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	1 0.5%	0 0.0%	71.4% 28.6%
5	0 0.0%	0 0.0%	1 0.5%	0 0.0%	12 5.6%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	92.3% 7.7%
6	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	12 5.6%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	100% 0.0%
7	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	10 4.6%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	100% 0.0%
8	0 0.0%	0 0.0%	0 0.0%	2 0.9%	0 0.0%	0 0.0%	0 0.0%	10 4.6%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	33.3% 16.7%
9	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	12 5.6%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	1 0.5%	0 0.0%	0 0.0%	0 0.0%	92.3% 7.7%
10	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	12 5.6%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	100% 0.0%
11	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	12 5.6%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	100% 0.0%
12	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	11 5.1%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	100% 0.0%
13	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	12 5.6%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	100% 0.0%
14	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	11 5.1%	0 0.0%	1 0.5%	0 0.0%	91.7% 8.3%
15	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	1 0.5%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	12 5.6%	0 0.0%	0 0.0%	92.3% 7.7%
16	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	11 5.1%	0 0.0%	100% 0.0%
17	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	11 5.1%	100% 0.0%
18	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	0 0.0%	100% 0.0%
	100% 0.0%	100% 0.0%	91.7% 8.3%	83.3% 16.7%	100% 0.0%	100% 0.0%	83.3% 16.7%	83.3% 16.7%	100% 0.0%	100% 0.0%	100% 0.0%	91.7% 8.3%	100% 0.0%	91.7% 8.3%	100% 0.0%	91.7% 8.3%	91.7% 8.3%	94.4% 5.6%

Figure 1: Confusion matrix of the classification result.