

## Machine Learning Worksheet 2

### Decision Trees and $k$ -Nearest Neighbors

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#### 1 Dataset

You are free to do the following exercises completely by hand or use a computer to help speed things up (python, MATLAB, R, Excel, ...). You should, however, show the basic steps of your work and implement your own “helpers” instead of blindly using code. Using a machine learning toolbox and copying the result will not help you understand.

The table below gives you a feature matrix  $\mathbf{X}$  together with the output  $z_i$  for every row  $i$  of the feature matrix. This data is also available in Piazza as `02_homework_dataset.csv`. The column with names for each datapoint  $i$  can help you reference specific data points.

$i$	$x_{i,1}$	$x_{i,2}$	$x_{i,3}$	$z_i$
A	5.5	0.5	4.5	2
B	7.4	1.1	3.6	0
C	5.9	0.2	3.4	2
D	9.9	0.1	0.8	0
E	6.9	-0.1	0.6	2
F	6.8	-0.3	5.1	2
G	4.1	0.3	5.1	1
H	1.3	-0.2	1.8	1
I	4.5	0.4	2.0	0
J	0.5	0.0	2.3	1
K	5.9	-0.1	4.4	0
L	9.3	-0.2	3.2	0
M	1.0	0.1	2.8	1
N	0.4	0.1	4.3	1
O	2.7	-0.5	4.2	1

#### 2 Decision Trees

**Problem 1:** Build a decision tree  $T$  for your data  $\mathbf{X}$ . Consider all possible feature tests and use the Gini index to build your tree. Build the tree only to a depth of two! Provide at least the value of the final Gini index at each node and the distribution of classes at each leaf.

**Problem 2:** Use the final tree  $T$  from the previous problem to classify the vectors  $\mathbf{x}_a = (4.1, -0.1, 2.2)^T$  and  $\mathbf{x}_b = (6.1, 0.4, 1.3)^T$ . Provide both your classification  $y_a$  and  $y_b$  and their respective probabilities  $p(c = y_a \mid \mathbf{x}_a, T)$  and  $p(c = y_b \mid \mathbf{x}_b, T)$

### 3 $k$ -Nearest Neighbor

**Problem 3:** Load the notebook `02_homework_kNN.ipynb` from piazza. Fill in the missing code and run the notebook. Convert the evaluated notebook to pdf and add it to the printout of your homework.

*Note: For information on IPython notebooks and how to convert them to other formats, consult the Jupyter documentation and nbconvert documentation.*

**Problem 4:** Classify the two vectors  $\mathbf{x}_a$  and  $\mathbf{x}_b$  given in Problem 2 with the  $k$ -nearest neighbors algorithm. Use  $k = 3$  and Euclidean distance.

**Problem 5:** Now, consider  $z_i$  to be real-valued labels rather than classes. Perform 3-NN regression to label the vectors from Problem 2.

**Problem 6:** Look at the data. Which problem do you see w.r.t. building a Euclidean distance-based  $k$ -NN model on  $\mathbf{X}$ ? How can you compensate for this problem? Does this problem also arise when training a decision tree?