Assignment 1. Solutions

1. Construct the root and the first level of a decision tree for the contact lenses data. Show the details of your construction. Then, check your solution with Weka (the data file is included with Weka).

For the root we have the following choices:

```
Age
```

young: 2/4/2 (8)

pre-presbyopic: 1/5/2 (8) presbyopic: 1/6/1 (8)

Spectacle-Prescription

myope: 3/7/2 (12)

hypermetrope: 1/8/3 (12)

Astigmatism

yes: 4/8/0 (12) no: 7/5/0 (12)

Tear-prod-rate

normal: 4/3/5 (12) reduced: 0/0/12 (12)

x/y/z means that we have x instances of some class, y instances of another class, and z instances of yet another class. The order x/y/z doesn't matter for computing entropies.

In the following I am computing the entropies using the natural logarithm l(.) (i.e. the one with base e). In order to get the values for logarithms with base 2, we have to divide with l(2). The reason I am not using directly logarithms with base 2 is that the calculators don't provide us log with base 2.

Age entropies:

```
entropy(2/4/2) = (-(2/8)*1(2/8)-(4/8)*1(4/8)-(2/8)*1(2/8))/1(2) = 1.5
entropy(1/5/2) = (-(1/8)*1(1/8)-(5/8)*1(5/8)-(2/8)*1(2/8))/1(2) = 1.3
entropy(1/6/1) = (-(1/8)*1(1/8)-(6/8)*1(6/8)-(1/8)*1(1/8))/1(2) = 1.06
avg_entropy = (8/24)*1.5 + (8/24)*1.3 + (8/24)*1.06 = 1.287 bits
```

Spectacle-Prescription entropies:

```
entropy(3/7/2) = (-(3/12)*l(3/12)-(7/12)*l(7/12)-(2/12)*l(2/12))/l(2) = 1.384
entropy(1/8/3) = (-(1/12)*l(1/12)-(8/12)*l(8/12)-(3/12)*l(3/12))/l(2) = 1.1887
avg_entropy = (12/24)*1.384 + (12/24)*1.1887 = 1.28635 bits
```

Astigmatism entropies:

```
entropy(4/8/0) = (-(4/12)*1(4/12)-(8/12)*1(8/12)-0)/1(2) = .918
```

```
entropy(7/5/0) = (-(7/12)*1(7/12)-(5/12)*1(5/12)-0)/1(2) = .9799 avg_entropy = (12/24)* .918 + (12/24)* .9799 = .94895 bits Tear-prod-rate entropies: entropy(4/3/5) = (-(4/12)*1(4/12)-(3/12)*1(3/12)-(5/12)*1(5/12))/1(2) = 1.555 entropy(0/0/12) = (0-0-0)/1(2) = 0 avg_entropy = (12/24)* 1.555 + (12/24)*0 = .7775 bits
```

The smallest average entropy is for Tear-prod-rate, so we choose it for the root. Now we have two branches (Tear-prod-rate=reduced) and (Tear-prod-rate=normal) The first branch ends up to be a leaf because all the instances going that branch are "contact-lenses = none".

For the other branch (Tear-prod-rate=normal) we have the following data instances:

| | spectacle- | | | |
|----------------|--------------|-------------|----------------|----------------|
| age | prescrip | astigmatism | tear-prod-rate | contact-lenses |
| pre-presbyopic | myope | yes | normal | hard |
| presbyopic | myope | yes | normal | hard |
| young | hypermetrope | yes | normal | hard |
| young | myope | yes | normal | hard |
| pre-presbyopic | hypermetrope | yes | normal | none |
| presbyopic | hypermetrope | yes | normal | none |
| presbyopic | myope | no | normal | none |
| pre-presbyopic | hypermetrope | no | normal | soft |
| pre-presbyopic | myope | no | normal | soft |
| presbyopic | hypermetrope | no | normal | soft |
| young | hypermetrope | no | normal | soft |
| young | myope | no | normal | soft |
| | | | | |

We have the following choices to break further:

Age

young: 2/2/0 (4)

pre-presbyopic: 1/1/2 (4) presbyopic: 1/2/1 (4)

Spectacle-Prescription

myope: 1/2/3 (6)

hypermetrope: 3/1/2 (6)

Astigmatism

yes: 4/2/0 (6) no: 1/5/0 (6)

Age entropies:

entropy(2/2/0) = (-(2/4)*1(2/4)-(2/4)*1(2/4)-0)/1(2) = .999

```
entropy(1/1/2) = (-(1/4)*1(1/4)-(1/4)*1(1/4)-(2/4)*1(2/4))/1(2) = 1.5 entropy(1/2/1) = (-(1/4)*1(1/4)-(2/4)*1(2/4)-0)/1(2) = .999 avg_entropy = (4/12)* .999 + (4/12)* 1.5 + (4/12)* .999 = 1.166 bits
```

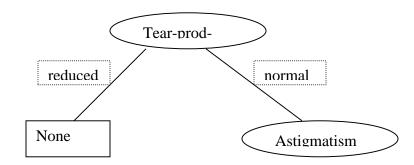
Spectacle-Prescription entropies:

```
entropy(1/2/3) = (-(1/6)*1(1/6)-(2/6)*1(2/6)-(3/6)*1(3/6))/1(2) = 1.459 entropy(3/1/2) = (-(3/6)*1(3/6)-(1/6)*1(1/6)-(2/6)*1(2/6))/1(2) = 1.459 avg_entropy = (6/12)*1.459+(6/12)*1.459=1.459 bits
```

Astigmatism entropies:

entropy(
$$4/2/0$$
) = $(-(4/6)*1(4/6)-(2/6)*1(2/6)-0)/1(2)$ = .918
entropy($1/5/0$) = $(-(1/6)*1(1/6)-(5/6)*1(5/6)-0)/1(2)$ = .65
avg_entropy = $(12/24)*.918 + (12/24)*.65 = .784$ bits

So, we choose astigmatism for the next level. The tree so far is:



2. Construct two rules using PRISM for the weather data. Show the details of your construction. Then, check your solution with Weka.

Rule we seek: If? then play=yes

(Of course, we could have alternatively started with **If? then play=no**. Actually, according to the heuristic mentioned in the PRISM pseudocode, we should start with the latter. However, in order to be compatible with WEKA, I am starting with the former. WEKA unfortunately ignores the heuristic. Both solutions will be considered fine when marking this question.)

| Outlook=sunny | 2/5 |
|------------------|-----|
| Outlook=overcast | 4/4 |
| Outlook=rainy | 3/5 |
| | |
| Temp=cool | 3/4 |
| Temp=mild | 4/6 |

| Temp=hot | 2/4 | |
|-----------------|-----|--|
| Humidity=normal | 6/7 | |
| Humidity=high | 3/7 | |
| Windy=true | 3/6 | |
| Windy=false | 6/8 | |

Best accuracy Outlook=overcast 4/4, so we choose it as a condition:

If Outlook=overcast then play=yes

We consider the instances covered by this rule, i.e. those that make the antecedent true. We observe that all them have play=yes, so we are done with the first rule.

Now, let's consider the instances not covered by this rule.

| Outlook | Temperature Humidity | | Windy | Play |
|---------|----------------------|--------|-------|------|
| rainy | mild | high | TRUE | no |
| rainy | cool | normal | TRUE | no |
| rainy | mild | high | FALSE | yes |
| rainy | cool | normal | FALSE | yes |
| rainy | mild | normal | FALSE | yes |
| sunny | hot | high | FALSE | no |
| sunny | mild | high | FALSE | No |
| sunny | hot | high | TRUE | No |
| sunny | cool | normal | FALSE | yes |
| sunny | mild | normal | TRUE | yes |

Rule we seek: If? then play=yes

| Outlook=sunny | 2/5 | |
|-----------------|-----|--|
| Outlook=rainy | 3/5 | |
| Temp=cool | 2/3 | |
| Temp=mild | 3/5 | |
| Temp=hot | 0/2 | |
| Humidity=normal | 4/5 | |
| Humidity=high | 1/5 | |
| Windy=true | 1/4 | |
| Windy=false | 4/6 | |

The best to choose is Humidity=normal 4/5, so we get

If Humidity=normal then play=yes

The covered instances (i.e. those that satisfy the antecedent) are:

| outlook | tempera | ture humidity | windy | play |
|---------|---------|---------------|-------|------|
| rainy | cool | normal | TRUE | no |
| rainy | cool | normal | FALSE | yes |
| sunny | cool | normal | FALSE | yes |
| rainy | mild | normal | FALSE | yes |
| sunny | mild | normal | TRUE | yes |

It's a pretty good rule and we might want to stop here, but let's say we want full accuracy.

We want to extend the rule as: If Humidity=normal and? then play=yes

| Outlook=sunny | 2/2 |
|--------------------|-----------|
| Outlook=rainy | 2/3 |
| | 2 (2 |
| Temp=cool | 2/3 |
| Temp=mild | 2/2 |
| Temp=hot | undefined |
| Windy=true | 1/2 |
| Windy=false | 3/3 |
| 11 111d y —1 1115C | 515 |

Between Temp=mild and Windy=false we choose Windy=false because it has greater coverage. So our second rule becomes:

If Humidity=normal and Windy=false then play=yes

We consider the instances covered by this rule, i.e. those that make the antecedent true. We observe that all them have play=yes, so we are done with the second rule.

3. (4 points) Classify using Naïve Bayes method (on contact lenses data) the data item: *pre-presbyopic, hypermetrope, yes, reduced,*?

$$\begin{split} &P(Hard|E) = (1+1)/(4+3) * (1+1)/(4+2) * (4+1)/(4+2) * (0+1)/(4+2) * (4+1)/(24+3) \\ &= alpha * .00244953948657652361 \\ &P(Soft|E) = (2+1)/(5+3) * (3+1)/(5+2) * (0+1)/(5+2) * (0+1)/(5+2) * (5+1)/(24+3) \\ &= alpha * .00097181729834791059 \end{split}$$

```
P(None|E) = (5+1)/(15+3) * (8+1)/(15+2) * (8+1)/(15+2) * (12+1)/(15+2) * (15+1)/(24+3) = alpha * .04233665784652961530
```

 $alpha = 1/(.00244953948657652361 + .00097181729834791059 \\ + .04233665784652961530) = 21.85409502694201713124$

P(Hard|E) = .00244953948657652361 * 21.85409502694201713124 = .05353246871189010655

~ 0.054

P(Soft|E) = .00097181729834791059 * 21.85409502694201713124 = .02123818758692129938 ~ 0.021

P(None|E) = .04233665784652961530 * 21.85409502694201713124 = .92522934370118859406 ~ 0.925