

### ASSIGNMENT 3

Assignment Date	19 September 2022
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Maximum Marks	2 Marks

## Artificial intelligence

### A startlingly effective assistant

I was given early “preview” access to Copilot about a year ago, and I’ve been using it on and off. It takes some practice to learn exactly how to frame your requests in English so the Copilot AI gives the most useful code output, but it can be startlingly effective.

However, we’re still a *long* way from “Hey Siri, make me a million dollar iPhone app”. It’s still necessary to use my software design skills to figure out what the different bits of code should do in my app.

To understand the level Copilot is working at, imagine writing an essay. You can’t just throw the essay question at it and expect it to produce a useful, well-argued piece. But if you figure out the argument and maybe write the topic sentence for each paragraph, it will often do a pretty good job at filling in the rest of each paragraph automatically.

Depending on the type of coding I’m doing, this can sometimes be a huge time- and brainpower-saver.

Out[1]:

In [2]:

Out[2]:

	school	sex	age	addresses	family size	status	married	Fedu	Mjob	Fjob	...	interest	romantic	family	frequency	group	Dalc	Walc	health	abuses	parents
0	GP	F	18	U	GT3	A	4	4	at_home	teacher	...	no	no	4	3	4	1	1	3	4	0
1	GP	F	17	U	GT3	T	1	1	at_home	other	...	yes	no	5	3	3	1	1	3	2	0
2	GP	F	15	U	LE3	T	1	1	at_home	other	...	yes	no	4	3	2	2	3	3	6	1
3	GP	F	15	U	GT3	T	4	2	health	services	...	yes	yes	3	2	2	1	1	5	0	1

```

sc h s a ad fa Ps M F int ro fa fre g D W h ab p
h o e g dr m ta e e Mj Fj . er ma m re o a l c e al se a
o o x e s s e s u u ob ob . ne nti rel e th nc s
ol ol

4 G F 1 U GT ot ot .
P 6 3 T 3 he he .
r r .
no no 4 3 2 1 2 5 0 1

```

5 rows × 31 columns

In [3]:

```

# use one-hot encoding on categorical columns
d = pd.get_dummies(d, columns=['sex', 'school', 'address', 'famsize', 'Pstatus', 'Mjob', 'Fjob',
                              'reason', 'guardian', 'schoolsup', 'famsup', 'paid', 'activities',
                              'nursery', 'higher', 'internet', 'romantic'])
d.head()

```

Out[3]:

```

a M F tr st f f fr g D . act act nu nu hi hi int int ro ro
g e d e av u ai l a e e o a . ivi ivit rs rs er er ne ne m ma
e u u e el dy ti l r e r et i o l c . tie ies _y _y _n _n _n _n _n _n
s e e m m e s l e m e t c . s _yes _no _yes _no _no _no _no _no _no _no
yes

0 1 4 4 2 2 0 4 3 4 1 . 1 0 0 1 0 1 1 0 1 0
8

1 1 1 1 2 0 5 3 3 1 . 1 0 1 0 0 1 0 1 1 0
7

2 1 1 1 2 0 4 3 2 2 . 1 0 0 1 0 1 0 1 1 0
5

3 1 4 2 1 3 0 3 2 2 1 . 0 1 0 1 0 1 0 0 1
5

```

	age	gender	gender	average	study	failures	family	freem	group	Dalc	activities	activities	nursery	nursery	high	high	internet	internet	romantic	romantic
4	16	3	3	1	2	0	4	3	2	1	1	0	0	1	0	1	1	0	1	0

5 rows × 57 columns

In [4]:

```
# shuffle rows
d = d.sample(frac=1)
# split training and testing data
d_train = d[:500]
d_test = d[500:]

d_train_att = d_train.drop(['pass'], axis=1)
d_train_pass = d_train['pass']

d_test_att = d_test.drop(['pass'], axis=1)
d_test_pass = d_test['pass']

d_att = d.drop(['pass'], axis=1)
d_pass = d['pass']

# number of passing students in whole dataset:
import numpy as np
print("Passing: %d out of %d (%.2f%%)" % (np.sum(d_pass), len(d_pass), 100*float(np.sum(d_pass)) / len(d_pass)))
Passing: 328 out of 649 (50.54%)
```

In [5]:

```
# fit a decision tree
from sklearn import tree
t = tree.DecisionTreeClassifier(criterion="entropy", max_depth=5)
t = t.fit(d_train_att, d_train_pass)
```

In [6]:

```
# visualize tree
import graphviz
```

```
dot_data = tree.export_graphviz(t, out_file=None, label="all", impurity=False, proportion=True,
                                feature_names=list(d_train_att), class_names=["fail", "pass"],
                                filled=True, rounded=True)
graph = graphviz.Source(dot_data)
graph
```

Out[6]:

```
b'\n'
```

In [7]:

```
# save tree
tree.export_graphviz(t, out_file="student-performance.dot", label="all", impurity=False,
                    proportion=True,
                    feature_names=list(d_train_att), class_names=["fail", "pass"],
                    filled=True, rounded=True)
```

In [8]:

```
t.score(d_test_att, d_test_pass)
```

Out[8]:

```
0.59731543624161076
```

In [9]:

```
from sklearn.model_selection import cross_val_score
scores = cross_val_score(t, d_att, d_pass, cv=5)
# show average score and +/- two standard deviations away (covering 95% of scores)
print("Accuracy: %0.2f (+/- %0.2f)" % (scores.mean(), scores.std() * 2))
Accuracy: 0.67 (+/- 0.06)
```

In [10]:

```
for max_depth in range(1, 20):
    t = tree.DecisionTreeClassifier(criterion="entropy", max_depth=max_depth)
    scores = cross_val_score(t, d_att, d_pass, cv=5)
    print("Max depth: %d, Accuracy: %0.2f (+/- %0.2f)" % (max_depth, scores.mean(), scores.std() * 2))
```

```
Max depth: 1, Accuracy: 0.64 (+/- 0.05)
Max depth: 2, Accuracy: 0.69 (+/- 0.08)
Max depth: 3, Accuracy: 0.69 (+/- 0.09)
Max depth: 4, Accuracy: 0.66 (+/- 0.10)
Max depth: 5, Accuracy: 0.67 (+/- 0.06)
Max depth: 6, Accuracy: 0.64 (+/- 0.08)
Max depth: 7, Accuracy: 0.67 (+/- 0.02)
Max depth: 8, Accuracy: 0.67 (+/- 0.07)
Max depth: 9, Accuracy: 0.67 (+/- 0.06)
Max depth: 10, Accuracy: 0.63 (+/- 0.12)
Max depth: 11, Accuracy: 0.65 (+/- 0.07)
Max depth: 12, Accuracy: 0.63 (+/- 0.07)
Max depth: 13, Accuracy: 0.63 (+/- 0.07)
Max depth: 14, Accuracy: 0.63 (+/- 0.08)
Max depth: 15, Accuracy: 0.64 (+/- 0.06)
Max depth: 16, Accuracy: 0.62 (+/- 0.05)
Max depth: 17, Accuracy: 0.64 (+/- 0.09)
Max depth: 18, Accuracy: 0.63 (+/- 0.08)
```

Max depth: 19, Accuracy: 0.63 (+/- 0.06)

In [11]:

```
depth_acc = np.empty((19,3), float)
i = 0
for max_depth in range(1, 20):
    t = tree.DecisionTreeClassifier(criterion="entropy", max_depth=max_depth)
    scores = cross_val_score(t, d_att, d_pass, cv=5)
    depth_acc[i,0] = max_depth
    depth_acc[i,1] = scores.mean()
    depth_acc[i,2] = scores.std() * 2
    i += 1
```

depth\_acc

Out[11]:

```
array([[ 1.    , 0.63790456, 0.04848398],
       [ 2.    , 0.68559869, 0.07148267],
       [ 3.    , 0.68710174, 0.0865951 ],
       [ 4.    , 0.6669467 , 0.10726248],
       [ 5.    , 0.66261518, 0.05307124],
       [ 6.    , 0.65018859, 0.07040891],
       [ 7.    , 0.66564494, 0.02029519],
       [ 8.    , 0.67474598, 0.05984916],
       [ 9.    , 0.6640118 , 0.03746891],
       [10.    , 0.6346137 , 0.09657669],
       [11.    , 0.6484015 , 0.10475147],
       [12.    , 0.64545485, 0.05529647],
       [13.    , 0.64544256, 0.08167465],
       [14.    , 0.6346614 , 0.07458128],
       [15.    , 0.63463773, 0.08162646],
       [16.    , 0.62853141, 0.05926906],
       [17.    , 0.63622335, 0.05390067],
       [18.    , 0.62548936, 0.06050112],
       [19.    , 0.63004547, 0.07022296]])
```

In [12]:

```
import matplotlib.pyplot as plt
fig, ax = plt.subplots()
ax.errorbar(depth_acc[:,0], depth_acc[:,1], yerr=depth_acc[:,2])
plt.show()
```

In [ ]