

In [1]:

```
import pandas as pd
import numpy as np
from matplotlib import pyplot as plt
from sklearn.model_selection import GridSearchCV
from sklearn.metrics import accuracy_score
from sklearn.metrics import roc_auc_score
#显示中文
plt.rcParams['font.sans-serif'] = ['Arial Unicode MS']
```

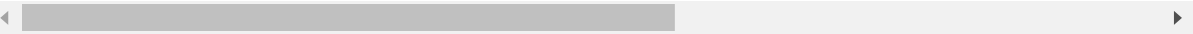
In [2]:

```
dpath = "./data/"
df = pd.read_csv(dpath + "mushrooms.csv")
df.head()
```

Out[2]:

	class	cap- shape	cap- surface	cap- color	bruises	odor	gill- attachment	gill- spacing	gill- size	gill- color	...	stalk- surface- below- ring
0	p	x	s	n	t	p	f	c	n	k	...	s
1	e	x	s	y	t	a	f	c	b	k	...	s
2	e	b	s	w	t	l	f	c	b	n	...	s
3	p	x	y	w	t	p	f	c	n	n	...	s
4	e	x	s	g	f	n	f	w	b	k	...	s

5 rows × 23 columns



In [3]:

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 8124 entries, 0 to 8123
Data columns (total 23 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   class                                8124 non-null   object
1   cap-shape                            8124 non-null   object
2   cap-surface                          8124 non-null   object
3   cap-color                           8124 non-null   object
4   bruises                             8124 non-null   object
5   odor                                8124 non-null   object
6   gill-attachment                      8124 non-null   object
7   gill-spacing                        8124 non-null   object
8   gill-size                           8124 non-null   object
9   gill-color                          8124 non-null   object
10  stalk-shape                         8124 non-null   object
11  stalk-root                          8124 non-null   object
12  stalk-surface-above-ring            8124 non-null   object
13  stalk-surface-below-ring            8124 non-null   object
14  stalk-color-above-ring              8124 non-null   object
15  stalk-color-below-ring              8124 non-null   object
16  veil-type                           8124 non-null   object
17  veil-color                          8124 non-null   object
18  ring-number                         8124 non-null   object
19  ring-type                           8124 non-null   object
20  spore-print-color                   8124 non-null   object
21  population                          8124 non-null   object
22  habitat                             8124 non-null   object
dtypes: object(23)
memory usage: 1.4+ MB
```

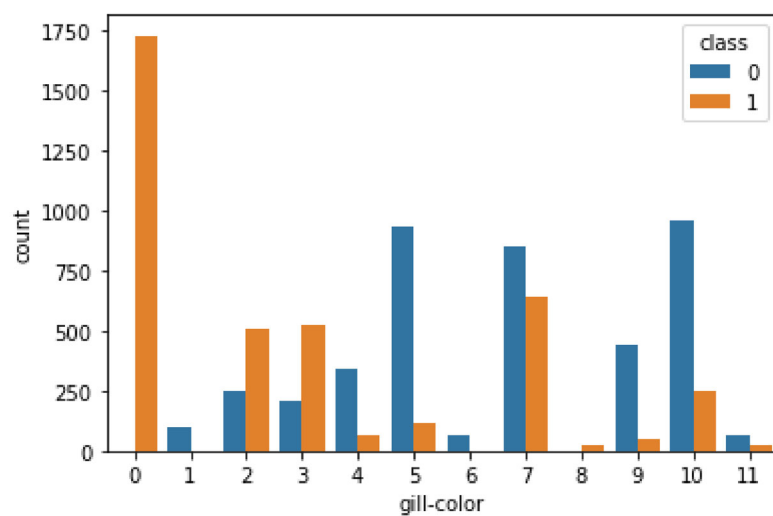
In [4]:

```
from sklearn.preprocessing import LabelEncoder
labelencoder=LabelEncoder()
for col in df.columns:
    df[col] = labelencoder.fit_transform(df[col])
df.head()
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
sns.countplot(x="gill-color", hue="class", data=df)
```

Out[4]:

<AxesSubplot:xlabel='gill-color', ylabel='count'>

findfont: Font family ['sans-serif'] not found. Falling back to DejaVu Sans.



In [5]:

```
#数据集是一个文件，我们自己分出一部分来做测试吧（不是校验集）
y = df['class']
X = df.drop('class', axis = 1)
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=4)
columns = X_train.columns
#体验一下分裂及gini 计算
df1 = pd.DataFrame({'gill-color':X_train['gill-color'], 'class':y_train})
df1.sort_values(by=['gill-color'], ascending=True)
gini=[]
for i in range(11) :
    threshold = i + 0.5
    dl = df[df['gill-color'] <= threshold]
    dr = df[df['gill-color'] > threshold]
    dl_0 = (dl['class'] == 0).sum()
    dl_1 = (dl['class'] == 1).sum()
    dr_0 = (dr['class'] == 0).sum()
    dr_1 = (dr['class'] == 1).sum()
    total_l = dl_0 + dl_1
    total_r = dr_0 + dr_1
    total = total_l + total_r
    h_l = 2.0 * dl_0/total_l * dl_1/total_l
    h_r = 2.0 * dr_0/total_r * dr_1/total_r
    gini.append(h_l*total_l/total + h_r*total_r/total)
print(gini)
```

```
[0.3543844641946022, 0.3739663947329344, 0.3621154280725757, 0.32883812359574915, 0.
36911839170001565, 0.4482379747002899, 0.452776520958944, 0.4501570405964006, 0.4473
229852206424, 0.47107276743242, 0.49825903532938304]
```

In [6]:

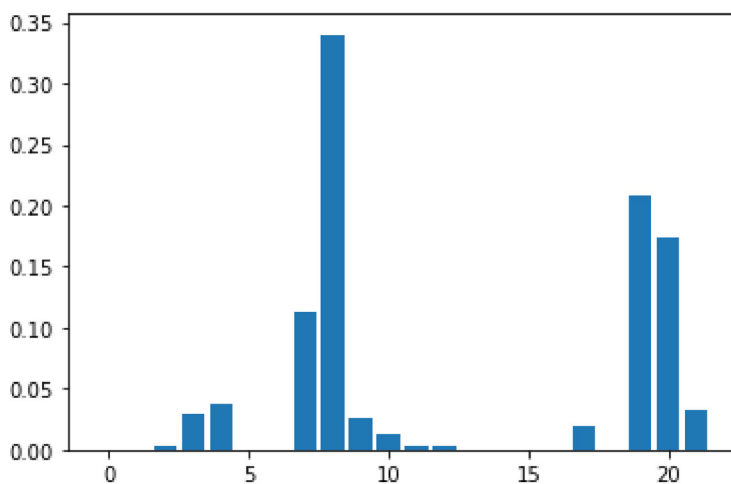
#默认参数的决策树

```

from sklearn.tree import DecisionTreeClassifier
model_tree = DecisionTreeClassifier()
model_tree.fit(X_train, y_train)
y_prob = model_tree.predict_proba(X_test)[:,-1]
# This will give you positive class prediction probabilities
y_pred = np.where(y_prob > 0.5, 1, 0)
# This will threshold the probabilities to give class predictions.
model_tree.score(X_test, y_pred)
print('The AUC of default Desicion Tree is', roc_auc_score(y_test, y_pred))
df = pd.DataFrame({"columns":list(columns), "importance":list(model_tree.feature_importances_.T)})
df.sort_values(by=['importance'], ascending=False)
plt.bar(range(len(model_tree.feature_importances_)), model_tree.feature_importances_)
plt.show()

```

The AUC of default Desicion Tree is 1.0



In [7]:

#超参数调优

```

from sklearn.tree import DecisionTreeClassifier
model_DD = DecisionTreeClassifier()
max_depth = range(3, 10, 1)
min_samples_leaf = range(5, 15, 1)
tuned_parameters = dict(max_depth=max_depth, min_samples_leaf=min_samples_leaf)

```

In [8]:

```

from sklearn.model_selection import GridSearchCV
gird = GridSearchCV(model_DD, tuned_parameters, scoring='roc_auc', cv=10)
gird.fit(X_train, y_train)
print("Best: %f using %s" % (gird.best_score_, gird.best_params_))
y_prob = gird.best_estimator_.predict_proba(X_test)[:,-1]
# This will give you positive class prediction probabilities
y_pred = np.where(y_prob > 0.5, 1, 0)
# This will threshold the probabilities to give class predictions.
gird.score(X_test, y_pred)

```

Best: 1.000000 using {'max_depth': 7, 'min_samples_leaf': 5}

Out[8]:

1.0

In [9]:

```
print ('The AUC of GridSearchCV Desicion Tree is', roc_auc_score(y_test,y_pred))
import graphviz
import pydotplus
from sklearn import tree
#dotfile = StringIO.StringIO()
dot_data = tree.export_graphviz(gird.best_estimator_, out_file='best_tree.dot',feature_names=columns
!dot -Tpng best_tree.dot -o best_tree.png
```

The AUC of GridSearchCV Desicion Tree is 1.0