```
In [139]: import numpy as np
   import pandas as pd
   import sklearn
   from sklearn.utils import shuffle
   from sklearn.neighbors import KNeighborsClassifier
   from sklearn import linear_model
   from sklearn import preprocessing
   from sklearn.model_selection import train_test_split
```

Out[140]:

	buying	maint	door	persons	lug_boot	safety	class
0	vhigh	vhigh	2	2	small	low	unacc
1	vhigh	vhigh	2	2	small	med	unacc
2	vhigh	vhigh	2	2	small	high	unacc
3	vhigh	vhigh	2	2	med	low	unacc
4	vhigh	vhigh	2	2	med	med	unacc
1723	low	low	5more	more	med	med	good
1724	low	low	5more	more	med	high	vgood
1725	low	low	5more	more	big	low	unacc
1726	low	low	5more	more	big	med	good
1727	low	low	5more	more	big	high	vgood

1728 rows × 7 columns

```
In [141]: len(data)
```

Out[141]: 1728

```
In [142]: data.dtypes
```

Out[142]: buying object maint object door object persons object lug_boot object safety object class object dtype: object

In [143]: data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1728 entries, 0 to 1727
Data columns (total 7 columns):

#	Column	Non-Null Count	Dtype
0	buying	1728 non-null	object
1	maint	1728 non-null	object
2	door	1728 non-null	object
3	persons	1728 non-null	object
4	lug_boot	1728 non-null	object
5	safety	1728 non-null	object
6	class	1728 non-null	object

dtypes: object(7)
memory usage: 94.6+ KB

In [144]: data.describe()

Out[144]:

	buying	maint	door	persons	lug_boot	safety	class
count	1728	1728	1728	1728	1728	1728	1728
unique	4	4	4	3	3	3	4
top	vhigh	vhigh	2	2	small	low	unacc
freq	432	432	432	576	576	576	1210

In [145]: data.corr()

Out[145]:

In [146]: data.head()

Out[146]:

	buying	maint	door	persons	lug_boot	safety	class
0	vhigh	vhigh	2	2	small	low	unacc
1	vhigh	vhigh	2	2	small	med	unacc
2	vhigh	vhigh	2	2	small	high	unacc
3	vhigh	vhigh	2	2	med	low	unacc
4	vhiah	vhiah	2	2	med	med	unacc

```
In [147]: data.tail()
Out[147]:
                   buying maint
                                   door persons lug_boot safety
                                                                   class
             1723
                      low
                             low
                                 5more
                                                      med
                                                             med
                                                                    good
                                            more
             1724
                                 5more
                                                      med
                                                             high
                                                                   vgood
                      low
                             low
                                            more
             1725
                                 5more
                                                       big
                                                              low
                                                                   unacc
                      low
                             low
                                            more
             1726
                                 5more
                                                       big
                                                             med
                      ow
                             low
                                            more
                                                                    good
             1727
                      low
                             low
                                 5more
                                            more
                                                       big
                                                             high vgood
In [148]: data.index
Out[148]: RangeIndex(start=0, stop=1728, step=1)
In [149]: data.isna()
Out[149]:
                           maint door persons lug_boot safety class
                   buying
                0
                    False
                           False False
                                                            False
                                                                  False
                                           False
                                                    False
                1
                    False
                           False False
                                           False
                                                    False
                                                            False
                                                                  False
                2
                    False
                           False False
                                           False
                                                    False
                                                            False
                                                                 False
                3
                    False
                           False False
                                           False
                                                    False
                                                            False
                                                                  False
                4
                    False
                           False False
                                           False
                                                    False
                                                            False False
             1723
                    False
                           False
                                 False
                                           False
                                                    False
                                                            False
                                                                  False
             1724
                    False
                           False
                                 False
                                           False
                                                    False
                                                            False
                                                                  False
             1725
                    False
                           False False
                                                            False
                                                                 False
                                           False
                                                    False
             1726
                    False
                           False False
                                           False
                                                    False
                                                            False
                                                                  False
             1727
                    False False
                                           False
                                                    False
                                                            False False
            1728 rows × 7 columns
In [150]: data.isna().sum()
Out[150]:
            buying
                          0
            maint
                          0
            door
                          0
            persons
                          0
            lug_boot
                          0
            safety
            class
            dtype: int64
```

```
In [151]: le = preprocessing.LabelEncoder()
In [152]:
          buying = le.fit_transform(list(data['buying']))
          maint = le.fit_transform(list(data['maint']))
          door = le.fit_transform(list(data['door']))
          persons = le.fit_transform(list(data['persons']))
          lug_boot = le.fit_transform(list(data['lug_boot']))
          safety = le.fit_transform(list(data['safety']))
          cls = le.fit_transform(list(data['class']))
In [153]: |print(buying)
          [3 3 3 ... 1 1 1]
In [154]: | predict = 'class'
In [155]: | x=list(zip(buying, maint,door,persons,lug boot,safety))
          y=list(cls)
In [156]: | x_train, x_test, y_train, y_test = train_test_split(x,y,test_size=0.1)
In [157]: | model = KNeighborsClassifier(n neighbors=9)
In [158]: |model.fit(x train,y train)
Out[158]: KNeighborsClassifier(n neighbors=9)
In [159]: | acc = model.score(x_test,y_test)
          print(acc)
          0.953757225433526
          C:\Users\Noah\anaconda3\lib\site-packages\sklearn\neighbors\_classification.p
          y:228: FutureWarning: Unlike other reduction functions (e.g. `skew`, `kurtosi
          s`), the default behavior of `mode` typically preserves the axis it acts alon
          g. In SciPy 1.11.0, this behavior will change: the default value of `keepdims
            will become False, the `axis` over which the statistic is taken will be eli
```

minated, and the value None will no longer be accepted. Set `keepdims` to Tru

e or False to avoid this warning.

mode, _ = stats.mode(_y[neigh_ind, k], axis=1)

LabelEncoder

```
In [160]: predicted = model.predict(x test)
         C:\Users\Noah\anaconda3\lib\site-packages\sklearn\neighbors\_classification.p
         y:228: FutureWarning: Unlike other reduction functions (e.g. `skew`, `kurtosi
         s`), the default behavior of `mode` typically preserves the axis it acts alon
          g. In SciPy 1.11.0, this behavior will change: the default value of `keepdims
          `will become False, the `axis` over which the statistic is taken will be eli
         minated, and the value None will no longer be accepted. Set `keepdims` to Tru
         e or False to avoid this warning.
           mode, _ = stats.mode(_y[neigh_ind, k], axis=1)
In [161]: names = ["unacc","acc","good","vgood"]
In [165]: for x in range(len(predicted)):
             print("Predicted: ", names[predicted[x]], "Data: ", x_test[x],"Actual: ",
             n = model.kneighbors(([x test[x]]),9,True)
             print(n)
                                 , 1. , 1.
          (array([[1.
                            , 1.
                                                              , 1.
                                     , 1.41421356, 1.41421356]]), array([[ 501,
                 1.
                           , 1.
          326, 710, 1261, 978, 863, 418, 299, 1364]],
               dtype=int64))
         Predicted: good Data: (2, 3, 1, 1, 1, 1) Actual: good
          (array([[1., 1., 1., 1., 1., 1., 1., 1.]]), array([[ 378, 1513, 1484,
         1142, 765, 1304, 1526, 815, 82]],
               dtype=int64))
         Predicted: good Data: (2, 0, 0, 2, 1, 1) Actual: good
          (array([[1., 1., 1., 1., 1., 1., 1., 1.]]), array([[ 986, 706, 1365,
         153, 1028, 1349, 1489, 99, 868]],
               dtype=int64))
         Predicted: good Data: (0, 0, 1, 0, 1, 1) Actual: good
          (array([[1.
                           , 1.
                                       , 1. , 1.
                                                              , 1.
                                      , 1.
                           , 1.
                                             , 1.41421356]]), array([[ 519,
          302, 1067, 899, 1481, 310, 628, 79, 1259]],
               dtype=int64))
         Predicted: unacc Data: (0, 1, 3, 1, 1, 0) Actual: unacc
                                     , 1.
          (array([[1.
                                               , 1.
                           , 1.
                           , 1.
                                      , 1.
                                                  , 1.41421356]]), array([[ 182, 1 ▼
```