

```
In [1]: import pandas as pd
```

```
In [2]: df = pd.read_csv('D:\\dataset\\kidney_disease.csv')
```

```
In [3]: df.head(5)
```

```
Out[3]:
```

	id	age	bp	sg	al	su	rbc	pc	pcc	ba	...	pcv	wc	rc
0	0	48.0	80.0	1.020	1.0	0.0	NaN	normal	notpresent	notpresent	...	44.0	7800.0	5.2
1	1	7.0	50.0	1.020	4.0	0.0	NaN	normal	notpresent	notpresent	...	38.0	6000.0	NaN
2	2	62.0	80.0	1.010	2.0	3.0	normal	normal	notpresent	notpresent	...	31.0	7500.0	NaN
3	3	48.0	70.0	1.005	4.0	0.0	normal	abnormal	present	notpresent	...	32.0	6700.0	3.9
4	4	51.0	80.0	1.010	2.0	0.0	normal	normal	notpresent	notpresent	...	35.0	7300.0	4.6

5 rows × 26 columns



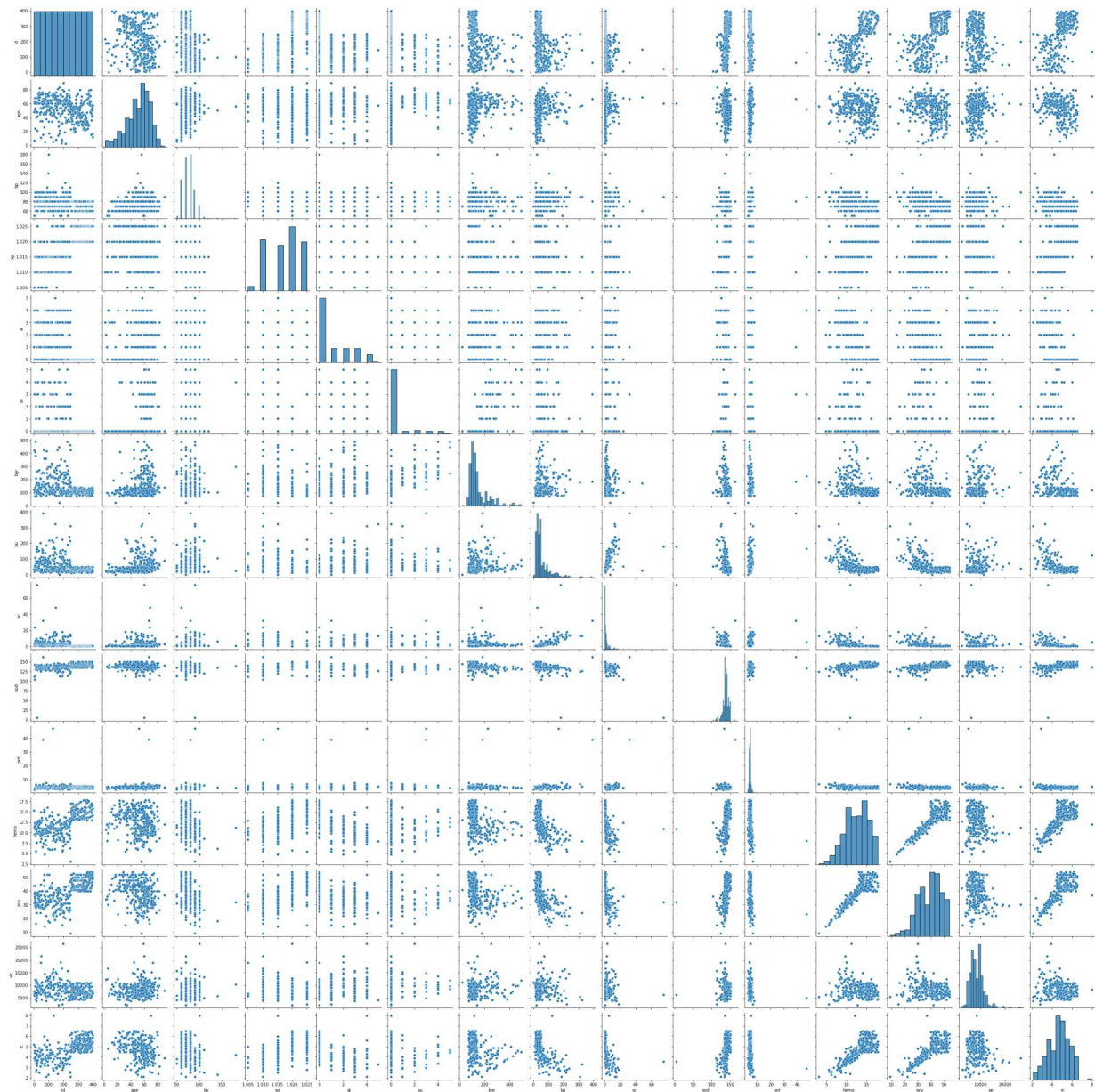
```
In [4]: df.shape
```

```
Out[4]: (400, 26)
```

```
In [5]: import seaborn as sns
```

```
In [6]: sns.pairplot(df)
```

```
Out[6]: <seaborn.axisgrid.PairGrid at 0x80a8748>
```



```
In [7]: df.isnull().any()
```

```
Out[7]: id          False
age          True
bp           True
sg           True
al           True
su           True
rbc          True
pc           True
pcc          True
ba           True
bgr          True
bu           True
sc           True
sod          True
pot          True
hemo         True
pcv          True
wc           True
rc           True
htn          True
dm           True
cad          True
appet        True
pe           True
ane          True
```

```
classification    False
dtype: bool
```

```
In [8]: df['age']=df['age'].fillna(df['age'].median())
df['bp'] = df['bp'].fillna(df['bp'].median())
df['sg'] = df['sg'].fillna(df['sg'].median())
df['al'] = df['al'].fillna(df['al'].median())
df['su'] = df['su'].fillna(df['su'].median())
df['rbc']=df['rbc'].fillna(df['rbc'].mode())
```

```
In [9]: df.isnull().sum()
```

```
Out[9]: id                0
age                0
bp                0
sg                0
al                0
su                0
rbc              151
pc               65
pcc              4
ba               4
bgr             44
bu              19
sc              17
sod             87
pot             88
hemo            52
pcv             71
wc             106
rc             131
htn              2
dm               2
cad              2
appet            1
pe              1
ane              1
classification    0
dtype: int64
```

```
In [12]: df['rbc'].mode()
```

```
Out[12]: 0    normal
dtype: object
```

```
In [13]: df['rbc']=df['rbc'].fillna('normal')
```

```
In [14]: df.isnull().sum()
```

```
Out[14]: id                0
age                0
bp                0
sg                0
al                0
su                0
rbc              0
pc               65
pcc              4
ba               4
bgr             44
bu              19
sc              17
sod             87
pot             88
hemo            52
pcv             71
wc             106
```

```
rc          131
htn         2
dm          2
cad         2
appet       1
pe          1
ane         1
classification 0
dtype: int64
```

```
In [15]: df['pc'].mode()
```

```
Out[15]: 0    normal
dtype: object
```

```
In [16]: df['pc']=df['pc'].fillna('normal')
```

```
In [17]: df['pcc'].mode()
```

```
Out[17]: 0    notpresent
dtype: object
```

```
In [18]: df['pcc']=df['pcc'].fillna('notpresent')
```

```
In [19]: df['ba'].mode()
```

```
Out[19]: 0    notpresent
dtype: object
```

```
In [20]: df['ba']=df['ba'].fillna('notpresent')
```

```
In [21]: df['bgr'].median()
```

```
Out[21]: 121.0
```

```
In [22]: df['bgr']=df['bgr'].fillna(df['bgr'].median())
```

```
In [23]: df['bu']=df['bu'].fillna(df['bu'].median())
```

```
In [24]: df['sc']=df['sc'].fillna(df['sc'].median())
```

```
In [25]: df['sod']=df['sod'].fillna(df['sod'].median())
```

```
In [26]: df['pot']=df['pot'].fillna(df['pot'].median())
```

```
In [27]: df['hemo']=df['hemo'].fillna(df['hemo'].median())
```

```
In [28]: df['pcv']=df['pcv'].fillna(df['pcv'].median())
```

```
In [29]: df['wc']=df['wc'].fillna(df['wc'].median())
```

```
In [30]: df['rc']=df['rc'].fillna(df['rc'].median())
```

```
In [31]: df.isnull().sum()
```

```
Out[31]: id          0
age          0
bp          0
sg          0
al          0
```

```
su          0
rbc         0
pc          0
pcc         0
ba          0
bgr         0
bu          0
sc          0
sod         0
pot         0
hemo        0
pcv         0
wc          0
rc          0
htn         2
dm          2
cad         2
appet       1
pe          1
ane         1
classification 0
dtype: int64
```

```
In [32]: df['htn'].mode()
```

```
Out[32]: 0    no
dtype: object
```

```
In [33]: df['htn']=df['htn'].fillna('no')
```

```
In [34]: df['dm'].mode()
```

```
Out[34]: 0    no
dtype: object
```

```
In [35]: df['dm']=df['dm'].fillna('no')
```

```
In [36]: df['cad'].mode()
```

```
Out[36]: 0    no
dtype: object
```

```
In [37]: df['cad']=df['cad'].fillna('no')
```

```
In [38]: df['appet'].mode()
```

```
Out[38]: 0    good
dtype: object
```

```
In [39]: df['appet']=df['appet'].fillna('good')
```

```
In [40]: df['pe'].mode()
```

```
Out[40]: 0    no
dtype: object
```

```
In [41]: df['pe']=df['pe'].fillna('no')
```

```
In [42]: df['ane'].mode()
```

```
Out[42]: 0    no
dtype: object
```

```
In [43]: df['ane']=df['ane'].fillna('no')
```

```
In [44]: df.isnull().sum()
```

```
Out[44]: id          0
age          0
bp           0
sg           0
al           0
su           0
rbc          0
pc           0
pcc          0
ba           0
bgr          0
bu           0
sc           0
sod          0
pot          0
hemo         0
pcv          0
wc           0
rc           0
htn          0
dm           0
cad          0
appet        0
pe           0
ane          0
classification 0
dtype: int64
```

```
In [45]: from sklearn.preprocessing import LabelEncoder
```

```
In [46]: df.columns
```

```
Out[46]: Index(['id', 'age', 'bp', 'sg', 'al', 'su', 'rbc', 'pc', 'pcc', 'ba', 'bgr',
               'bu', 'sc', 'sod', 'pot', 'hemo', 'pcv', 'wc', 'rc', 'htn', 'dm', 'cad',
               'appet', 'pe', 'ane', 'classification'],
              dtype='object')
```

```
In [47]: df['rbc']=LabelEncoder().fit_transform(df['rbc'])
```

```
In [48]: df['pc']=LabelEncoder().fit_transform(df['pc'])
```

```
In [49]: df['pcc']=LabelEncoder().fit_transform(df['pcc'])
```

```
In [50]: df['ba']=LabelEncoder().fit_transform(df['ba'])
```

```
In [51]: df['htn']=LabelEncoder().fit_transform(df['htn'])
```

```
In [52]: df['dm']=LabelEncoder().fit_transform(df['dm'])
```

```
In [53]: df['cad']=LabelEncoder().fit_transform(df['cad'])
```

```
In [54]: df['appet']=LabelEncoder().fit_transform(df['appet'])
```

```
In [55]: df['pe']=LabelEncoder().fit_transform(df['pe'])
```

```
In [56]: df['ane']=LabelEncoder().fit_transform(df['ane'])
```

```
In [57]: df['classification']=LabelEncoder().fit_transform(df['classification'])
```

In [58]: `df.head()`

```
Out[58]:
```

	id	age	bp	sg	al	su	rbc	pc	pcc	ba	...	pcv	wc	rc	htn	dm	cad	appet	pe
0	0	48.0	80.0	1.020	1.0	0.0	1	1	0	0	...	44.0	7800.0	5.2	1	4	1	0	0
1	1	7.0	50.0	1.020	4.0	0.0	1	1	0	0	...	38.0	6000.0	4.8	0	3	1	0	0
2	2	62.0	80.0	1.010	2.0	3.0	1	1	0	0	...	31.0	7500.0	4.8	0	4	1	1	0
3	3	48.0	70.0	1.005	4.0	0.0	1	0	1	0	...	32.0	6700.0	3.9	1	3	1	1	1
4	4	51.0	80.0	1.010	2.0	0.0	1	1	0	0	...	35.0	7300.0	4.6	0	3	1	0	0

5 rows × 26 columns

In [59]:

```
Q1_df = df.quantile(0.25)
Q3_df = df.quantile(0.75)
IQR = Q3_df - Q1_df
print(IQR)
```

```
id          199.500
age          22.000
bp           10.000
sg            0.005
al            2.000
su            0.000
rbc           0.000
pc            0.000
pcc           0.000
ba            0.000
bgr          49.000
bu           34.750
sc            1.825
sod           6.000
pot           0.800
hemo          3.750
pcv           10.000
wc          2425.000
rc            0.600
htn           1.000
dm            1.000
cad           0.000
appet         0.000
pe            0.000
ane           0.000
classification 2.000
dtype: float64
```

In [60]:

```
df_out= df[~((df<(Q1_df-1.5* IQR))|(df>(Q3_df + 1.5 * IQR))).any(axis=1)]
df_out.shape
```

Out[60]: (148, 26)

In [61]: `df_out.columns`

Out[61]: Index(['id', 'age', 'bp', 'sg', 'al', 'su', 'rbc', 'pc', 'pcc', 'ba', 'bgr', 'bu', 'sc', 'sod', 'pot', 'hemo', 'pcv', 'wc', 'rc', 'htn', 'dm', 'cad', 'appet', 'pe', 'ane', 'classification'], dtype='object')

In [62]:

```
x = df_out[['id', 'age', 'bp', 'sg', 'al', 'su', 'rbc', 'pc', 'pcc', 'ba', 'bgr', 'bu', 'sc', 'sod', 'pot', 'hemo', 'pcv', 'wc', 'rc', 'htn', 'dm', 'cad', 'appet', 'pe', 'ane']]
```

```
In [63]: y = df_out[['classification']]
```

```
In [64]: from sklearn.model_selection import train_test_split
```

```
In [65]: x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
```

```
In [66]: from sklearn import tree
```

```
In [67]: model = tree.DecisionTreeRegressor()
```

```
In [68]: model.fit(x_train,y_train)
```

```
Out[68]: DecisionTreeRegressor()
```

```
In [69]: y_pred = model.predict(x_test)
```

```
In [70]: y_pred
```

```
Out[70]: array([2., 2., 2., 2., 0., 2., 2., 2., 2., 0., 2., 2., 2., 2., 2., 2., 0.,
                2., 2., 2., 2., 2., 2., 2., 2., 2., 2., 2., 0., 2., 2., 2., 2., 2.,
                2., 2., 2., 2., 2., 2., 2., 2., 0., 2., 2.])
```

```
In [71]: import numpy as np
from sklearn import metrics
print('Mean Absolute Error:', metrics.mean_absolute_error(y_test,y_pred))
print('Mean Squared Error:', metrics.mean_squared_error(y_test,y_pred))
print('Root Mean Squared Error:', np.sqrt(metrics.mean_squared_error(y_test,y_pred)))
```

Mean Absolute Error: 0.0

Mean Squared Error: 0.0

Root Mean Squared Error: 0.0

```
In [72]: from sklearn.linear_model import LogisticRegression
```

```
In [73]: model = LogisticRegression()
```

```
In [74]: model.fit(x_train,y_train)
```

D:\data science\lib\site-packages\sklearn\utils\validation.py:73: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_samples,), for example using ravel().

return f(**kwargs)

D:\data science\lib\site-packages\sklearn\linear_model_logistic.py:762: ConvergenceWarning: lbfgs failed to converge (status=1):

STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.

Increase the number of iterations (max_iter) or scale the data as shown in:

<https://scikit-learn.org/stable/modules/preprocessing.html>

Please also refer to the documentation for alternative solver options:

https://scikit-learn.org/stable/modules/linear_model.html#logistic-regression

n_iter_i = _check_optimize_result(

```
Out[74]: LogisticRegression()
```

```
In [86]: y_pred=model.predict(x_test)
```

```
In [87]: y_pred
```

```
Out[87]: array([2, 2, 2, 2, 0, 2, 2, 2, 2, 0, 2, 2, 2, 2, 2, 2, 0, 2, 2, 2, 2, 2,
                2, 2, 2, 2, 2, 0, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 0, 2,
                2])
```



```
In [88]: from sklearn import metrics
```

```
In [89]: cm = metrics.confusion_matrix(y_test,y_pred)
```

```
In [90]: cm
```

```
Out[90]: array([[ 5,  0],  
               [ 0, 40]], dtype=int64)
```

```
In [91]: metrics.accuracy_score(y_test,y_pred)
```

```
Out[91]: 1.0
```

```
In [96]: y_test = y_true
```

```
-----  
NameError                                Traceback (most recent call last)  
<ipython-input-96-687f3cc81c8d> in <module>  
----> 1 y_test = y_true  
  
NameError: name 'y_true' is not defined
```

```
In [ ]:
```