```
# Importing Libraries:
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt

dataset=pd.read_csv(r"/content/kidney_disease.csv")

# Top 5 records:
dataset.head()
```

	id	age	bp	sg	al	su	rbc	рс	рсс	ba	bgr	bu	sc	sod	pot	hemo	pcv	WC	rc	htn
0	0	48.0	80.0	1.020	1.0	0.0	NaN	normal	notpresent	notpresent	121.0	36.0	1.2	NaN	NaN	15.4	44	7800	5.2	yes
1	1	7.0	50.0	1.020	4.0	0.0	NaN	normal	notpresent	notpresent	NaN	18.0	8.0	NaN	NaN	11.3	38	6000	NaN	no
2	2	62.0	80.0	1.010	2.0	3.0	normal	normal	notpresent	notpresent	423.0	53.0	1.8	NaN	NaN	9.6	31	7500	NaN	no
3	3	48.0	70.0	1.005	4.0	0.0	normal	abnormal	present	notpresent	117.0	56.0	3.8	111.0	2.5	11.2	32	6700	3.9	yes
4	4	51.0	80.0	1.010	2.0	0.0	normal	normal	notpresent	notpresent	106.0	26.0	1.4	NaN	NaN	11.6	35	7300	4.6	no

```
# Dropping unneccsary feature :
dataset = dataset.drop('id', axis=1)

# Shape of dataset:
dataset.shape

    (400, 25)
```

```
# Cheaking Missing (NaN) Values:
dataset.isnull().sum()
```

age	9
bp	12
sg	47
al	46
su	49
rbc	152
pc	65
pcc	4
ba	4
bgr	44
bu	19
SC	17
sod	87
pot	88
hemo	52
pcv	70
WC	105
rc	130
htn	2
dm	2
cad	2
appet	1
pe	1
ane	1
classification	0
dtype: int64	

Description:
dataset.describe()

	age	bp	sg	al	su	bgr	bu	sc	sod	pot	
count	391.000000	388.000000	353.000000	354.000000	351.000000	356.000000	381.000000	383.000000	313.000000	312.000000	348.00
mean	51.483376	76.469072	1.017408	1.016949	0.450142	148.036517	57.425722	3.072454	137.528754	4.627244	12.52
std	17.169714	13.683637	0.005717	1.352679	1.099191	79.281714	50.503006	5.741126	10.408752	3.193904	2.9
min	2.000000	50.000000	1.005000	0.000000	0.000000	22.000000	1.500000	0.400000	4.500000	2.500000	3.1(
25%	42.000000	70.000000	1.010000	0.000000	0.000000	99.000000	27.000000	0.900000	135.000000	3.800000	10.30

Datatypes:
dataset.dtypes

age	float64
bp	float64
sg	float64
al	float64
su	float64
rbc	object
рс	object
рсс	object
ba	object
bgr	float64
bu	float64
SC	float64
sod	float64
pot	float64
hemo	float64
pcv	object
WC	object
rc	object
htn	object
dm	object
cad	object
appet	object
pe	object
ane	object
classification	object
dtype: object	

https://colab.research.google.com/drive/17SUYeU8XzpypFXf0AXT7DOmEFcmwb88U#scrollTo=9839ee0e&printMode=true

dataset.head()

```
sg al su
                             rbc
                                                                                          rc htn
age
       bp
                                        рс
                                                  pcc
                                                             ba
                                                                   bgr
                                                                             pcv
                                                                                                    dm
                                                                                                       cad
                                                                                                             appet
                                                                                                                     pe
                                                                                                                        ane
     80.0 1.020 1.0 0.0
48.0
                             NaN
                                    normal
                                            notpresent notpresent 121.0
                                                                              44
                                                                                  7800
                                                                                         5.2
                                                                                              ves
                                                                                                   ves
                                                                                                              good
                                                                                                         no
                                                                                                                     no
                                                                                                                          no
     50.0
          1.020 4.0 0.0
                                            notpresent notpresent
                                                                                  6000
 7.0
                            NaN
                                                                              38
                                                                                        NaN
                                    normal
                                                                  NaN
                                                                                               no
                                                                                                    no
                                                                                                         no
                                                                                                              good
                                                                                                                     no
                                                                                                                          no
     0.08
          1.010 2.0 3.0
                                            notpresent notpresent 423.0
                                                                                  7500
                          normal
                                                                              31
                                                                                        NaN
                                    normal
                                                                                               no
                                                                                                   ves
                                                                                                         no
                                                                                                              poor
                                                                                                                     no
                                                                                                                         ves
48.0
     70.0 1.005 4.0 0.0 normal abnormal
                                               present notpresent 117.0
                                                                              32
                                                                                  6700
                                                                                         3.9
                                                                                              ves
                                                                                                    no
                                                                                                         no
                                                                                                              poor
                                                                                                                    ves
                                                                                                                         ves
51.0 80.0 1.010 2.0 0.0 normal
                                    normal notpresent notpresent 106.0
                                                                              35
                                                                                  7300
                                                                                         4.6
                                                                                               no
                                                                                                    no
                                                                                                         no
                                                                                                              good
                                                                                                                     no
                                                                                                                          no
```

5 rows x 25 columns

```
dataset['rbc'].value counts()
     normal
                 201
                  47
     abnormal
     Name: rbc, dtype: int64
dataset['rbc'] = dataset['rbc'].replace(to replace = {'normal' : 0, 'abnormal' : 1})
dataset['pc'].value counts()
     normal
                 259
     abnormal
                  76
     Name: pc, dtype: int64
dataset['pc'] = dataset['pc'].replace(to replace = {'normal' : 0, 'abnormal' : 1})
dataset['pcc'].value counts()
     notpresent
                   354
```

```
42
     present
     Name: pcc, dtype: int64
dataset['pcc'] = dataset['pcc'].replace(to_replace = {'notpresent':0,'present':1})
dataset['ba'].value counts()
     notpresent
                   374
     present
                    22
     Name: ba, dtype: int64
dataset['ba'] = dataset['ba'].replace(to replace = {'notpresent':0,'present':1})
dataset['htn'].value counts()
     no
            251
            147
     yes
     Name: htn, dtype: int64
dataset['htn'] = dataset['htn'].replace(to replace = {'yes' : 1, 'no' : 0})
dataset['dm'].value counts()
              258
     no
              134
     yes
     \tno
                3
     \tyes
                1
     yes
     Name: dm, dtype: int64
dataset['dm'] = dataset['dm'].replace(to_replace = {'\tyes':'yes', ' yes':'yes', '\tno':'no'})
dataset['dm'] = dataset['dm'].replace(to_replace = {'yes' : 1, 'no' : 0})
```

```
dataset['cad'].value_counts()
             362
     no
              34
     ves
     \tno
               2
     Name: cad, dtype: int64
dataset['cad'] = dataset['cad'].replace(to replace = {'\tno':'no'})
dataset['cad'] = dataset['cad'].replace(to replace = {'yes' : 1, 'no' : 0})
dataset['appet'].unique()
     array(['good', 'poor', nan], dtype=object)
dataset['appet'] = dataset['appet'].replace(to replace={'good':1,'poor':0,'no':np.nan})
dataset['pe'].value_counts()
            323
     no
             76
     yes
     Name: pe, dtype: int64
dataset['pe'] = dataset['pe'].replace(to replace = {'yes' : 1, 'no' : 0})
dataset['ane'].value_counts()
     no
            339
     yes
             60
     Name: ane, dtype: int64
```

dataset.head()

	age	bp	sg	al	su	rbc	рс	рсс	ba	bgr	• • •	pcv	WC	rc	htn	dm	cad	appet	pe	ane	classification
0	48.0	80.0	1.020	1.0	0.0	NaN	0.0	0.0	0.0	121.0		44	7800	5.2	1.0	1.0	0.0	1.0	0.0	0.0	1
1	7.0	50.0	1.020	4.0	0.0	NaN	0.0	0.0	0.0	NaN		38	6000	NaN	0.0	0.0	0.0	1.0	0.0	0.0	1
2	62.0	80.0	1.010	2.0	3.0	0.0	0.0	0.0	0.0	423.0		31	7500	NaN	0.0	1.0	0.0	0.0	0.0	1.0	1
3	48.0	70.0	1.005	4.0	0.0	0.0	1.0	1.0	0.0	117.0		32	6700	3.9	1.0	0.0	0.0	0.0	1.0	1.0	1
4	51.0	80.0	1.010	2.0	0.0	0.0	0.0	0.0	0.0	106.0		35	7300	4.6	0.0	0.0	0.0	1.0	0.0	0.0	1

5 rows × 25 columns

Datatypes: dataset.dtypes

age	float64
bp	float64
sg	float64
al	float64
su	float64

```
float64
     rbc
                       float64
     рс
                       float64
     рсс
                       float64
     ba
                       float64
     bgr
                       float64
     bu
                       float64
     sc
                       float64
     sod
                       float64
     pot
                       float64
     hemo
                        object
     pcv
                        object
     WC
                        object
     rc
     htn
                       float64
                       float64
     dm
                       float64
     cad
     appet
                       float64
                       float64
     pe
                       float64
     ane
     classification
                         int64
     dtype: object
dataset['pcv'] = pd.to numeric(dataset['pcv'], errors='coerce')
dataset['wc'] = pd.to numeric(dataset['wc'], errors='coerce')
dataset['rc'] = pd.to numeric(dataset['rc'], errors='coerce')
# Datatypes:
dataset.dtypes
                       float64
     age
                       float64
     bp
                       float64
     sg
                       float64
     al
                       float64
     su
                       float64
     rbc
                       float64
     рс
                       float64
     рсс
     ba
                       float64
     bgr
                       float64
```

float64 bu float64 sc float64 sod float64 pot hemo float64 float64 pcv float64 WC float64 rc htn float64 float64 dm cad float64 float64 appet float64 pe float64 ane classification int64

dtype: object

Description:
dataset.describe()

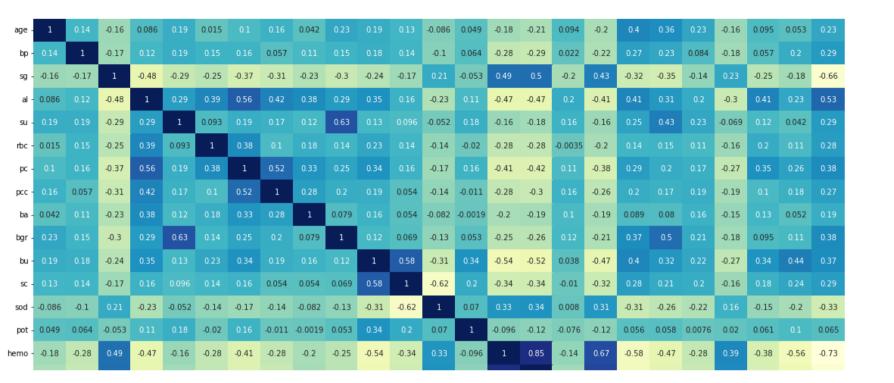
рс

pcc

ba

al bp rbc age sg su # Cheaking Missing (NaN) Values: dataset.isnull().sum().sort_values(ascending=False) rbc 152 rc 131 WC 106 88 pot 87 sod 71 pcv 65 рс 52 hemo 49 su 47 sg al 46 bgr 44 bu 19 17 SC bp 12 9 age ba рсс htn dm 2 cad 1 appet 1 pe 1 ane classification dtype: int64 dataset.columns Index(['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')

bgr ...



dataset.drop('pcv', axis=1, inplace=True)

- 02 03 04 04 04 03 03 03 03 03 03 03 03 03 03 03 03 03

dataset.head()

	age	bp	sg	al	su	rbc	рс	рсс	ba	bgr	• • •	hemo	WC	rc	htn	dm	cad	appet	pe	ane	classification
0	48.0	80.0	1.020	1.0	0.0	0.0	0.0	0.0	0.0	121.0		15.4	7800.0	5.2	1.0	1.0	0.0	1.0	0.0	0.0	1
1	7.0	50.0	1.020	4.0	0.0	0.0	0.0	0.0	0.0	121.0		11.3	6000.0	4.8	0.0	0.0	0.0	1.0	0.0	0.0	1
2	62.0	80.0	1.010	2.0	3.0	0.0	0.0	0.0	0.0	423.0		9.6	7500.0	4.8	0.0	1.0	0.0	0.0	0.0	1.0	1
3	48.0	70.0	1.005	4.0	0.0	0.0	1.0	1.0	0.0	117.0		11.2	6700.0	3.9	1.0	0.0	0.0	0.0	1.0	1.0	1
4	51.0	80.0	1.010	2.0	0.0	0.0	0.0	0.0	0.0	106.0		11.6	7300.0	4.6	0.0	0.0	0.0	1.0	0.0	0.0	1

5 rows × 24 columns

- 0.8

- 0.4

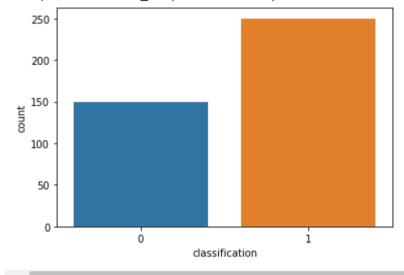
- 0.2

- 0.0

```
# Target feature:
sns.countplot(dataset['classification'])
```

/usr/local/lib/python3.7/dist-packages/seaborn/_decorators.py:43: FutureWarning: Pass the following variable as a keyword arg: FutureWarning

<matplotlib.axes._subplots.AxesSubplot at 0x7f82460d0790>



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