load dataset

```
In [1]:
         import pandas as pd
         import numpy as np
         df = pd.read_csv('kidney_disease.csv')
         df = df.sample(frac=0.2, random_state=42)
         df.info()
        <class 'pandas.core.frame.DataFrame'>
        Index: 80 entries, 209 to 289
       Data columns (total 26 columns):
             Column
                     Non-Null Count Dtype
             -----
                             -----
        ---
        0
             id
                             80 non-null
                                                int64
                             76 non-null
        1
             age
                                               float64
                             79 non-null float64
        2
           bp
                             70 non-null float64
        3 sg
                            70 non-null float64
70 non-null float64
48 non-null object
        4 al
        5
            su
        6 rbc
                            67 non-null object
80 non-null object
80 non-null object
        7
            рс
        8
            рсс
        9
             ba
                            71 non-null
76 non-null
                                              float64
        10 bgr
        11 bu
                                              float64
                            76 non-null float64
67 non-null float64
67 non-null float64
71 non-null float64
67 non-null object
        12 sc
                                              float64
                                              float64
        13 sod
        14 pot
                                               float64
        15 hemo
                                              float64
        16 pcv
        17 wc
                            64 non-null object
                            58 non-null object
80 non-null object
80 non-null object
        18 rc
        19 htn
        20 dm
                            80 non-null
79 non-null
79 non-null
        21 cad
                                                object
        22 appet
                                                object
        23
             pe
                                                object
        24 ane
                              79 non-null
                                                object
        25 classification 80 non-null
                                                object
        dtypes: float64(11), int64(1), object(14)
       memory usage: 16.9+ KB
In [2]: df = pd.read csv('kidney disease.csv')
         df
```

Out[2]:		id	age	bp	sg	al	su	rbc	рс	рсс	l	ba	•••	pcv
	0	0	48.0	80.0	1.020	1.0	0.0	NaN	normal	notpresent	notprese	ent		44
	1	1	7.0	50.0	1.020	4.0	0.0	NaN	normal	notpresent	notprese	ent		38
	2	2	62.0	80.0	1.010	2.0	3.0	normal	normal	notpresent	notprese	ent		31
	3	3	48.0	70.0	1.005	4.0	0.0	normal	abnormal	present	notprese	ent		32
	4	4	51.0	80.0	1.010	2.0	0.0	normal	normal	notpresent	notprese	ent		35
	•••													
	395	395	55.0	80.0	1.020	0.0	0.0	normal	normal	notpresent	notprese	ent		47
	396	396	42.0	70.0	1.025	0.0	0.0	normal	normal	notpresent	notprese	ent		54
	397	397	12.0	80.0	1.020	0.0	0.0	normal	normal	notpresent	notprese	ent		49
	398	398	17.0	60.0	1.025	0.0	0.0	normal	normal	notpresent	notprese	ent		51
	399	399	58.0	80.0	1.025	0.0	0.0	normal	normal	notpresent	notprese	ent		53
	400 rows × 26 columns													
														•
														ŕ
In [3]:	<pre>df = df.drop(['id', 'age'], axis=1)</pre>													
In [4]:	df.s	ample	()											
Out[4]:		bp	sg	al	su		rbc	рс	рсс	ba	bgr	bu	•••	рсл
	346	60.0	NaN	NaN	NaN	nor	mal	normal	notpresent	notpresent	130.0	41.0		52
	1 rows × 24 columns													
	4													•
In [5]:	df.i	nfo()												

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 400 entries, 0 to 399
Data columns (total 24 columns):

```
Column
                Non-Null Count Dtype
   -----
                 -----
---
0
    bp
                 388 non-null
                               float64
                 353 non-null float64
1
    sg
2
                 354 non-null float64
   al
                 351 non-null float64
3
   su
4
   rbc
                 248 non-null object
5
                335 non-null object
   рс
                396 non-null object
6
   рсс
7
                 396 non-null
                               object
    ba
                356 non-null
8
    bgr
                               float64
9
                381 non-null float64
    bu
10 sc
                383 non-null
                               float64
                313 non-null
                               float64
11 sod
12 pot
                312 non-null float64
13 hemo
                348 non-null float64
14 pcv
                330 non-null object
15
                295 non-null
                               object
   WC
16 rc
                270 non-null
                               object
17 htn
                398 non-null
                               object
                398 non-null
18 dm
                               object
                398 non-null object
19 cad
                399 non-null
20 appet
                               object
21 pe
                 399 non-null
                               object
22
                 399 non-null
                               object
23 classification 400 non-null
                               object
dtypes: float64(10), object(14)
memory usage: 75.1+ KB
```

```
In [7]: numerical
Out[7]: ['bp', 'sg', 'al', 'su', 'bgr', 'bu', 'sc', 'sod', 'pot', 'hemo']
In [8]: catgcols
```

```
Out[8]: ['rbc',
           'pc',
           'pcc',
           'ba',
           'pcv',
           'wc',
           'rc',
           'htn',
           'dm',
           'cad',
           'appet',
           'pe',
           'ane',
           'classification']
 In [9]: df['classification'].value_counts()
 Out[9]: classification
          ckd
                    248
          notckd
                   150
          ckd\t
                      2
          Name: count, dtype: int64
In [10]: | df['classification'] = df['classification'].replace(['ckd\t'],'ckd')
In [11]: df['classification'].value_counts()
Out[11]: classification
                  250
          ckd
          notckd
                   150
          Name: count, dtype: int64
In [12]: ind_col = [col for col in df.columns if col != 'classification']
         dep_col = 'classification'
In [13]: df[dep col].value counts()
Out[13]: classification
                    250
          ckd
          notckd
                    150
          Name: count, dtype: int64
```

transformasi dtaset

```
In [14]: from sklearn.preprocessing import LabelEncoder
    le = LabelEncoder()
    for col in catgcols:
        df[col] = le.fit_transform(df[col])

In [15]: df['classification'] = le.fit_transform(df['classification'])

In [16]: x = df[ind_col] #feature
    y = df[dep_col] #Label

In [17]: df.head()
```

sg

al

bp

Out[17]:

In []:

```
80.0 1.020
                              0.0
                                                     121.0 36.0
                                                                          72
                                                                              34
                                                                                               1
                         1.0
                                             0
                                                  0
                                                                      32
                                                                                     1
                                                                                          4
             50.0 1.020 4.0
                              0.0
                                                                              34
                                                                                          3
                                             0
                                                    121.0 18.0
                                                                      26
                                                                          56
                                                                                     0
                                                                                               1
             80.0 1.010 2.0
                              3.0
                                             0
                                                  0
                                                    423.0
                                                          53.0
                                                                          70
                                                                              34
                                                                                     0
                                                                                          4
                                                                                               1
                                                                      19
             70.0 1.005 4.0
                                                                              19
                                                                                               1
                              0.0
                                                  0
                                                    117.0
                                                           56.0
                                                                      20
                                                                          62
                                                                                          3
             80.0 1.010 2.0 0.0
                                                    106.0 26.0
                                                                                     0
                                                                                          3
                                                                                               1
                                     1
                                         1
                                             0
                                                  0
                                                                      23
                                                                          68
                                                                              27
         5 rows × 24 columns
          df.to_csv('kidney_clean.csv', index=False)
In [18]:
In [19]: from sklearn.model_selection import train_test_split
          x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.35, random
```

bgr

bu

rc htn

pcv wc

dm

cad

su rbc pc pcc ba

Membuat Model Decision Tree

```
In [20]: from sklearn.metrics import accuracy_score, confusion_matrix, classification_rep
         from sklearn.tree import DecisionTreeClassifier
         from sklearn import tree
         dtc = DecisionTreeClassifier(
             ccp_alpha=0.0, class_weight=None, criterion='entropy',
             max_depth=4, max_features=None, max_leaf_nodes=None,
             min_impurity_decrease=0.0, min_samples_leaf=1,
             min_samples_split=2, min_weight_fraction_leaf=0.0,
         random state=42, splitter='best'
         model = dtc.fit(x_train, y_train)
         dtc_acc = accuracy_score(y_test, dtc.predict(x_test))
         print(f"Akurasi Data Training = {accuracy_score(y_train, dtc.predict(x_train))}"
         print(f"Akurasi Data Testing = {dtc_acc} \n")
         print(f"Confusion Matrix : \n{confusion_matrix(y_test, dtc.predict(x_test))}\n")
         confusion = confusion_matrix(y_test, dtc.predict(x_test))
         tn, fp, fn, tp = confusion.ravel()
         print(f"Nilai Classfication Report : \n {classification report(y test, dtc.predi
```

```
Akurasi Data Training = 0.9923076923076923
Akurasi Data Testing = 0.9928571428571429
Confusion Matrix :
[[85 1]
[ 0 54]]
Nilai Classfication Report :
                         recall f1-score support
             precision
                         0.99
          0
                 1.00
                                    0.99
                                               86
                 0.98
                         1.00
                                    0.99
                                               54
                                    0.99
                                              140
   accuracy
                 0.99
                                    0.99
                                              140
  macro avg
                        0.99
weighted avg
                 0.99
                         0.99
                                    0.99
                                              140
```

SIMULASI MODEL

```
In [21]: input_data = (80,1.02,1,0,1,1,0,0,121,36,1.2,138,4.4,15.4,32,72,34,1,4,1,0,0,0)
    input_data_as_numpy_array = np.array(input_data)
    input_data_reshape = input_data_as_numpy_array.reshape(1,-1)
    prediction = model.predict(input_data_reshape)
    print(prediction)

if (prediction[0]==0):
    print ('Pasien Tidak Terkena Batu Ginjal')

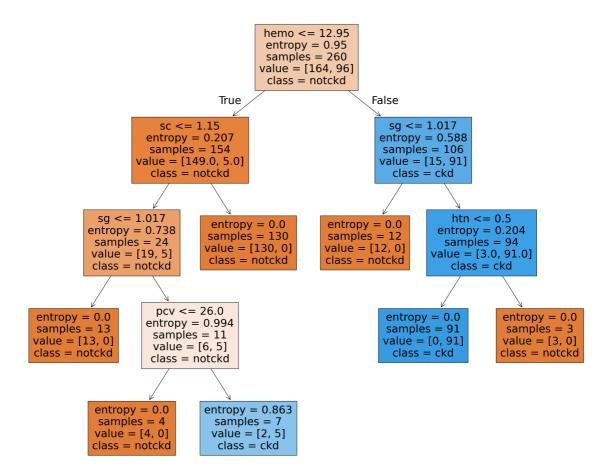
else:
    print ('Pasien Terkena Batu Ginjal')

[0]
    Pasien Tidak Terkena Batu Ginjal

c:\Users\acer\AppData\Local\Programs\Python\Python310\lib\site-packages\sklearn\utils\validation.py:2739: UserWarning: X does not have valid feature names, but De cisionTreeClassifier was fitted with feature names
```

visualisasi Decision Tree

warnings.warn(



Export to pkl

```
In [23]: import joblib as jb
In [24]: jb.dump(model, 'decision_tree_model.pkl')
Out[24]: ['decision_tree_model.pkl']
```