

```
In [122]: import pandas as pd
import matplotlib.pyplot as plt
import numpy as np
import warnings
import seaborn as sns
warnings.filterwarnings('ignore')
plt.style.use('fivethirtyeight')
%matplotlib inline
pd.set_option('display.max_columns', 26)
```

```
In [9]: df = pd.read_csv('kidney_disease.csv')
df.head()
```

Out[9]:

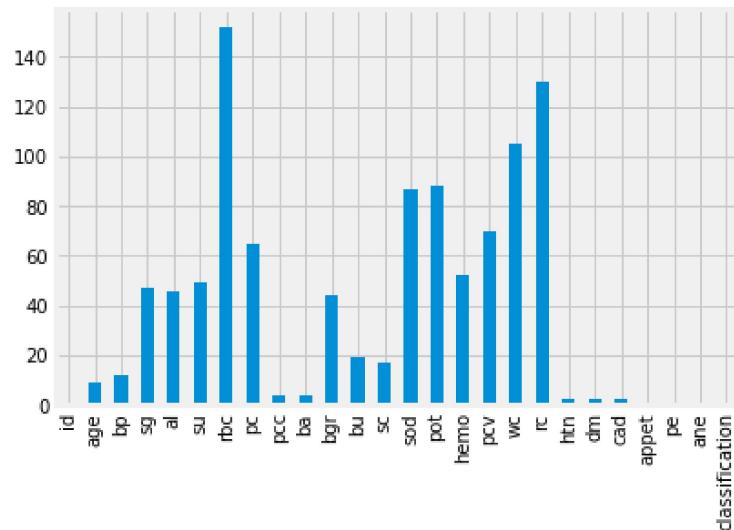
bp	sg	al	su	rbc	pc	pcc	ba	bgr	bu	sc	sod	pot	hemo	pcv	wc	rc
30.0	1.020	1.0	0.0	NaN	normal	notpresent	notpresent	121.0	36.0	1.2	NaN	NaN	15.4	44	7800	5.2
50.0	1.020	4.0	0.0	NaN	normal	notpresent	notpresent	NaN	18.0	0.8	NaN	NaN	11.3	38	6000	NaN
30.0	1.010	2.0	3.0	normal	normal	notpresent	notpresent	423.0	53.0	1.8	NaN	NaN	9.6	31	7500	NaN
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
30.0	1.010	2.0	0.0	normal	normal	notpresent	notpresent	106.0	26.0	1.4	NaN	NaN	11.6	35	7300	4.6



```
In [10]: # dropping id column
df.drop('id', axis = 1, inplace = True)
```

```
In [13]: #rename
df.columns = ['age', 'blood_pressure', 'specific_gravity', 'albumin', 'sugar', 'red_blood_
'pus_cell_clumps', 'bacteria', 'blood_glucose_random', 'blood_urea', 'serum_
'potassium', 'haemoglobin', 'packed_cell_volume', 'white_blood_cell_count',
'hypertension', 'diabetes_mellitus', 'coronary_artery_disease', 'appetite',
'aanemia', 'class']
```

In [120]: # Here we are plotting the graph to see the null values in the dataset.
`p = df.isnull().sum().plot.bar()`



In [14]: `df.head()`

Out[14]:

	age	blood_pressure	specific_gravity	albumin	sugar	red_blood_cells	pus_cell	pus_cell_clumps	bacteria
0	48.0	80.0	1.020	1.0	0.0		NaN	normal	notpresent
1	7.0	50.0	1.020	4.0	0.0		NaN	normal	notpresent
2	62.0	80.0	1.010	2.0	3.0		normal	normal	notpresent
3	48.0	70.0	1.005	4.0	0.0		normal	abnormal	present
4	51.0	80.0	1.010	2.0	0.0		normal	normal	notpresent

In [15]: `df.describe()`

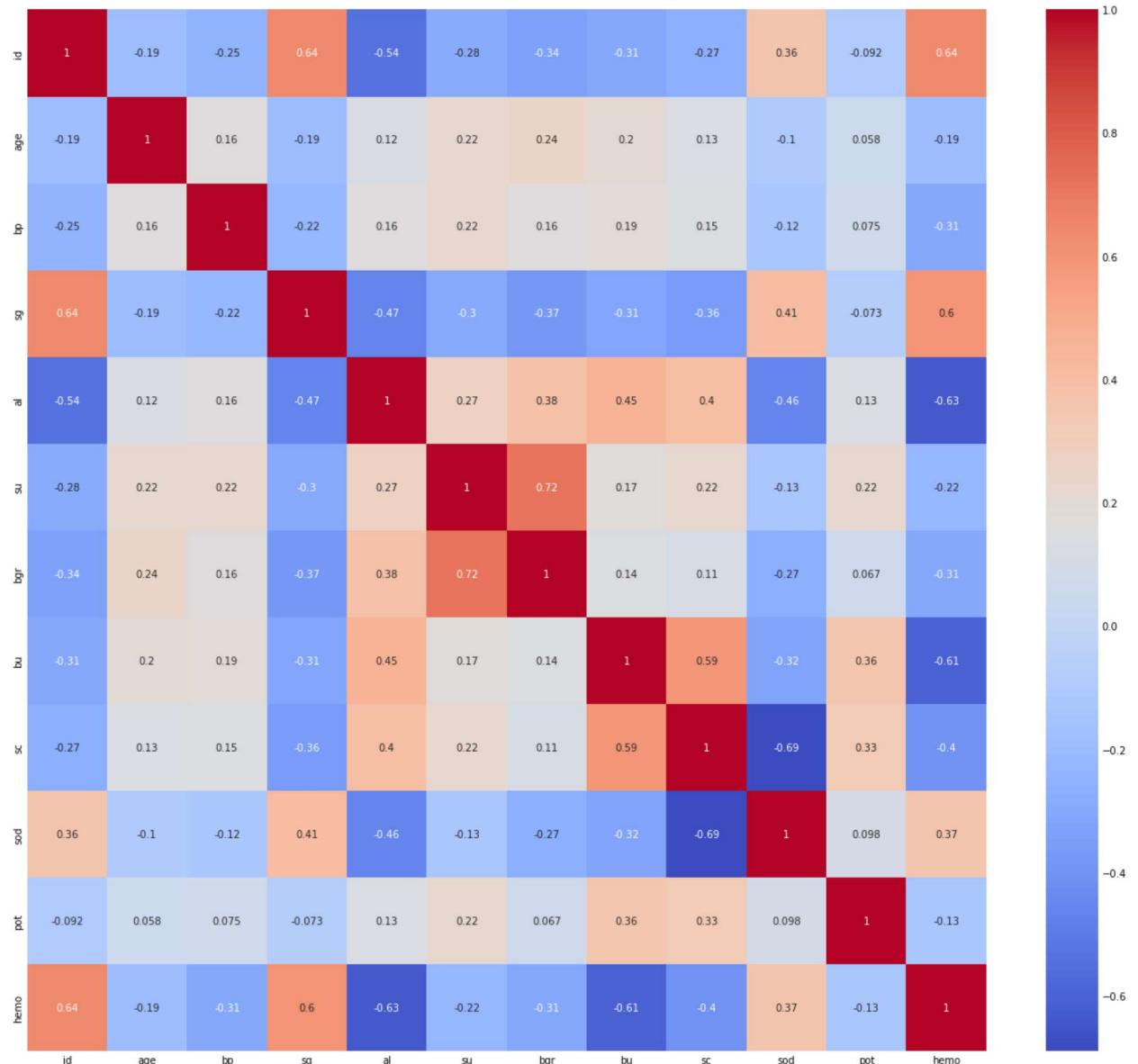
Out[15]:

age	blood_pressure	specific_gravity	albumin	sugar	blood_glucose_random	blood_urea	serum_c
391.000000	388.000000	353.000000	354.000000	351.000000	356.000000	381.000000	381.000000
51.483376	76.469072	1.017408	1.016949	0.450142	148.036517	57.425722	
17.169714	13.683637	0.005717	1.352679	1.099191	79.281714	50.503006	
2.000000	50.000000	1.005000	0.000000	0.000000	22.000000	1.500000	
42.000000	70.000000	1.010000	0.000000	0.000000	99.000000	27.000000	
55.000000	80.000000	1.020000	0.000000	0.000000	121.000000	42.000000	
64.500000	80.000000	1.020000	2.000000	0.000000	163.000000	66.000000	
90.000000	180.000000	1.025000	5.000000	5.000000	490.000000	391.000000	7

In [123]: #Finding the Correlation between the plots

```
plt.figure(figsize = (19,19))
sns.heatmap(data.corr(), annot = True, cmap = 'coolwarm') # looking for strong correlation.
```

Out[123]: <AxesSubplot:>



```
In [4]: # checking for null values
```

```
df.isna().sum().sort_values(ascending = False)
```

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

```
In [17]: df['packed_cell_volume'] = pd.to_numeric(df['packed_cell_volume'], errors='coerce')
df['white_blood_cell_count'] = pd.to_numeric(df['white_blood_cell_count'], errors='coerce')
df['red_blood_cell_count'] = pd.to_numeric(df['red_blood_cell_count'], errors='coerce')
```

```
In [18]: # Extracting categorical and numerical columns
```

```
cat_cols = [col for col in df.columns if df[col].dtype == 'object']
num_cols = [col for col in df.columns if df[col].dtype != 'object']
```

```
In [19]: # Looking at unique values in categorical columns
```

```
for col in cat_cols:
    print(f"{col} has {df[col].unique()} values\n")
```

```
red_blood_cells has [nan 'normal' 'abnormal'] values
```

```
pus_cell has ['normal' 'abnormal' nan] values
```

```
pus_cell_clumps has ['notpresent' 'present' nan] values
```

```
bacteria has ['notpresent' 'present' nan] values
```

```
hypertension has ['yes' 'no' nan] values
```

```
diabetes_mellitus has ['yes' 'no' 'yes' '\tno' '\tyes' nan] values
```

```
coronary_artery_disease has ['no' 'yes' '\tno' nan] values
```

```
appetite has ['good' 'poor' nan] values
```

```
peda_edema has ['no' 'yes' nan] values
```

```
aanemia has ['no' 'yes' nan] values
```

```
class has ['ckd' 'ckd\t' 'notckd'] values
```

```
In [21]: df['diabetes_mellitus'].replace(to_replace = {'\tno':'no', '\tyes':'yes', ' yes':'yes'}, inplace=True)
```

```
df['coronary_artery_disease'] = df['coronary_artery_disease'].replace(to_replace = '\tno', value='no')
```

```
df['class'] = df['class'].replace(to_replace = {'ckd\t': 'ckd', 'notckd': 'not ckd'})
```

```
In [22]: df['class'] = df['class'].map({'ckd': 0, 'not ckd': 1})
```

```
df['class'] = pd.to_numeric(df['class'], errors='coerce')
```

```
In [23]: cols = ['diabetes_mellitus', 'coronary_artery_disease', 'class']
```

```
for col in cols:
```

```
    print(f"{col} has {df[col].unique()} values\n")
```

```
diabetes_mellitus has ['yes' 'no' nan] values
```

```
coronary_artery_disease has ['no' 'yes' nan] values
```

```
class has [0 1] values
```

```
In [24]: df.isna().sum().sort_values(ascending = False)
```

```
Out[24]: red_blood_cells      152
red_blood_cell_count       131
white_blood_cell_count     106
potassium                  88
sodium                     87
packed_cell_volume         71
pus_cell                   65
haemoglobin                52
sugar                      49
specific_gravity           47
albumin                    46
blood_glucose_random       44
blood_urea                  19
serum_creatinine            17
blood_pressure               12
age                         9
bacteria                   4
pus_cell_clumps             4
hypertension                 2
diabetes_mellitus            2
coronary_artery_disease      2
appetite                    1
peda_edema                  1
aanemia                     1
class                       0
dtype: int64
```

```
In [25]: df[num_cols].isnull().sum()
```

```
Out[25]: age                  9
blood_pressure              12
specific_gravity            47
albumin                    46
sugar                      49
blood_glucose_random        44
blood_urea                  19
serum_creatinine             17
sodium                     87
potassium                  88
haemoglobin                52
packed_cell_volume          71
white_blood_cell_count      106
red_blood_cell_count        131
dtype: int64
```

```
In [26]: df[cat_cols].isnull().sum()
```

```
Out[26]: red_blood_cells      152
          pus_cell             65
          pus_cell_clumps       4
          bacteria              4
          hypertension           2
          diabetes_mellitus      2
          coronary_artery_disease 2
          appetite               1
          peda_edema             1
          aanemia                1
          class                  0
          dtype: int64
```

```
In [28]: # filling null values, we will use two methods, random sampling for higher null values and
```

```
# mean/mode sampling for lower null values
```

```
def random_value_imputation(feature):
    random_sample = df[feature].dropna().sample(df[feature].isna().sum())
    random_sample.index = df[df[feature].isnull()].index
    df.loc[df[feature].isnull(), feature] = random_sample

def impute_mode(feature):
    mode = df[feature].mode()[0]
    df[feature] = df[feature].fillna(mode)
```

```
In [29]: # filling num_cols null values using random sampling method
```

```
for col in num_cols:
    random_value_imputation(col)
```

```
In [30]: df[num_cols].isnull().sum()
```

```
Out[30]: age                  0
          blood_pressure        0
          specific_gravity       0
          albumin                0
          sugar                  0
          blood_glucose_random   0
          blood_urea              0
          serum_creatinine        0
          sodium                 0
          potassium              0
          haemoglobin            0
          packed_cell_volume      0
          white_blood_cell_count  0
          red_blood_cell_count    0
          dtype: int64
```

```
In [31]: # filling "red_blood_cells" and "pus_cell" using random sampling method and rest of cat_co
```

```
random_value_imputation('red_blood_cells')
random_value_imputation('pus_cell')

for col in cat_cols:
    impute_mode(col)
```

```
In [32]: df[cat_cols].isnull().sum()
```

```
Out[32]: red_blood_cells      0
pus_cell          0
pus_cell_clumps  0
bacteria         0
hypertension      0
diabetes_mellitus 0
coronary_artery_disease 0
appetite          0
peda_edema        0
aanemia           0
class             0
dtype: int64
```

```
In [33]: for col in cat_cols:
    print(f"{col} has {df[col].nunique()} categories\n")
```

```
red_blood_cells has 2 categories
pus_cell has 2 categories
pus_cell_clumps has 2 categories
bacteria has 2 categories
hypertension has 2 categories
diabetes_mellitus has 2 categories
coronary_artery_disease has 2 categories
appetite has 2 categories
peda_edema has 2 categories
aanemia has 2 categories
class has 2 categories
```

```
In [34]: from sklearn.preprocessing import LabelEncoder
```

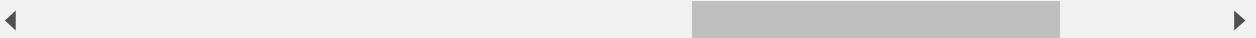
```
le = LabelEncoder()

for col in cat_cols:
    df[col] = le.fit_transform(df[col])
```

In [35]: `df.head()`

Out[35]:

acked_cell_volume	white_blood_cell_count	red_blood_cell_count	hypertension	diabetes_mellitus	coronary_artery_
44.0	7800.0	5.2	1	1	
38.0	6000.0	5.1	0	0	
31.0	7500.0	3.9	0	1	
32.0	6700.0	3.9	1	0	
35.0	7300.0	4.6	0	0	



In []:

In []:

In []:

In [38]:

```
ind_col = [col for col in df.columns if col != 'class']
dep_col = 'class'
X = df[ind_col]
y = df[dep_col]
```

In [39]:

```
# splitting data intp training and test set
```

```
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.30, random_state =
```

K-NEAREST NEIGHBOUR

In [40]:

```
#knn
from sklearn.neighbors import KNeighborsClassifier
from sklearn.metrics import accuracy_score, confusion_matrix, classification_report

knn = KNeighborsClassifier()
knn.fit(X_train, y_train)

# accuracy score, confusion matrix and classification report of knn

knn_acc = accuracy_score(y_test, knn.predict(X_test))

print(f"Training Accuracy of KNN is {accuracy_score(y_train, knn.predict(X_train))}")
print(f"Test Accuracy of KNN is {knn_acc} \n")

print(f"Confusion Matrix :- \n{confusion_matrix(y_test, knn.predict(X_test))}\n")
print(f"Classification Report :- \n {classification_report(y_test, knn.predict(X_test))}\n")
```

Training Accuracy of KNN is 0.7964285714285714

Test Accuracy of KNN is 0.6666666666666666

Confusion Matrix :-

```
[[52 20]
 [20 28]]
```

Classification Report :-

	precision	recall	f1-score	support
0	0.72	0.72	0.72	72
1	0.58	0.58	0.58	48
accuracy			0.67	120
macro avg	0.65	0.65	0.65	120
weighted avg	0.67	0.67	0.67	120

DECISION TREE CLASSIFIER

```
In [41]: from sklearn.tree import DecisionTreeClassifier

dtc = DecisionTreeClassifier()
dtc.fit(X_train, y_train)

# accuracy score, confusion matrix and classification report of decision tree

dtc_acc = accuracy_score(y_test, dtc.predict(X_test))

print(f"Training Accuracy of Decision Tree Classifier is {accuracy_score(y_train, dtc.predict(X_train))}")
print(f"Test Accuracy of Decision Tree Classifier is {dtc_acc} \n")

print(f"Confusion Matrix :- \n{confusion_matrix(y_test, dtc.predict(X_test))}\n")
print(f"Classification Report :- \n {classification_report(y_test, dtc.predict(X_test))}\n")
```

Training Accuracy of Decision Tree Classifier is 1.0
Test Accuracy of Decision Tree Classifier is 0.9416666666666667

Confusion Matrix :-

```
[[71  1]
 [ 6 42]]
```

Classification Report :-

	precision	recall	f1-score	support
0	0.92	0.99	0.95	72
1	0.98	0.88	0.92	48
accuracy			0.94	120
macro avg	0.95	0.93	0.94	120
weighted avg	0.94	0.94	0.94	120

RANDOM FOREST CLASSIFIER

```
In [42]: from sklearn.ensemble import RandomForestClassifier  
  
rd_clf = RandomForestClassifier(criterion = 'entropy', max_depth = 11, max_features = 'auto')  
rd_clf.fit(X_train, y_train)  
  
# accuracy score, confusion matrix and classification report of random forest  
  
rd_clf_acc = accuracy_score(y_test, rd_clf.predict(X_test))  
  
print(f"Training Accuracy of Random Forest Classifier is {accuracy_score(y_train, rd_clf.predict(X_train))}")  
print(f"Test Accuracy of Random Forest Classifier is {rd_clf_acc} \n")  
  
print(f"Confusion Matrix :- \n{confusion_matrix(y_test, rd_clf.predict(X_test))}\n")  
print(f"Classification Report :- \n {classification_report(y_test, rd_clf.predict(X_test))}
```

Training Accuracy of Random Forest Classifier is 1.0
Test Accuracy of Random Forest Classifier is 0.9666666666666667

Confusion Matrix :-

```
[[72  0]  
 [ 4 44]]
```

Classification Report :-

	precision	recall	f1-score	support
0	0.95	1.00	0.97	72
1	1.00	0.92	0.96	48
accuracy			0.97	120
macro avg	0.97	0.96	0.96	120
weighted avg	0.97	0.97	0.97	120

In []:

In [118]:

```
In [119]: x = df[ind_col]
y = df[dep_col]
xtrain, xtest, ytrain, ytest = train_test_split(x, y, test_size=0.15)
```

```
-----
KeyError Traceback (most recent call last)
~\AppData\Local\Temp\ipykernel_29180\3116842793.py in <module>
----> 1 x = df[ind_col]
      2 y = df[dep_col]
      3 xtrain, xtest, ytrain, ytest = train_test_split(x, y, test_size=0.15)

~\OneDrive\Documents\lib\site-packages\pandas\core\frame.py in __getitem__(self, key)
 3462         if is_iterator(key):
 3463             key = list(key)
-> 3464         indexer = self.loc._get_listlike_indexer(key, axis=1)[1]
 3465
 3466     # take() does not accept boolean indexers

~\OneDrive\Documents\lib\site-packages\pandas\core\indexing.py in _get_listlike_indexer(self, key, axis)
 1312         keyarr, indexer, new_indexer = ax._reindex_non_unique(keyarr)
 1313
-> 1314     self._validate_read_indexer(keyarr, indexer, axis)
 1315
 1316     if needs_i8_conversion(ax.dtype) or isinstance(
~\OneDrive\Documents\lib\site-packages\pandas\core\indexing.py in _validate_read_indexer(self, key, indexer, axis)
 1375
 1376     not_found = list(ensure_index(key)[missing_mask.nonzero()[0]].unique
())
-> 1377     raise KeyError(f"{not_found} not in index")
 1378
 1379

KeyError: "['blood_pressure', 'specific_gravity', 'albumin', 'sugar', 'red_blood_cells', 'pus_cell', 'pus_cell_clumps', 'bacteria', 'blood_glucose_random', 'blood_urea', 'serum_creatinine', 'sodium', 'potassium', 'haemoglobin', 'packed_cell_volume', 'white_blood_cell_count', 'red_blood_cell_count', 'hypertension', 'diabetes_mellitus', 'coronary_artery_disease', 'appetite', 'peda_edema', 'aanemia'] not in index"
```

```
In [85]: svc = SVC()
print(svc)

SVC(C=1.0, break_ties=False, cache_size=200, class_weight=None, coef0=0.0,
    decision_function_shape='ovr', degree=3, gamma='scale', kernel='rbf',
    max_iter=-1, probability=False, random_state=None, shrinking=True,
    tol=0.001, verbose=False)
```

SVC()

Out[85]: SVC()

```
In [87]: svc.fit(xtrain, ytrain)
score = svc.score(xtrain, ytrain)
print("Score: ", score)
```

Score: 0.6176470588235294

In [77]:

```
ypred = svc.predict(xtest)

cm = confusion_matrix(ytest, ypred)
print(cm)
```

```
[[40  0]
 [20  0]]
```

LINEAR REGRESSION

In [111]:

```
from sklearn.linear_model import LinearRegression
from sklearn.metrics import r2_score
from sklearn.model_selection import train_test_split
import pandas as pd

# Load data into a Pandas DataFrame
data = pd.read_csv('data.csv')

# Separate the input (X) and output (y) variables
X = df.iloc[:, :-1].values
y = df.iloc[:, -1].values

# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)

# Create a Linear Regression object and fit it to the training data
regressor = LinearRegression()
regressor.fit(X_train, y_train)

# Predict the output values for the testing set
y_pred = regressor.predict(X_test)

# Calculate the R-squared score
accuracy = r2_score(y_test, y_pred)

# Print the accuracy score
print("Accuracy (R-squared):", accuracy)
```

Accuracy (R-squared): 0.7045578655846056

```
from sklearn.linear_model import LinearRegression from sklearn.metrics import r2_score from sklearn.model_selection import train_test_split import pandas as pd data = pd.read_csv('kidney_disease.csv') X = df.iloc[:, :-1].values y = df.iloc[:, -1].values X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0) regressor = LinearRegression() regressor.fit(X_train, y_train) y_pred = regressor.predict(X_test) accuracy = r2_score(y_test, y_pred) print("Accuracy (R-squared):", accuracy)
```

POLYNOMIAL REGRESSION

```
In [113]: from sklearn.preprocessing import PolynomialFeatures
from sklearn.linear_model import LinearRegression
from sklearn.metrics import r2_score
import numpy as np

# Transform the input data to include polynomial features up to degree 2
poly = PolynomialFeatures(degree=2)
X_poly = poly.fit_transform(X)

# Create a Linear Regression object and fit it to the training data
regressor = LinearRegression()
regressor.fit(X_train, y_train)

# Predict the output values for the testing set
y_pred = regressor.pre
dict(X_test)

# Calculate the R-squared score
accuracy = r2_score(y_test, y_pred)

# Print the accuracy score
print("Accuracy (R-squared):", accuracy)
```

```
-----
AttributeError                                     Traceback (most recent call last)
~\AppData\Local\Temp\ipykernel_29180/3116478538.py in <module>
      16
      17 # Predict the output values for the testing set
---> 18 y_pred = regressor.pre
      19
      20 dict(X_test)

AttributeError: 'LinearRegression' object has no attribute 'pre'
```

LOGISTIC REGRESSION

```
In [112]: from sklearn.linear_model import LogisticRegression
from sklearn.metrics import accuracy_score
from sklearn.model_selection import train_test_split
import pandas as pd

# Load data into a Pandas DataFrame
data = pd.read_csv('kidney_disease.csv')

# Separate the input (X) and output (y) variables
X = df.iloc[:, :-1].values
y = df.iloc[:, -1].values

# Split the data into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)

# Create a Logistic Regression object and fit it to the training data
c
classifier = LogisticRegression()
classifier.fit(X_train, y_train)

# Predict the output values for the testing set
y_pred = classifier.predict(X_test)

# Calculate the accuracy score
accuracy = accuracy_score(y_test, y_pred)

# Print the accuracy score
print("Accuracy:", accuracy)
```

Accuracy: 0.925

RANDOM FOREST

```
In [125]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score, confusion_matrix, classification_report
```

```
In [127]: X = df.iloc[:, :-1].values
y = df.iloc[:, -1].values
```

```
In [128]: x_train, x_test, y_train, y_test = train_test_split(x,y,test_size = 0.2,random_state=0)
```

```
In [129]: # Instantiate and fit the RandomForestClassifier
forest = RandomForestClassifier()
forest.fit(x_train, y_train)
```

```
Out[129]: RandomForestClassifier()
```

```
In [130]: # Make predictions for the test set  
y_pred_test = forest.predict(x_test)
```

```
In [132]: # View accuracy score  
accuracy_score(y_test, y_pred_test)
```

Out[132]: 0.9625

```
In [133]: # View confusion matrix for test data and predictions  
confusion_matrix(y_test, y_pred_test)
```

```
Out[133]: array([[51,  0,  0],  
                  [ 1,  0,  0],  
                  [ 2,  0, 26]], dtype=int64)
```

```
In [135]: # View the classification report for test data and predictions  
print(classification_report(y_test, y_pred_test))
```

	precision	recall	f1-score	support
ckd	0.94	1.00	0.97	51
ckd	0.00	0.00	0.00	1
notckd	1.00	0.93	0.96	28
accuracy			0.96	80
macro avg	0.65	0.64	0.64	80
weighted avg	0.95	0.96	0.96	80

DECISION TREE

```
In [150]: import sys
from sklearn.tree import DecisionTreeClassifier
from sklearn.metrics import accuracy_score

import numpy as np
import pylab as pl
X = df.iloc[:, :-1].values
y = df.iloc[:, -1].values

clf = DecisionTreeClassifier()
clf = clf.fit(X,y)
labels_test = clf.predict(X_test)

acc = accuracy_score(y_test, y_train)
```

```
-----  
ValueError Traceback (most recent call last)  
~\AppData\Local\Temp\ipykernel_29180\3347318169.py in <module>  
      9  
     10 clf = DecisionTreeClassifier()  
--> 11 clf = clf.fit(X,y)  
     12 labels_test = clf.predict(X_test)  
     13  
  
~\OneDrive\Documents\lib\site-packages\sklearn\tree\_classes.py in fit(self, X, y, sample_weight, check_input, X_idx_sorted)  
  901         """  
  902  
--> 903         super().fit(  
  904             X, y,  
  905             sample_weight=sample_weight,  
  
~\OneDrive\Documents\lib\site-packages\sklearn\tree\_classes.py in fit(self, X, y, sample_weight, check_input, X_idx_sorted)  
155         check_X_params = dict(dtype=DTYPE, accept_sparse="csc")  
156         check_y_params = dict(ensure_2d=False, dtype=None)  
--> 157         X, y = self._validate_data(X, y,  
158                         validate_separately=(check_X_params,  
159                         check_y_params))  
  
~\OneDrive\Documents\lib\site-packages\sklearn\base.py in _validate_data(self, X, y, reset, validate_separately, **check_params)  
428         # :(  
429         check_X_params, check_y_params = validate_separately  
--> 430         X = check_array(X, **check_X_params)  
431         y = check_array(y, **check_y_params)  
432     else:  
  
~\OneDrive\Documents\lib\site-packages\sklearn\utils\validation.py in inner_f(*args, **kwargs)  
61         extra_args = len(args) - len(all_args)  
62         if extra_args <= 0:  
--> 63             return f(*args, **kwargs)  
64  
65         # extra_args > 0  
  
~\OneDrive\Documents\lib\site-packages\sklearn\utils\validation.py in check_array(array, accept_sparse, accept_large_sparse, dtype, order, copy, force_all_finite, ensure_2d, allow_nd, ensure_min_samples, ensure_min_features, estimator)  
671         array = array.astype(dtype, casting="unsafe", copy=False)  
672     else:  
--> 673         array = np.asarray(array, order=order, dtype=dtype)  
674     except ComplexWarning as complex_warning:  
675         raise ValueError("Complex data not supported\n")  
  
~\OneDrive\Documents\lib\site-packages\numpy\core\_asarray.py in asarray(a, dtype, order, like)  
100         return _asarray_with_like(a, dtype=dtype, order=order, like=like)  
101  
--> 102     return array(a, dtype, copy=False, order=order)  
103  
104  
ValueError: could not convert string to float: 'normal'
```

POLYNOMIAL REGRESSION

```
In [154]: from sklearn.preprocessing import PolynomialFeatures
import numpy as np
X = df.iloc[:, :-1].values
y = df.iloc[:, -1].values
x_train, x_test, y_train, y_test = train_test_split(x,y,test_size = 0.2,random_state=0)
# Transform the input data to include polynomial features up to degree 2
poly = PolynomialFeatures(degree=2)
X_poly = poly.fit_transform(X)
lin_reg_2 = LinearRegression()
lin_reg_2.fit(X_poly, y)
```

```

-----
ValueError                                Traceback (most recent call last)
~\AppData\Local\Temp\ipykernel_29180\62937563.py in <module>
      6 # Transform the input data to include polynomial features up to degree 2
      7 poly = PolynomialFeatures(degree=2)
-->  8 X_poly = poly.fit_transform(X)
      9 lin_reg_2 = LinearRegression()
     10 lin_reg_2.fit(X_poly, y)

~\OneDrive\Documents\lib\site-packages\sklearn\base.py in fit_transform(self, X, **fit_params)
   697         if y is None:
   698             # fit method of arity 1 (unsupervised transformation)
--> 699             return self.fit(X, **fit_params).transform(X)
   700         else:
   701             # fit method of arity 2 (supervised transformation)

~\OneDrive\Documents\lib\site-packages\sklearn\preprocessing\_data.py in fit(self, X, y)
  1702         Fitted transformer.
  1703         """
--> 1704     n_samples, n_features = self._validate_data(
  1705         X, accept_sparse=True).shape
  1706     combinations = self._combinations(n_features, self.degree,

~\OneDrive\Documents\lib\site-packages\sklearn\base.py in _validate_data(self, X, y, reset, validate_separately, **check_params)
  419         out = X
  420     elif isinstance(y, str) and y == 'no_validation':
--> 421         X = check_array(X, **check_params)
  422         out = X
  423     else:

~\OneDrive\Documents\lib\site-packages\sklearn\utils\validation.py in inner_f(*args, **kwargs)
   61         extra_args = len(args) - len(all_args)
   62         if extra_args <= 0:
--> 63             return f(*args, **kwargs)
   64
   65         # extra_args > 0

~\OneDrive\Documents\lib\site-packages\sklearn\utils\validation.py in check_array(array, accept_sparse, accept_large_sparse, dtype, order, copy, force_all_finite, ensure_2d, allow_nd, ensure_min_samples, ensure_min_features, estimator)
  671         array = array.astype(dtype, casting="unsafe", copy=False)
  672     else:
--> 673         array = np.asarray(array, order=order, dtype=dtype)
  674     except ComplexWarning as complex_warning:
  675         raise ValueError("Complex data not supported\n"

~\OneDrive\Documents\lib\site-packages\numpy\core\asarray.py in asarray(a, dtype, order, like)
  100     return _asarray_with_like(a, dtype=dtype, order=order, like=like)
  101
--> 102     return array(a, dtype, copy=False, order=order)
  103
  104

ValueError: could not convert string to float: 'normal'

```

```
In [155]: from sklearn.linear_model import LinearRegression
from sklearn.preprocessing import PolynomialFeatures
from sklearn.metrics import mean_squared_error, r2_score
from sklearn.datasets import load_boston
from sklearn.model_selection import train_test_split

# Load the Boston Housing dataset
df=pd.read_csv('kidney_disease.csv')

# Split the dataset into training and testing sets
X_train, X_test, y_train, y_test = train_test_split(X,y,test_size=0.2, random_state=42)

# Define a polynomial regression model
poly_features = PolynomialFeatures(degree=2)
X_train_poly = poly_features.fit_transform(X_train)
X_test_poly = poly_features.transform(X_test)
model = LinearRegression()
model.fit(X_train_poly, y_train)

# Use the trained model to make predictions on the testing data
y_pred = model.predict(X_test_poly)

# Calculate the mean squared error and R-squared as accuracy metrics
mse = mean_squared_error(y_test, y_pred)
r2 = r2_score(y_test, y_pred)
print('Mean Squared Error:', mse)
print('R-squared:', r2)
```

```

-----
ValueError                                Traceback (most recent call last)
~\AppData\Local\Temp\ipykernel_29180\517723795.py in <module>
      13 # Define a polynomial regression model
      14 poly_features = PolynomialFeatures(degree=2)
--> 15 X_train_poly = poly_features.fit_transform(X_train)
      16 X_test_poly = poly_features.transform(X_test)
      17 model = LinearRegression()

~\OneDrive\Documents\lib\site-packages\sklearn\base.py in fit_transform(self, X, y, **fit_params)
   697         if y is None:
   698             # fit method of arity 1 (unsupervised transformation)
--> 699             return self.fit(X, **fit_params).transform(X)
   700         else:
   701             # fit method of arity 2 (supervised transformation)

~\OneDrive\Documents\lib\site-packages\sklearn\preprocessing\_data.py in fit(self, X, y)
  1702         Fitted transformer.
  1703         """
--> 1704     n_samples, n_features = self._validate_data(
  1705         X, accept_sparse=True).shape
  1706     combinations = self._combinations(n_features, self.degree,

~\OneDrive\Documents\lib\site-packages\sklearn\base.py in _validate_data(self, X, y, reset, validate_separately, **check_params)
  419         out = X
  420     elif isinstance(y, str) and y == 'no_validation':
--> 421         X = check_array(X, **check_params)
  422         out = X
  423     else:

~\OneDrive\Documents\lib\site-packages\sklearn\utils\validation.py in inner_f(*args, **kwargs)
   61         extra_args = len(args) - len(all_args)
   62         if extra_args <= 0:
--> 63             return f(*args, **kwargs)
   64
   65         # extra_args > 0

~\OneDrive\Documents\lib\site-packages\sklearn\utils\validation.py in check_array(array, accept_sparse, accept_large_sparse, dtype, order, copy, force_all_finite, ensure_2d, allow_nd, ensure_min_samples, ensure_min_features, estimator)
  671         array = array.astype(dtype, casting="unsafe", copy=False)
  672     else:
--> 673         array = np.asarray(array, order=order, dtype=dtype)
  674     except ComplexWarning as complex_warning:
  675         raise ValueError("Complex data not supported\n"

~\OneDrive\Documents\lib\site-packages\numpy\core\asarray.py in asarray(a, dtype, order, like)
  100     return _asarray_with_like(a, dtype=dtype, order=order, like=like)
  101
--> 102     return array(a, dtype, copy=False, order=order)
  103
  104

ValueError: could not convert string to float: 'normal'

```

In []:

In []: