Milestone2:

```
import pandas as pd
import numpy as np
from collections import Counter as c
import matplotlib.pyplot as plt
import seaborn as sns
import missingno as msno
from sklearn.metrics import accuracy_score, confusion_matrix
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
from sklearn.linear_model import LogisticRegression
import pickle
data=pd.read csv('/content/kidney disease[1].csv')
data.head()
        id age
                         sg al su
                                        rbc
                                                                       ba ... pcv
                   bp
                                                            рсс
                                                   рс
        0 48.0 80.0 1.020 1.0 0.0
                                        NaN
                                               normal notpresent notpresent
                                                                                    780
                                                                                 44
            7.0 50.0 1.020 4.0 0.0
         1
                                        NaN
                                               normal notpresent notpresent
                                                                                 38
                                                                                    600
         2 62.0 80.0 1.010 2.0 3.0 normal
                                               normal notpresent notpresent
                                                                                 31 750
         3 48.0 70.0 1.005 4.0 0.0 normal
                                             abnormal
                                                         present notpresent
                                                                                 32 670
        4 51.0 80.0 1.010 2.0 0.0 normal
                                                                                 35 730
                                               normal notpresent notpresent
     5 rows × 26 columns
      1
data.columns
     'appet', 'pe', 'ane', 'classification'],
           dtype='object')
'hemoglobin','packed_cell_volume','white_blood_cell_count','red_blood_cell_count',
'hypertension','diabetesmellitus','coronary_artery_disease','appetite',
              'pedal_edema', 'anemia', 'class']
data.columns
     Index(['age', 'blood_pressure', 'specific_gravity', 'albumin', 'sugar',
             red_blood_cells', 'pus_cell', 'pus_cell_clumps', 'bacteria',
            'blood glucose random', 'blood_urea', 'serum_creatine', 'sodium',
            'potassium', 'hemoglobin', 'packed_cell_volume', 'white_blood_cell_count', 'red_blood_cell_count', 'hypertension',
            'diabetesmellitus', 'coronary_artery_disease', 'appetite',
           'pedal_edema', 'anemia', 'class'], dtype='object')
data.info()
     <class 'pandas.core.frame.DataFrame'>
     RangeIndex: 400 entries, 0 to 399
     Data columns (total 25 columns):
         Column
                                  Non-Null Count Dtype
      #
     ---
      0
          age
                                  391 non-null
                                                  float64
          blood_pressure
                                  388 non-null
                                                  float64
      1
                                  353 non-null
          specific_gravity
                                  354 non-null
      3
          albumin
                                                  float64
                                  351 non-null
      4
          sugar
                                                  float64
          red_blood_cells
                                  248 non-null
                                                  object
      5
      6
          pus_cell
                                  335 non-null
                                                  object
          pus_cell_clumps
                                  396 non-null
                                                  object
          bacteria
                                   396 non-null
                                                  object
      9
          blood glucose random
                                   356 non-null
                                                  float64
      10 blood_urea
                                  381 non-null
                                                  float64
         serum_creatine
                                  383 non-null
```

```
313 non-null
                                                  float64
         sodium
         potassium
                                                  float64
                                  312 non-null
     13
                                  348 non-null
     14
         hemoglobin
                                                  float64
     15
         packed_cell_volume
                                  330 non-null
                                                  object
         white_blood_cell_count
                                  295 non-null
                                                  object
     16
         red_blood_cell_count
                                  270 non-null
     17
                                                  object
     18 hypertension
                                  398 non-null
                                                  object
     19
        diabetesmellitus
                                  398 non-null
                                                  object
     20 coronary_artery_disease 398 non-null
                                                  object
     21 appetite
                                  399 non-null
                                                  object
     22 pedal_edema
                                  399 non-null
                                                  object
     23
                                  399 non-null
         anemia
                                                  object
                                  400 non-null
     24 class
                                                  object
     dtypes: float64(11), object(14)
    memory usage: 78.2+ KB
data.isnull().sum()
                                 9
    blood_pressure
                                12
     specific_gravity
                                47
     albumin
                                46
     sugar
                                49
     red_blood_cells
                               152
    pus cell
                                65
     pus cell clumps
                                 4
                                 4
     bacteria
     blood glucose random
                                44
    blood urea
                                19
     serum_creatine
                                17
     sodium
                                87
     potassium
                                88
                                52
     hemoglobin
    packed_cell_volume
     white_blood_cell_count
                               105
     red_blood_cell_count
                               130
     hypertension
                                 2
     diabetesmellitus
                                 2
     {\tt coronary\_artery\_disease}
     appetite
                                 1
     pedal_edema
                                 1
     anemia
                                 1
                                 0
     dtype: int64
data['blood glucose random'].fillna(data['blood glucose random'].mean(),inplace=True)
data['blood_pressure'].fillna(data['blood_pressure'].mean(),inplace=True)
data['blood_urea'].fillna(data['blood_urea'].mean(),inplace=True)
data['hemoglobin'].fillna(data['hemoglobin'].mean(),inplace=True)
data['packed_cell_volume'].fillna(data['packed_cell_volume'].mean(),inplace=True)
data['potassium'].fillna(data['potassium'].mean(),inplace=True)
data['red_blood_cell_count'].fillna(data['red_blood_cell_count'].mean(),inplace=True)
data['serum_creatine'].fillna(data['serum_creatine'].mean(),inplace=True)
data['sodium'].fillna(data['sodium'].mean(),inplace=True)
data['white_blood_cell_count'].fillna(data['white_blood_cell_count'].mean(),inplace=True)
data['age'].fillna(data['age'].mode()[0],inplace=True)
{\tt data['hypertension'].fillna(data['hypertension'].mode()[0],inplace=True)}
data['pus_cell_clumps'].fillna(data['pus_cell_clumps'].mode()[0],inplace=True)
data['appetite'].fillna(data['appetite'].mode()[0],inplace=True)
data['albumin'].fillna(data['albumin'].mode()[0],inplace=True)
data['pus_cell'].fillna(data['pus_cell'].mode()[0],inplace=True)
data['red_blood_cells'].fillna(data['red_blood_cells'].mode()[0],inplace=True)
data['bacteria'].fillna(data['bacteria'].mode()[0],inplace=True)
data['anemia'].fillna(data['anemia'].mode()[0],inplace=True)
data['sugar'].fillna(data['sugar'].mode()[0],inplace=True)
data['diabetesmellitus'].fillna(data['diabetesmellitus'].mode()[0],inplace=True)
data['pedal_edema'].fillna(data['pedal_edema'].mode()[0],inplace=True)
data['specific_gravity'].fillna(data['specific_gravity'].mode()[0],inplace=True)
contcols=set(data.dtypes[data.dtypes!='0'].index.values)
#contcols=pd.DataFrame(data,columns=contcols)
print(contcols)
     {'sg', 'rbc', 'id', 'pot', 'bgr', 'pcc', 'su', 'sc', 'pe', 'wc', 'rc', 'ba', 'sod', 'classification', 'pcv', 'pc', 'age', 'dm', 'ht
    4
for i in catcols:
  print("Colummns:",i)
```

```
print(c(data[i]))
 print('*'*120+'\n')
    Colummns: sugar
    Counter({0.0: 339, 2.0: 18, 3.0: 14, 4.0: 13, 1.0: 13, 5.0: 3})
    Colummns: specific_gravity
    Counter({1.02: 153, 1.01: 84, 1.025: 81, 1.015: 75, 1.005: 7})
    Colummns: albumin
    Counter({0.0: 245, 1.0: 44, 2.0: 43, 3.0: 43, 4.0: 24, 5.0: 1})
#'specific_gravity','albumin','sugar'(as these columns are numerical it is removed)
from sklearn.preprocessing import LabelEncoder
for i in catcols:
 print("LABEL ENCODING OF:",i)
 LEi=LabelEncoder()
 print(c(data[i]))
 data[i] = LEi.fit_transform(data[i])
 print(c(data[i]))
 print("*"*100)
contcols=set(data.dtypes[data.dtypes!='0'].index.values)
#contcols=pd.DataFrame(data,columns=contcols)
print(contcols)
    {'sg', 'rbc', 'id', 'pot', 'bgr', 'pcc', 'su', 'sc', 'pe', 'wc', 'rc', 'ba', 'sod', 'classification', 'pcv', 'pc', 'age', 'dm', 'ht
for i in contcols:
 print("Continous Columns:",i)
 print(c(data[i]))
 print('*'*120+'\n')
```

```
Continous Columns: al
           Counter({0.0: 199, 1.0: 44, 2.0: 43, 3.0: 43, 4.0: 24, nan: 1, nan: 1,
           Continous Columns: bp
           Counter({80.0: 116, 70.0: 112, 60.0: 71, 90.0: 53, 100.0: 25, 50.0: 5, 110.0: 3, nan: 1, nan: 1, 140.0: 1, 180.0: 1, nan: 1, nan
           Continous Columns: appet
           Counter({'good': 317, 'poor': 82, nan: 1})
contcols.remove('specific_gravity')
contcols.remove('albumin')
contcols.remove('sugar')
print(contcols)
           {'blood_urea', 'pus_cell_clumps', 'red_blood_cells', 'coronary_artery_disease', 'appetite', 'potassium', 'pedal_edema', 'anemia',
contcols.add('red_blood_cell_count')
contcols.add('packed_cell_volume')
contcols.add('white_blood_cell_count')
print(contcols)
           {'sg', 'rbc', 'id', 'pot', 'bgr', 'pcc', 'su', 'sc', 'pe', 'wc', 'rc', 'ba', 'white_blood_cell_count', 'red_blood_cell_count', 'pac
catcols.add('specific_gravity')
catcols.add('albumin')
catcols.add('sugar')
print(catcols)
           {'sugar', 'specific_gravity', 'albumin'}
data['coronary_artery_disease'] = data.coronary_artery_disease.replace('\tno','no')
           NameError
                                                                                                    Traceback (most recent call last)
           <ipython-input-183-b86ea066d7e2> in <cell line: 1>()
            ----> 1 data['coronary_artery_disease'] = dataset.coronary_artery_disease.replace('\tno','no')
           NameError: name 'dataset' is not defined
             SEARCH STACK OVERFLOW
data['diabetesmellitus'] = data.diabetesmellitus.replace(to_replace={'\tno':'no', \types':'yes'.'yes':})
c(data[diabetesmellitus])
               File "<ipython-input-180-d85ff3d59297>", line 1
                   data['diabetesmellitus'] = data.diabetesmellitus.replace(to_replace={'\tno':'no','\types':'yes'.'})
           SyntaxError: invalid syntax
             SEARCH STACK OVERFLOW
 milestone3
data.describe()
```

blood age blood_pressure specific_gravity albumin sugar glucose blood_urea serum_creatine sodium pota random

---- E4 400070 76 400070 4 047400 4 046040 0 450440 440 006647 E7 406700 2 070464 407 500764 407 50

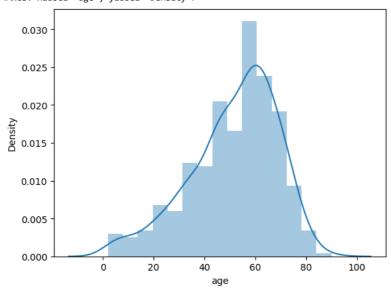
<ipython-input-45-868c85374ad7>:1: UserWarning:

`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751

sns.distplot(data.age)
<Axes: xlabel='age', ylabel='Density'>



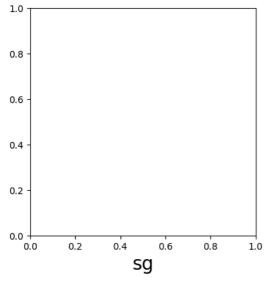
```
import matplotlib.pyplot as plt
fig=plt.figure(figsize=(5,5))
plt.scatter(data['age'],data['blood_pressure'],color='blue')
plt.xlabel('age')
plt.ylabel('blood pressure')
plt.title("age vs blood scatter plot")
```

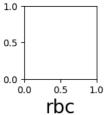
Text(0.5, 1.0, 'age vs blood scatter plot')

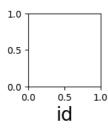


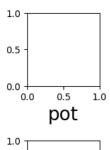
```
plt.figure(figsize=(20,15),facecolor='white')
plotnumber=1

for column in contcols:
   if plotnumber<=11:
      ax=plt.subplot(3,4,plotnumber)
      plt.xlabel(column,fontsize=20)
      plotnumber+=1
      plt.show()</pre>
```



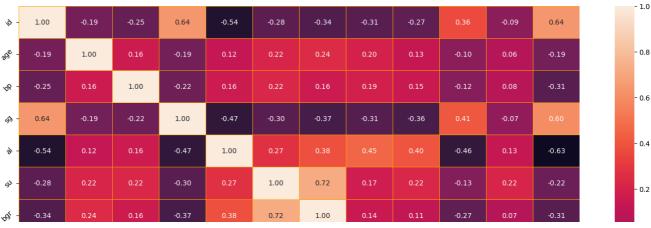






#HEAT MAP #correlation of parameters
f,ax=plt.subplots(figsize=(18,10))
sns.heatmap(data.corr(),annot=True,fmt=".2f",ax=ax,linewidths=0.5,linecolor="orange")
plt.xticks(rotation=45)
plt.yticks(rotation=45)
plt.show()

<ipython-input-147-9d5ae916c1e3>:3: FutureWarning: The default value of numeric_only in DataFrame.corr is deprecated. In a future \(\) sns.heatmap(data.corr(),annot=True,fmt=".2f",ax=ax,linewidths=0.5,linecolor="orange")



```
sns.countplot(data['class'])
     KeyError
                                                Traceback (most recent call last)
     /usr/local/lib/python3.9/dist-packages/pandas/core/indexes/base.py in get_loc(self, key, method, tolerance)
        3801
        3802
                             return self._engine.get_loc(casted_key)
        3803
                         except KeyError as err:
                                     4 frames -
     pandas/_libs/hashtable_class_helper.pxi in pandas._libs.hashtable.PyObjectHashTable.get_item()
     pandas/_libs/hashtable_class_helper.pxi in pandas._libs.hashtable.PyObjectHashTable.get_item()
     KeyError: 'class'
     The above exception was the direct cause of the following exception:
                                                Traceback (most recent call last)
     KeyError
     /usr/local/lib/python3.9/dist-packages/pandas/core/indexes/base.py in get_loc(self, key, method, tolerance)
                             return self._engine.get_loc(casted_key)
        3802
                         except KeyError as err:
        3803
     -> 3804
                              raise KeyError(key) from err
        3805
                         except TypeError:
                              # If we have a listlike key, _check_indexing_error will raise
        3806
     KeyError: 'class'
      SEARCH STACK OVERFLOW
\# perfroming feature scaling operation using standard scaller on X part of the dataset because
# there different type of values in the columns
from sklearn.preprocessing import StandardScaler
sc=StandardScaler()
x_bal=sc.fit_transform
selcols=['red_blood_cells','pus_cell','blood glucose random','blood_urea',
          'pedal_edema','anemia','diabetesmellitus','coronary_artery_disease']
x=pd.DataFrame(data,columns=selcols)
y=pd.DataFrame(data,columns=['class'])
print(x.shape)
print(y.shape)
     (400, 8)
from sklearn.model_selection import train_test_split
x\_train, x\_test, y\_train, y\_test=train\_test\_split(x,y,test\_size=0.2, random\_state=2)
print(x_train.shape)
print(y_train.shape)
print(x_test.shape)
print(y_test.shape)
     (320, 8)
     (320, 1)
     (80, 8)
     (80, 1)
```

Milestone4

```
#importing the keras libraries and packages
import tensorflow
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Dense
#creating ANN skleton view
classification = Sequential()
classification.add(Dense(30,activation='relu'))
classification.add(Dense(128,activation='relu'))
classification.add(Dense(64,activation='relu'))
classification.add(Dense(1,activation='sigmoid'))
#Compiling the ANN model
classification.compile(optimizer='adam',loss='binary_crossentropy',metrics=['accuracy'])
classification.fit(x_train,y_train,batch_size=10,validation_split=0.3,epochs=100)
     ValueError
                                              Traceback (most recent call last)
     <ipython-input-178-85d4caec70a2> in <cell line: 1>()
      ---> 1 classification.fit(x_train,y_train,batch_size=10,validation_split=0.3,epochs=100)
                                     - 💲 1 frames
     /usr/local/lib/python3.9/dist-packages/tensorflow/python/framework/constant_op.py in convert_to_eager_tensor(value, ctx, dtype)
         101
                 dtype = dtypes.as_dtype(dtype).as_datatype_enum
         102
               ctx.ensure_initialized()
              return ops.EagerTensor(value, ctx.device_name, dtype)
         104
         105
     ValueError: Failed to convert a NumPy array to a Tensor (Unsupported object type float).
      SEARCH STACK OVERFLOW
from sklearn.ensemble import RandomForestClassifier
rfc = RandomForestClassifier(n_estimators=10,criterion='entropy')
rfc.fit(x_train,y_train)
     ValueError
                                               Traceback (most recent call last)
     <ipython-input-165-b87bb2ba9825> in <cell line: 1>()
     ----> 1 rfc.fit(x_train,y_train)
                                     – 💲 5 frames -
     /usr/local/lib/python3.9/dist-packages/pandas/core/generic.py in __array__(self, dtype)
        2068
        2069
                 def __array__(self, dtype: npt.DTypeLike | None = None) -> np.ndarray:
     -> 2070
                     return np.asarray(self._values, dtype=dtype)
        2071
                 def __array_wrap__(
     ValueError: could not convert string to float: 'normal'
      SEARCH STACK OVERFLOW
y_predict = rfc.predict(x_test)
```

```
y_predict_train = rfc.predict(x_train)
     _____
                                              Traceback (most recent call last)
     <ipython-input-159-6d0b7f55f6ea> in <cell line: 1>()
     ----> 1 y_predict_train = rfc.predict(x_train)
                                   — 💲 6 frames 🗕
     /usr/local/lib/python3.9/dist-packages/pandas/core/generic.py in __array__(self, dtype)
        2068
        2069
                def __array__(self, dtype: npt.DTypeLike | None = None) -> np.ndarray:
     -> 2070
                     return np.asarray(self._values, dtype=dtype)
        2071
                def __array_wrap__(
     ValueError: could not convert string to float: 'normal'
      SEARCH STACK OVERFLOW
from sklearn.tree import DecisionTreeClassifier
dtc = DecisionTreeClassifier(max_depth=4,splitter='best,criterion='entropy')
       File "<ipython-input-160-aa38db87d527>", line 2
         dtc = DecisionTreeClassifier(max_depth=4,splitter='best,criterion='entropy')
     SyntaxError: invalid syntax
      SEARCH STACK OVERFLOW
dtc.fit(x train,y train)
                                             Traceback (most recent call last)
     <ipython-input-161-8bc0353a9ffd> in <cell line: 1>()
     ----> 1 dtc.fit(x_train,y_train)
     NameError: name 'dtc' is not defined
     SEARCH STACK OVERFLOW
y predict= dtc.predict(x test)
y_predict
                                              Traceback (most recent call last)
     <ipython-input-162-b3c3150919dc> in <cell line: 1>()
     ----> 1 y_predict= dtc.predict(x_test)
          2 y_predict
     NameError: name 'dtc' is not defined
      SEARCH STACK OVERFLOW
y_predict_train = dtc.predict(x_train)
     NameError
                                              Traceback (most recent call last)
     <ipython-input-163-4e71ba4a880d> in <cell line: 1>()
     ---> 1 y_predict_train = dtc.predict(x_train)
     NameError: name 'dtc' is not defined
      SEARCH STACK OVERFLOW
from sklearn.linear_model import LogisticRegression
lgr = LogisticRegression()
lgr.fit(x_train,y_train)
```

```
ValueError
                                                Traceback (most recent call last)
     <ipython-input-169-0be1d0fe3673> in <cell line: 1>()
      ----> 1 lgr.fit(x_train,y_train)
                                      – 💲 5 frames –
     /usr/local/lib/python3.9/dist-packages/pandas/core/generic.py in __array__(self, dtype)
        2069
                       _array__(self, dtype: npt.DTypeLike | None = None) -> np.ndarray:
      -> 2070
                      return np.asarray(self._values, dtype=dtype)
        2071
        2072
                 def array wrap (
     ValueError: could not convert string to float: 'normal'
from sklearn.metrics import accuracy_score,classification_report
y_predict = lgr.predict(x_test)
     ValueError
                                                Traceback (most recent call last)
     <ipython-input-171-361d7e5c1e35> in <cell line: 1>()
     ----> 1 y_predict = lgr.predict(x_test)
                                      – 💲 5 frames –
     /usr/local/lib/python3.9/dist-packages/pandas/core/generic.py in __array__(self, dtype)
        2068
        2069
                       array (self, dtype: npt.DTypeLike | None = None) -> np.ndarray:
      -> 2070
                      return np.asarray(self._values, dtype=dtype)
        2071
        2072
                 def __array_wrap__(
     ValueError: could not convert string to float: 'normal'
      SEARCH STACK OVERFLOW
y_pred = lgr.predict([[1,1,121.000000,36.0,0,0,1,0]])
print(y_pred)
y_pred
     /usr/local/lib/python3.9/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but LogisticRegressic
       warnings.warn(
                                                 Traceback (most recent call last)
     AttributeError
     <ipython-input-172-0bf88aa12548> in <cell line: 1>()
       --> 1 y_pred = lgr.predict([[1,1,121.000000,36.0,0,0,1,0]])
           2 print(y_pred)
           3 y_pred
                                      - 💲 1 frames -
     /usr/local/lib/python3.9/dist-packages/sklearn/linear_model/_base.py in decision_function(self, X)
         399
                      X = self._validate_data(X, accept_sparse="csr", reset=False)
         400
                      scores = safe_sparse_dot(X, self.coef_.T, dense_output=True) + self.intercept_
     --> 401
         402
                      return xp.reshape(scores, -1) if scores.shape[1] == 1 else scores
         403
     AttributeError: 'LogisticRegression' object has no attribute 'coef_'
      SEARCH STACK OVERFLOW
y_pred = dtc.predict([[1,1,121.000000,36.0,0,0,1,0]])
print(v predv)
y_predict_train
                                                 Traceback (most recent call last)
     cipython-input-173-d8b5d3f4ae4d> in <cell line: 1>()
---> 1 y_pred = dtc.predict([[1,1,121.000000,36.0,0,0,1,0]])
           2 print(y_predy)
           3 y_predict_train
     NameError: name 'dtc' is not defined
      SEARCH STACK OVERFLOW
y_pred = rfc.predict([[1,1,121.000000,36.0,0,0,1,0]])
print(y_pred)
```

```
y_pred
classification.save("ckd.h5")
y pred = classification.predict(x test)
v pred
y_pred = (y_pred > 0.5)
y_pred
def predict_exit(sample_value):
  # Convert list to numpy array
  sample_value = np.array(sample_value)
  #Reshape because sample_value contains only 1 record
  sample_value = sample_value.reshape(1,-1)
  #Feature Scaling
  sample value = sc,transform(sample value)
  return classifier.predict(sample_value)
test=classification.predict([[1,1,121,000000,36,0,0,0,1,0]])
if test==1:
 print('Prediction: Highh chance of CKD!')
else:
 print('Prediction:Low chance of CKD.')
milestone5
from sklearn import model_selection
dfs =[]
models =[
    ('LogReg',LogisticRegression()),
    ('RF',RandomForestClassifier()),
    ('DecisionTree',DecisionTreeClassifier)),
results =[]
 names =[]
 scoring =['accuracy','precision_weighted','recall_weighted','f1_weighted','roc_auc']
 target_names =['No CKD','CKD']
 for name, model in models:
   kfold=model_selection.kFkkold(n_splits=5,shuffle=True,random_state=90210)
   cv_results = model_selection.cross_validate(model,x_train,y_train,cv=kfold,scoring=scoring)
   clf = model.fit(x_train,y_train)
   y_pred = clf.predict(x_test)
   print(name)
   print(classification_report(y_test,y_pred,target_names=target_names))
   results.append(cv_results)
   names.append(name)
   this_df = pd.DataFrame(cv_results)
   this_df['model'] = name
   dfs.append(this_df)
final = pd.concat(dfs,ignore_index=True)
return final
       File "<ipython-input-132-e9b81462d286>", line 5
         ('DecisionTree', DecisionTreeClassifier)),
     SyntaxError: closing parenthesis ')' does not match opening parenthesis '[' on line 2
      SEARCH STACK OVERFLOW
from sklearn.metrics import confision_matrix
cm = confusion_matrix(y_test,y_predict)
cm
```

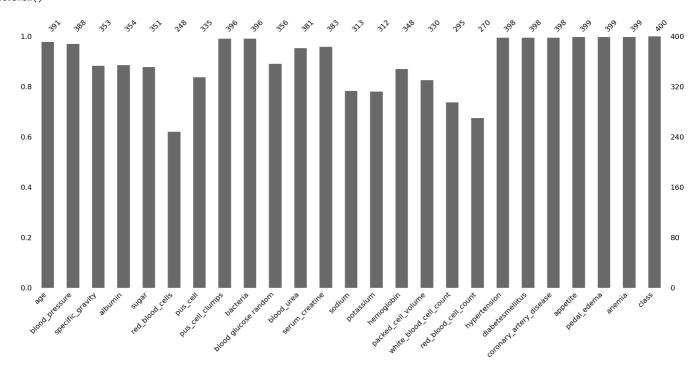
```
# Plotting confusion matrix
plt.figure(figsize=(8,6))
sns.heatmap(cm.cmap='Blues',annot=True,xticklabels=['no ckd','ckd'],yticklabels=['no,ckd','ckd'])
plt.xlabel('predicted values')
plt.ylabel('Actual values')
plt.title('Confusion Matrix for Logistic Regression model')
plt.show()
from sklearn.metrics import confision matrix
cm = confusion_matrix(y_test,y_predict)
cm
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plt.figure(figsize=(8,6))
sns.heatmap(cm.cmap='Blues',annot=True,xticklabels=['no ckd','ckd'],yticklabels=['no,ckd','ckd'])
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plt.xlabel('predicted values')
plt.ylabel('Actual values')
plt.title('Confusion Matrix for Logistic Regression model')
plt.show()
from google.colab import files
uploaded = files.upload()
     Choose Files No file chosen
                                       Upload widget is only available when the cell has been executed in the current browser session. Please rerun this cell to
     enable
     Saving kidney disease[1] csv to kidney disease[1] (1) csv
from sklearn.metrics import confision_matrix
cm = confusion_matrix(y_test,y_predict)
cm
# Plotting confusion matrix
plt.figure(figsize=(8,6))
sns.heatmap(cm.cmap='Blues',annot=True,xticklabels=['no ckd','ckd'],yticklabels=['no,ckd','ckd'])
plt.xlabel('predicted values')
plt.ylabel('Actual values')
plt.title('Confusion Matrix for Logistic Regression model')
plt.show()
bootstaps -[]
for model in list(set(final.model.values)):
 model_df = final.loc[final.model == model]
  bootstrap = madel_df.sample(n=30, replace=True)
  bootstraps.append(bootstrap)
bootstrap_df = pd.concat(bootstraps,ignore_ignore_index=True)
results_long = pd.melt(bootstrap_df,id_vars=['model'],var_name='metrics',value_name='values')
time_metrics+['fit_time','score_time']
## PERFORMANCE METRICS
results_long_nofit = results_long.loc[~results_long['metrics'].isin(time_matrics)]
results_long_nofit = results_long_nofit.sort_values(by='values')
```

data.isnull().sum()

```
## TIME METRICS
results_long_nofit = results_long.loc[~results_long['metrics'].isin(time_matrics)]
results_long_nofit = results_long_nofit.sort_values(by='values')
import matplotlib.pyplot as plt
import seaborn as sns
plt.figure(figsize=(20,12))
sns.set(font scale=2.5)
g = sns.boxplot(x="model",y="values",hue="metrics",data=results_long_nofit,palette="Set3")
plt.legend(bbox_to_anchor=(1.05,1), loc=2, borderxespad=0)
plt.title("Comparison of Model by Classification Metric")
plt.savefig('./benchmark_models_performance.png',dpi=300)
Milestone6
pickle.dump(lgr, open('CKD.pk1','wb'))
data.tail()
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     10 rows × 26 columns
data.drop(["id"],axis=1,inplace=True)
data['class'].unique()
     array(['ckd', 'ckd\t', 'notckd'], dtype=object)
data['class']=data['class'].replace("ckd\t","ckd")
data['class'].unique()
     array(['ckd', 'notckd'], dtype=object)
np.unique(data.dtypes,return_counts=True)
     (array([dtype('float64'), dtype('0')], dtype=object), array([11, 14]))
```

age	9
blood_pressure	12
specific_gravity	47
albumin	46
sugar	49
red_blood_cells	152
pus_cell	65
pus_cell_clumps	4
bacteria	4
blood glucose random	44
blood_urea	19
serum_creatine	17
sodium	87
potassium	88
hemoglobin	52
packed_cell_volume	70
white_blood_cell_count	105
red_blood_cell_count	130
hypertension	2
diabetesmellitus	2
coronary_artery_disease	2
appetite	1
pedal_edema	1
anemia	1
class	0
dtype: int64	

```
#sns.heatmap(data.isnull(), cbar=False)
msno.bar(data)
plt.show()
```



data.packed_cell_volume = pd.to_numeric(data.packed_cell_volume,errors='coerce')
data.white_blood_cell_count = pd.to_numeric(data.white_blood_cell_count,errors='coerce')
data.red_blood_cell_count = pd.to_numeric(data.red_blood_cell_count,errors='coerce')

data.isnull().sum()

```
0
blood_pressure
specific_gravity
                            0
albumin
                            0
sugar
red_blood_cells
                            0
pus_cell
                            0
                            0
0
pus_cell_clumps
bacteria
                            0
blood glucose random
blood_urea
serum_creatine
                            0
sodium
potassium
                            0
hemoglobin
packed_cell_volume
```

```
white_blood_cell_count
     red_blood_cell_count
     hypertension
     diabetesmellitus
     coronary_artery_disease
                                 0
     appetite
     pedal_edema
                                 a
     anemia
                                 0
     class
                                 0
     dtype: int64
data['class'].unique()
     array(['ckd', 'notckd'], dtype=object)
selcols=['red_blood_cells','pus_cell','blood glucose random','blood_urea',
         'pedal_edema','anemia','diabetesmellitus','coronary_artery_disease']
x=pd.DataFrame(data,columns=selcols)
y=pd.DataFrame(data,columns=['class'])
print(x.shape)
print(y.shape)
     (400, 8)
     (400, 1)
from sklearn.model_selection import train_test_split
x\_train, x\_test, y\_train, y\_test=train\_test\_split(x, y, test\_size=0.2, random\_state=2)
print(x_train.shape)
print(y_train.shape)
print(x_test.shape)
print(y_test.shape)
     (320, 8)
     (320, 1)
(80, 8)
     (80, 1)
from sklearn.preprocessing import LabelEncoder
for i in catcols:
  print("LABEL ENCODING OF:",i)
  LEi=LabelEncoder()
  print(c(data[i]))
  data[i]=LEi.fit_transform(data[i])
  print(c(data[i]))
  print("*"*100)
data.describe()
```

	age	blood_pressure	specific_gravity	albumin	sugar	blood glucose random	blood_urea	serum_creatine	sodium	potas
count	400.000000	400.000000	400.000000	400.00000	400.000000	400.000000	400.000000	400.000000	400.000000	400.00
mean	51.675000	76.469072	1.017712	0.90000	0.395000	148.036517	57.425722	3.072454	137.528754	4.62
std	17.022008	13.476298	0.005434	1.31313	1.040038	74.782634	49.285887	5.617490	9.204273	2.81
min	2.000000	50.000000	1.005000	0.00000	0.000000	22.000000	1.500000	0.400000	4.500000	2.50
25%	42.000000	70.000000	1.015000	0.00000	0.000000	101.000000	27.000000	0.900000	135.000000	4.00
50%	55.000000	78.234536	1.020000	0.00000	0.000000	126.000000	44.000000	1.400000	137.528754	4.62
75%	64.000000	80.000000	1.020000	2.00000	0.000000	150.000000	61.750000	3.072454	141.000000	4.80
max	90.000000	180.000000	1.025000	5.00000	5.000000	490.000000	391.000000	76.000000	163.000000	47.00