```
import numpy as np
import pandas as pd
from sklearn.impute import SimpleImputer
import collections
from sklearn.model selection import train test split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score,confusion_matrix,classification_report
df=pd.read csv('train.csv')
print('Shape of the data',df.shape)
    Shape of the data (9557, 143)
print(df.head())
                 Ιd
                         v2a1 hacdor rooms hacapo v14a refrig v18q
                                                                       v18q1 \
    0 ID 279628684 190000.0
                                                 0
                                                       1
                                                              1
                                                                         NaN
    1 ID f29eb3ddd 135000.0
                                   0
                                         4
                                                 0
                                                       1
                                                              1
                                                                         1.0
    2 ID 68de51c94
                         NaN
                                   a
                                         8
                                                 0
                                                      1
                                                              1
                                                                         NaN
    3 ID d671db89c 180000.0
                                         5
                                                      1
                                                              1
                                                                         1.0
    4 ID d56d6f5f5 180000.0
       r4h1 ... SQBescolari SQBage SQBhogar_total SQBedjefe SQBhogar nin \
          0 ...
                         100
                               1849
                                                 1
                                                           100
                                                                          0
          0 ...
                         144
                               4489
                                                 1
                                                           144
                                                                          0
          0 ...
                         121 8464
                                                            0
                                                                          0
          0 ...
                                                 16
    3
                          81
                                 289
                                                           121
                                                                          4
          0
                         121
                               1369
                                                 16
                                                           121
                                                                          4
        SQBovercrowding SQBdependency SQBmeaned agesq
                                                       Target
              1.000000
                                 0.0
                                         100.0
                                                 1849
              1.000000
                                64.0
                                         144.0
                                                 4489
    1
    2
              0.250000
                                64.0
                                         121.0
                                                 8464
                                                           4
    3
              1.777778
                                 1.0
                                         121.0
                                                  289
                                                           4
              1.777778
                                 1.0
                                         121.0
                                                 1369
    [5 rows x 143 columns]
data_train_info = pd.DataFrame(columns=['Name of Col', 'Num of Null', 'Dtype', 'N_Unique' , 'Null Perc'])
for i in range(0, len(df.columns)):
   data_train_info.loc[i] = [df.columns[i],
                       df[df.columns[i]].isnull().sum(),
                       df[df.columns[i]].dtypes,
                       df[df.columns[i]].nunique(),
                       df[df.columns[i]].isnull().sum()*100/df.shape[0]]
data_train_info
```

	Name of Col	Num of Null	Dtype	N_Unique	Null Perc
0	Id	0	object	9557	0.000000
1	v2a1	6860	float64	157	71.779847
2	hacdor	0	int64	2	0.000000
3	rooms	0	int64	11	0.000000
4	hacapo	0	int64	2	0.000000
138	SQBovercrowding	0	float64	38	0.000000
139	SQBdependency	0	float64	31	0.000000
140	SQBmeaned	5	float64	155	0.052318

data\_train\_info[data\_train\_info['Num of Null']>0]

	Name of Col	Num of Null	Dtype	N_Unique	Null Perc
1	v2a1	6860	float64	157	71.779847
8	v18q1	7342	float64	6	76.823271
21	rez_esc	7928	float64	6	82.954902
103	meaneduc	5	float64	155	0.052318
140	SQBmeaned	5	float64	155	0.052318

```
#Percentage of null values in v2a1, v18q1, rez_esc is more than 50%. So, these columns are dropped
df= df.drop(['v2a1','v18q1','rez_esc'],axis=1)
print(df.shape)
```

```
(9557, 140)
```

```
#Find Column with mixed values
df= df.drop(['Id'],axis=1)
df.describe(include='0')
```

```
idhogar dependency edjefe edjefa
      count
                  9557
                              9557
                                      9557
                                              9557
      unique
                  2988
                                31
                                        22
                                                22
# Dependency replace yes with 0.5 and no with 0
df.dependency = df.dependency.replace(to_replace=['yes','no'],value=[0.5,0]).astype('float')
# edjefe replace yes with median and no with zero
med 1=np.median(df.edjefe[df.edjefe.isin(['yes','no'])==False].astype('float'))
df.edjefe= df.edjefe.replace(to_replace=['yes','no'],value=[med_1,0]).astype('float')
# edjefa replace yes with median and no with zero
med 2=np.median(df.edjefa[df.edjefa.isin(['yes','no'])==False].astype('float'))
df.edjefa= df.edjefa.replace(to_replace=['yes','no'],value=[med_2,0]).astype('float')
df.describe(include='0')
               idhogar
      count
                  9557
      unique
                  2988
             fd8a6d014
       top
       freq
                    13
print(df.idhogar.nunique())
     2988
   3. Finding biasness in the dataset
df.Target.value_counts()
import collections
print(df.shape)
collections.Counter(df['Target'])
     (9557, 139)
     Counter({4: 5996, 2: 1597, 3: 1209, 1: 755})
   4. Checking whether all members of the house have the same poverty level. bold text
```

poverty\_level=(df.groupby('idhogar')['Target'].nunique()>1).index
print(poverty\_level)

## 5. Checking if there is a house without a family head.

# 6. Set poverty level of the members and the head of the house same in a family.

```
target_mean=df.groupby('idhogar')['Target'].mean().astype('int64').reset_index().rename(columns={'Target':'Target_mean'})
df=df.merge(target_mean,how='left',on='idhogar')
df.Target=df.Target_mean
df.drop('Target_mean',axis=1,inplace=True)
df.head()
```

	hacdor	rooms	hacapo	v14a	refrig	v18q	r4h1	r4h2	r4h3	r4m1	•••	SQBescolari	SQBage	SQBhogar
0	0	3	0	1	1	0	0	1	1	0		100	1849	
1	0	4	0	1	1	1	0	1	1	0		144	4489	
2	0	8	0	1	1	0	0	0	0	0		121	8464	
3	0	5	0	1	1	1	0	2	2	1		81	289	
4	0	5	0	1	1	1	0	2	2	1		121	1369	

5 rows × 139 columns



# **Deploying Random Forest Classifier**

```
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2,random_state=10)
rfc = RandomForestClassifier(criterion= 'gini', n estimators=100)
rfc.fit(x_train,y_train)
pred=rfc.predict(x test)
# Check the accuracy using random forest
print('Accuracy score: ', accuracy_score(pred,y_test))
print()
print('Confusion matrix:\n', confusion matrix(pred,y test))
print()
print('Classification report:\n', classification_report(pred,y_test))
    Accuracy score: 0.9325313807531381
    Confusion matrix:
     [[ 141  3  1  3]
     [ 3 267 3 5]
        0
            1 175 2]
        25 46 37 1200]]
    Classification report:
                   precision
                               recall f1-score
                                                 support
               1
                       0.83
                                0.95
                                          0.89
                                                    148
               2
                       0.84
                                0.96
                                          0.90
                                                    278
               3
                       0.81
                                0.98
                                          0.89
                                                    178
               4
                      0.99
                                0.92
                                                   1308
                                          0.95
                                          0.93
                                                   1912
        accuracy
       macro avg
                       0.87
                                0.95
                                          0.91
                                                   1912
     weighted avg
                       0.94
                                0.93
                                          0.93
                                                   1912
```

## Check the accuracy using random forest with cross validation.

```
from sklearn.model_selection import KFold,cross_val_score

seed=7
kfold=KFold(n_splits=5,random_state=seed,shuffle=True)

rmclassifier=RandomForestClassifier(random_state=10)
#print(cross_val_score(rmclassifier,x_train,y_train,cv=kfold,scoring='accuracy'))
```

```
results=cross_val_score(rmclassifier,x_train,y_train,cv=kfold,scoring='accuracy')
print(results)
print(results.mean()*100)

[> [0.91563113 0.89535644 0.8940484 0.90843689 0.91628515]
90.595160235448
```

#### **Predict for Test Data**

df test.describe(include='0')

	Id	idhogar	dependency	edjefe	edjefa	1
count	23856	23856	23856	23856	23856	
unique	23856	7352	35	22	22	
top	ID_2f6873615	8e9159699	yes	no	no	
freq	1	13	5388	9056	15845	

```
# Dependency replace yes with 0.5 and no with 0
df_test.dependency = df_test.dependency.replace(to_replace=['yes','no'],value=[0.5,0]).astype('float')
# edjefe replace yes with median and no with zero
med_l=np.median(df_test.edjefe[df_test.edjefe.isin(['yes','no'])==False].astype('float'))
df_test.edjefe= df_test.edjefe.replace(to_replace=['yes','no'],value=[med_1,0]).astype('float'))
# edjefa replace yes with median and no with zero
med_2=np.median(df_test.edjefa[df_test.edjefa.isin(['yes','no'])==False].astype('float'))
df_test.edjefa= df_test.edjefa.replace(to_replace=['yes','no'],value=[med_2,0]).astype('float')

df_test= df_test.drop(['idhogar'],axis=1)
df_test.shape
```

```
1/29/23, 10:10 AM
```

✓ 0s completed at 10:09 AM

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