#importing pandas library
import pandas as pd

#reading the csv file
data=pd.read_csv("/content/01.Data Cleaning and Preprocessing.csv")

#displaying the first five rows
data.head()

	Observation	Y- Kappa	ChipRate	BF- CMratio	BlowFlow	ChipLevel4	T- upperExt- 2	T- lowerExt- 2
0	31-00:00	23.10	16.520	121.717	1177.607	169.805	358.282	329.545
1	31-01:00	27.60	16.810	79.022	1328.360	341.327	351.050	329.067
2	31-02:00	23.19	16.709	79.562	1329.407	239.161	350.022	329.260
3	31-03:00	23.60	16.478	81.011	1334.877	213.527	350.938	331.142
4	31-04:00	22.90	15.618	93.244	1334.168	243.131	351.640	332.709
5 rc	ws × 23 column	ıs						

#displaying the last five rows
data.tail()

	Observation	Y- Kappa	ChipRate	BF- CMratio	BlowFlow	ChipLevel4	T- upperExt- 2	T lowerExt
319	10-16:00	23.75	12.667	93.450	1178.252	276.955	347.286	310.97
320	9-19:00	19.80	12.558	94.352	1184.119	297.071	399.135	319.57
321	9-20:00	23.01	12.550	90.842	1188.517	289.826	373.633	314.59
322	9-21:00	24.32	13.083	88.910	1192.879	318.006	364.081	308.55
323	9-22:00	25.75	13.417	85.451	1186.342	248.312	356.289	310.48
5 row	rs × 23 columns							

displaying information about the dataset
data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 324 entries, 0 to 323
Data columns (total 23 columns):

#	Column	Non-Null Count	Dtype
0	Observation	324 non-null	object
1	Y-Kappa	324 non-null	float64
2	ChipRate	319 non-null	float64
3	BF-CMratio	307 non-null	float64
4	BlowFlow	308 non-null	float64
5	ChipLevel4	323 non-null	float64
6	T-upperExt-2	322 non-null	float64
7	T-lowerExt-2	322 non-null	float64
8	UCZAA	299 non-null	float64
9	WhiteFlow-4	323 non-null	float64
10	AAWhiteSt-4	173 non-null	float64
11	AA-Wood-4	323 non-null	float64
12	ChipMoisture-4	323 non-null	float64
13	SteamFlow-4	323 non-null	float64
14	Lower-HeatT-3	322 non-null	float64
15	Upper-HeatT-3	322 non-null	float64
16	ChipMass-4	323 non-null	float64
17	WeakLiquorF	323 non-null	float64
18	BlackFlow-2	322 non-null	float64
19	WeakWashF	323 non-null	float64
20	SteamHeatF-3	322 non-null	float64
21	T-Top-Chips-4	323 non-null	float64
22	SulphidityL-4	173 non-null	float64
dtyp	es: float64(22),	object(1)	
memo	ry usage: 58.3+ K	(B	

#displaying the type of the data
type(data)

```
pandas.core.frame.DataFrame

def __init__(data=None, index: Axes | None=None, columns: Axes | None=None,

dtype: Dtype | None=None, copy: bool | None=None) -> None

/usr/local/lib/python3.10/dist-packages/pandas/core/frame.py.

Two-dimensional, size-mutable, potentially heterogeneous tabular data.

Data structure also contains labeled axes (rows and columns).

Arithmetic operations align on both row and column labels. Can be
```

#displaying the datatype of columns
data.dtypes

```
Observation
                   object
                  float64
Y-Kappa
ChipRate
                  float64
BF-CMratio
                  float64
BlowFlow
                  float64
ChipLevel4
                  float64
T-upperExt-2
                  float64
T-lowerExt-2
                  float64
UCZAA
                  float64
WhiteFlow-4
                  float64
                  float64
AAWhiteSt-4
AA-Wood-4
                  float64
ChipMoisture-4
                  float64
                  float64
SteamFlow-4
Lower-HeatT-3
                  float64
                  float64
Upper-HeatT-3
                  float64
ChipMass-4
WeakLiquorF
                  float64
BlackFlow-2
                  float64
WeakWashF
                  float64
SteamHeatF-3
                  float64
T-Top-Chips-4
                  float64
SulphidityL-4
                  float64
dtype: object
```

#displaying the number of rows and columns data.shape

(324, 23)

#column names
data.columns

#descriptive statistics
data.describe()

	Ү-Карра	ChipRate	BF- CMratio	BlowFlow	ChipLevel4	T- upperExt- 2	lowerE
count	324.000000	319.000000	307.000000	308.000000	323.000000	322.000000	322.000
mean	20.635370	14.347937	87.464456	1237.837614	258.164483	356.904295	324.020
std	3.070036	1.499095	7.995012	100.593735	87.987452	9.209290	7.621
min	12.170000	9.983000	68.645000	0.000000	0.000000	339.168000	284.633
25%	18.382500	13.358000	81.823000	1193.215250	213.527000	350.241250	321.420
50%	20.845000	14.308000	86.739000	1273.138500	271.792000	356.843000	325.669
75%	23.032500	15.517000	92.372000	1289.196000	321.680000	362.242250	329.175
max	27.600000	16.958000	121.717000	1351.240000	419.014000	399.135000	337.012
8 rows ×	22 columns						

#Checking for duplicate rows
data.duplicated()

```
False
        False
2
        False
3
        False
        False
        ...
True
319
320
         True
321
         True
322
         True
323 True
Length: 324, dtype: bool
```

#total number of duplicates
data.duplicated().sum()

23

#removing the duplicates
data.drop_duplicates()

	Observation	Y- Kappa	ChipRate	BF- CMratio	BlowFlow	ChipLevel4	T- upperExt- 2	T lowerExt
0	31-00:00	23.10	16.520	121.717	1177.607	169.805	358.282	329.54
1	31-01:00	27.60	16.810	79.022	1328.360	341.327	351.050	329.06
2	31-02:00	23.19	16.709	79.562	1329.407	239.161	350.022	329.26
3	31-03:00	23.60	16.478	81.011	1334.877	213.527	350.938	331.14
4	31-04:00	22.90	15.618	93.244	1334.168	243.131	351.640	332.70
298	12-09:00	20.90	15.167	84.640	1283.706	339.440	354.803	311.04
299	12-10:00	24.98	NaN	85.034	1278.345	368.564	357.723	321.38
300	12-11:00	21.00	NaN	88.013	1307.722	278.842	357.438	323.75
301	12-12:00	21.40	NaN	85.490	1255.986	273.484	361.365	322.68
307	31-05:00	20.89	14.308	94.172	1327.832	251.120	351.263	332.48
301 rc	ows × 23 column	ıs						

#check for null values
data.isnull()

	Observation	Y- Kappa	ChipRate	BF- CMratio	BlowFlow	ChipLevel4	T- upperExt- 2	T lowerExt
0	False	False	False	False	False	False	False	Fals
1	False	False	False	False	False	False	False	Fals
2	False	False	False	False	False	False	False	Fals
3	False	False	False	False	False	False	False	Fals
4	False	False	False	False	False	False	False	Fals
319	False	False	False	False	False	False	False	Fals
320	False	False	False	False	False	False	False	Fals
321	False	False	False	False	False	False	False	Fals
322	False	False	False	False	False	False	False	Fals
323	False	False	False	False	False	False	False	Fals
324 rd	ows × 23 column	ns						

#null values in each column
data.isnull().sum()

Observation 0
Y-Kanna 0

```
ChipRate
BF-CMratio
                 17
BlowFlow
                 16
ChipLevel4
T-upperExt-2
                  2
T-lowerExt-2
UCZAA
                  25
WhiteFlow-4
AAWhiteSt-4
                 151
AA-Wood-4
ChipMoisture-4
                  1
SteamFlow-4
                  1
Lower-HeatT-3
Upper-HeatT-3
ChipMass-4
WeakLiquorF
                  1
BlackFlow-2
WeakWashF
SteamHeatF-3
T-Top-Chips-4
SulphidityL-4
                 151
dtype: int64
```

#total number of null values in the dataset
data.isnull().sum().sum()

386

#check for non null values
data.notnull()

	Observation	Y- Kappa	ChipRate	BF- CMratio	BlowFlow	ChipLevel4	T- upperExt- 2	T lowerExt
0	True	True	True	True	True	True	True	Tru
1	True	True	True	True	True	True	True	Tru
2	True	True	True	True	True	True	True	Tru
3	True	True	True	True	True	True	True	Tru
4	True	True	True	True	True	True	True	Tru
319	True	True	True	True	True	True	True	Tru
320	True	True	True	True	True	True	True	Tru
321	True	True	True	True	True	True	True	Tru
322	True	True	True	True	True	True	True	Tru
323	True	True	True	True	True	True	True	Tru
324 rc	ows × 23 column	ıs						

#non null values in each column
data.notnull().sum()

Observation	324
Ү-Карра	324
ChipRate	319
BF-CMratio	307
BlowFlow	308
ChipLevel4	323
T-upperExt-2	322
T-lowerExt-2	322
UCZAA	299
WhiteFlow-4	323
AAWhiteSt-4	173
AA-Wood-4	323
ChipMoisture-4	323
SteamFlow-4	323
Lower-HeatT-3	322
Upper-HeatT-3	322
ChipMass-4	323
WeakLiquorF	323
BlackFlow-2	322
WeakWashF	323
SteamHeatF-3	322
T-Top-Chips-4	323
SulphidityL-4	173
dtype: int64	

#replacing null values with 0
d1=data.fillna(value=0)

	Observation	Y- Kappa	ChipRate	BF- CMratio	BlowFlow	ChipLevel4	T- upperExt- 2	T- lowerExt- 2	UCZAA	WhiteFlow- 4	•••	SteamFlow- 4	Lower- HeatT- 3
0	31-00:00	23.10	16.520	121.717	1177.607	169.805	358.282	329.545	1.443	599.253		67.122	329.432
1	31-01:00	27.60	16.810	79.022	1328.360	341.327	351.050	329.067	1.549	537.201		60.012	330.823
2	31-02:00	23.19	16.709	79.562	1329.407	239.161	350.022	329.260	1.600	549.611		61.304	329.140
3	31-03:00	23.60	16.478	81.011	1334.877	213.527	350.938	331.142	1.604	623.362		68.496	328.875
4	31-04:00	22.90	15.618	93.244	1334.168	243.131	351.640	332.709	0.000	638.672		70.022	328.352
319	10-16:00	23.75	12.667	93.450	1178.252	276.955	347.286	310.970	1.523	513.956		61.141	330.117
320	9-19:00	19.80	12.558	94.352	1184.119	297.071	399.135	319.576	1.451	570.058		67.667	330.848
321	9-20:00	23.01	12.550	90.842	1188.517	289.826	373.633	314.591	1.457	549.306		66.446	330.226
322	9-21:00	24.32	13.083	88.910	1192.879	318.006	364.081	308.559	1.523	504.852		61.054	327.346
323	9-22:00	25.75	13.417	85.451	1186.342	248.312	356.289	310.482	1.474	497.375		58.247	328.092
324 rc	ows × 23 column	ıs											

[#] Forward fill null values
d2=data.fillna(method="pad")
d2

	Observation	Y- Kappa	ChipRate	BF- CMratio	BlowFlow	ChipLevel4	T- upperExt- 2	T- lowerExt- 2	UCZAA	WhiteFlow-	•••	SteamFlow-	Lower- HeatT- 3
0	31-00:00	23.10	16.520	121.717	1177.607	169.805	358.282	329.545	1.443	599.253		67.122	329.432
1	31-01:00	27.60	16.810	79.022	1328.360	341.327	351.050	329.067	1.549	537.201		60.012	330.823
2	31-02:00	23.19	16.709	79.562	1329.407	239.161	350.022	329.260	1.600	549.611		61.304	329.140
3	31-03:00	23.60	16.478	81.011	1334.877	213.527	350.938	331.142	1.604	623.362		68.496	328.875
4	31-04:00	22.90	15.618	93.244	1334.168	243.131	351.640	332.709	1.604	638.672		70.022	328.352
								•••					
319	10-16:00	23.75	12.667	93.450	1178.252	276.955	347.286	310.970	1.523	513.956		61.141	330.117
320	9-19:00	19.80	12.558	94.352	1184.119	297.071	399.135	319.576	1.451	570.058		67.667	330.848
321	9-20:00	23.01	12.550	90.842	1188.517	289.826	373.633	314.591	1.457	549.306		66.446	330.226
322	9-21:00	24.32	13.083	88.910	1192.879	318.006	364.081	308.559	1.523	504.852		61.054	327.346
323	9-22:00	25.75	13.417	85.451	1186.342	248.312	356.289	310.482	1.474	497.375		58.247	328.092
324 rc	ows × 23 column	s											

[#] Backward fill null values
d3=data.fillna(method="bfill")
d3

	Observation	Y- Kappa	ChipRate	BF- CMratio	BlowFlow	ChipLevel4	T- upperExt- 2	T- lowerExt- 2	UCZAA	WhiteFlow- 4	•••	SteamFlow-	Lower- HeatT- 3
0	31-00:00	23.10	16.520	121.717	1177.607	169.805	358.282	329.545	1.443	599.253		67.122	329.432
1	31-01:00	27.60	16.810	79.022	1328.360	341.327	351.050	329.067	1.549	537.201		60.012	330.823
2	31-02:00	23.19	16.709	79.562	1329.407	239.161	350.022	329.260	1.600	549.611		61.304	329.140
3	31-03:00	23.60	16.478	81.011	1334.877	213.527	350.938	331.142	1.604	623.362		68.496	328.875
4	31-04:00	22.90	15.618	93.244	1334.168	243.131	351.640	332.709	1.436	638.672		70.022	328.352

319	10-16:00	23.75	12.667	93.450	1178.252	276.955	347.286	310.970	1.523	513.956		61.141	330.117
320	9-19:00	19.80	12.558	94.352	1184.119	297.071	399.135	319.576	1.451	570.058		67.667	330.848
321	9-20:00	23.01	12.550	90.842	1188.517	289.826	373.633	314.591	1.457	549.306		66.446	330.226
322	9-21:00	24.32	13.083	88.910	1192.879	318.006	364.081	308.559	1.523	504.852		61.054	327.346
323	9-22:00	25.75	13.417	85.451	1186.342	248.312	356.289	310.482	1.474	497.375		58.247	328.092
324 rc	ws × 23 column	ıs											

#selecting numeric columns
d11=data.select_dtypes(include="number")
d11

	Y- Kappa	ChipRate	BF- CMratio	BlowFlow	ChipLevel4	T- upperExt- 2	T- lowerExt- 2	UCZAA	WhiteFlow-	AAWhiteSt- 4	 SteamFlow-	Lower- HeatT- 3	l I
0	23.10	16.520	121.717	1177.607	169.805	358.282	329.545	1.443	599.253	NaN	 67.122	329.432	3
1	27.60	16.810	79.022	1328.360	341.327	351.050	329.067	1.549	537.201	6.076	 60.012	330.823	3
2	23.19	16.709	79.562	1329.407	239.161	350.022	329.260	1.600	549.611	NaN	 61.304	329.140	3
3	23.60	16.478	81.011	1334.877	213.527	350.938	331.142	1.604	623.362	6.054	 68.496	328.875	3
4	22.90	15.618	93.244	1334.168	243.131	351.640	332.709	NaN	638.672	6.110	 70.022	328.352	3
					***	***				***	 		
319	23.75	12.667	93.450	1178.252	276.955	347.286	310.970	1.523	513.956	6.068	 61.141	330.117	3
320	19.80	12.558	94.352	1184.119	297.071	399.135	319.576	1.451	570.058	6.190	 67.667	330.848	3
321	23.01	12.550	90.842	1188.517	289.826	373.633	314.591	1.457	549.306	NaN	 66.446	330.226	3
322	24.32	13.083	88.910	1192.879	318.006	364.081	308.559	1.523	504.852	6.128	 61.054	327.346	3
323	25.75	13.417	85.451	1186.342	248.312	356.289	310.482	1.474	497.375	NaN	 58.247	328.092	3
324 rc	ws × 22	columns											

[#] Replace null values with the mean of the column d4=data.fillna(d11.mean()) d4

	Observation	Y- Kappa	ChipRate	BF- CMratio	BlowFlow	ChipLevel4	T- upperExt- 2	T- lowerExt- 2	UCZAA	WhiteFlow-	•••	SteamFlow-	Lower- HeatT- 3
0	31-00:00	23.10	16.520	121.717	1177.607	169.805	358.282	329.545	1.44300	599.253		67.122	329.432
1	31-01:00	27.60	16.810	79.022	1328.360	341.327	351.050	329.067	1.54900	537.201		60.012	330.823
2	31-02:00	23.19	16.709	79.562	1329.407	239.161	350.022	329.260	1.60000	549.611		61.304	329.140
3	31-03:00	23.60	16.478	81.011	1334.877	213.527	350.938	331.142	1.60400	623.362		68.496	328.875
4	31-04:00	22.90	15.618	93.244	1334.168	243.131	351.640	332.709	1.49201	638.672		70.022	328.352
													•••
319	10-16:00	23.75	12.667	93.450	1178.252	276.955	347.286	310.970	1.52300	513.956		61.141	330.117
320	9-19:00	19.80	12.558	94.352	1184.119	297.071	399.135	319.576	1.45100	570.058		67.667	330.848
321	9-20:00	23.01	12.550	90.842	1188.517	289.826	373.633	314.591	1.45700	549.306		66.446	330.226
322	9-21:00	24.32	13.083	88.910	1192.879	318.006	364.081	308.559	1.52300	504.852		61.054	327.346
323	9-22:00	25.75	13.417	85.451	1186.342	248.312	356.289	310.482	1.47400	497.375		58.247	328.092
324 rc	ows × 23 column	ıs											

[#] Remove rows with any null values
d5=data.dropna()

	Observation	Y- Kappa	ChipRate	BF- CMratio	BlowFlow	ChipLevel4	T- upperExt- 2	T- lowerExt- 2	UCZAA	WhiteFlow-	 SteamFlow-	Lower- HeatT- 3
1	31-01:00	27.60	16.810	79.022	1328.360	341.327	351.050	329.067	1.549	537.201	 60.012	330.823
3	31-03:00	23.60	16.478	81.011	1334.877	213.527	350.938	331.142	1.604	623.362	 68.496	328.875
5	1-08:00	14.23	15.350	85.518	1171.604	198.538	344.014	325.195	1.436	628.245	 65.225	322.103
7	31-06:00	22.65	14.100	91.887	1307.852	288.989	352.321	331.162	1.468	625.549	 71.298	329.662
9	31-08:00	24.70	13.850	96.208	1334.892	362.511	352.372	327.358	1.515	553.172	 64.249	332.264
312	31-10:00	24.40	14.117	85.998	1330.104	394.234	348.089	319.027	1.429	540.558	 62.179	329.831
317	4-16:00	17.80	16.625	78.367	1276.082	202.744	360.127	329.266	1.488	698.486	 75.296	321.658
319	10-16:00	23.75	12.667	93.450	1178.252	276.955	347.286	310.970	1.523	513.956	 61.141	330.117
320	9-19:00	19.80	12.558	94.352	1184.119	297.071	399.135	319.576	1.451	570.058	 67.667	330.848
322	9-21:00	24.32	13.083	88.910	1192.879	318.006	364.081	308.559	1.523	504.852	 61.054	327.346
141 rows × 23 columns												

d1.isnull().sum()

Observation	0
Y-Kappa	0
ChipRate	0
BF-CMratio	0
BlowFlow	0
ChipLevel4	0
T-upperExt-2	0
T-lowerExt-2	0
UCZAA	0
WhiteFlow-4	0
AAWhiteSt-4	0
AA-Wood-4	0
ChipMoisture-4	0
SteamFlow-4	0
Lower-HeatT-3	0
Upper-HeatT-3	0
ChipMass-4	0
WeakLiquorF	0
BlackFlow-2	0
WeakWashF	0
SteamHeatF-3	0
T-Top-Chips-4	0
SulphidityL-4	0
dtype: int64	

```
4/26/24, 7:44 PM
   #removing observation column
   d1.drop(['Observation'],axis=1,inplace=True)
   d1.columns
       'BlackFlow-2', 'We 'SulphidityL-4'],
             dtype='object')
   import numpy as np
   # Calculate Q1, Q3, and IQR
   q1=d1.quantile(0.25)
   q3=d1.quantile(0.75)
   iqr=q3-q1
   print(iqr)
                        4.65000
        Y-Kappa
        ChipRate
                         2.25625
                    11.11225
98.43375
        BF-CMratio
        BlowFlow
       ChipLevel4
                      107.92275
                       11.96500
        T-upperExt-2
        T-lowerExt-2
                         7.82875
       UCZAA
                         0.13925
       WhiteFlow-4
                        98.59525
                        6.14000
        AAWhiteSt-4
        AA-Wood-4
                         1.45900
                        2.22000
9.04675
        ChipMoisture-4
        SteamFlow-4
                        8.46750
7.77050
        Lower-HeatT-3
        Upper-HeatT-3
                        19.70375
       ChipMass-4
                     174.05550
276.51675
        WeakLiquorF
       BlackFlow-2
       WeakWashF
                       271.44325
       SteamHeatF-3
                         6.94975
        T-Top-Chips-4
                         2.01025
        SulphidityL-4
                         30.40250
        dtype: float64
   # Define the boundaries for outliers
   lowerbound=q1-1.5*iqr
   upperbound=q3+1.5*iqr
   #filter outliers
```

 $outlier=d1[\sim((d1<lowerbound)|(d1>upperbound)).any(axis=1)]$ outlier

	Y- Kappa	ChipRate	BF- CMratio	BlowFlow	ChipLevel4	T- upperExt- 2	T- lowerExt- 2	UCZAA	WhiteFlow- 4	AAWhiteSt- 4		SteamFlow-	Lower- HeatT- 3	l I
1	27.60	16.810	79.022	1328.360	341.327	351.050	329.067	1.549	537.201	6.076		60.012	330.823	3
2	23.19	16.709	79.562	1329.407	239.161	350.022	329.260	1.600	549.611	0.000		61.304	329.140	3
3	23.60	16.478	81.011	1334.877	213.527	350.938	331.142	1.604	623.362	6.054		68.496	328.875	3
5	14.23	15.350	85.518	1171.604	198.538	344.014	325.195	1.436	628.245	6.020		65.225	322.103	2
6	13.49	13.700	98.186	1243.688	116.275	346.208	326.982	1.434	696.766	0.000		72.989	322.982	2
317	17.80	16.625	78.367	1276.082	202.744	360.127	329.266	1.488	698.486	6.126		75.296	321.658	2
318	18.20	16.283	83.508	1288.104	234.284	359.412	328.670	1.534	692.687	0.000		74.528	321.224	2
319	23.75	12.667	93.450	1178.252	276.955	347.286	310.970	1.523	513.956	6.068		61.141	330.117	3
321	23.01	12.550	90.842	1188.517	289.826	373.633	314.591	1.457	549.306	0.000		66.446	330.226	3
323	25.75	13.417	85.451	1186.342	248.312	356.289	310.482	1.474	497.375	0.000		58.247	328.092	3
241 rows × 22 columns														

#descriptive statistics outlier.describe()