THIS TASK INVOLVES USING THE PANDAS LIBRARY TO MANIPULATE DATA AND TO PERFORM OPERATIONS LIKE FILTERING DATA BASED ON CONDITIONS, HANDLING MISSING VALUES, AND CALCULATING SUMMARY STATISTICS.

· Reading CSV File

Importing pandas to perform data manipulation and analysis tasks import pandas as pd

Reading the CSV file
data = pd.read_csv('/content/01.Data Cleaning and Preprocessing.csv')

• Initial Exploration:

Displaying the first few rows of the DataFrame
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 rows × 23 columns

Displaying the shape of the DataFrame:
data.shape

(324, 23)

Displaying information about the DataFrame:
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

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14 Lower-HeatT-3
                        322 non-null
                                       float64
                        322 non-null
                                       float64
     15 Upper-HeatT-3
     16 ChipMass-4
17 WeakLiquorF
                                       float64
                        323 non-null
                        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
    dtypes: float64(22), object(1)
    memory usage: 58.3+ KB
# Displaying data types of columns:
data.dtypes
    Observation 0
                      object
                      float64
    Y-Kappa
    ChipRate
                     float64
    BF-CMratio
                     float64
    BlowFlow
                     float64
    ChipLevel4
                     float64
    T-upperExt-2
                     float64
    T-lowerExt-2
                     float64
    UCZAA
                     float64
                     float64
    WhiteFlow-4
                     float64
    AAWhiteSt-4
    AA-Wood-4
                     float64
    ChipMoisture-4
                     float64
    SteamFlow-4
                     float64
    Lower-HeatT-3
                     float64
    Upper-HeatT-3
                     float64
    ChipMass-4
                     float64
    WeakLiquorF
                     float64
    BlackFlow-2
                     float64
                     float64
    WeakWashF
    SteamHeatF-3
                     float64
    T-Top-Chips-4
                     float64
                     float64
    SulphidityL-4
    dtype: object
# Displaying column names:
data.columns
    dtype='object')
```

Displaying summary statistics of the DataFrame:

data.describe()

	Y-Kappa	ChipRate	BF-CMratio	BlowFlow	ChipLevel4	T-upperExt-2	T-lowerExt-2	U
count	324.000000	319.000000	307.000000	308.000000	323.000000	322.000000	322.000000	299.00
mean	20.635370	14.347937	87.464456	1237.837614	258.164483	356.904295	324.020180	1.49
std	3.070036	1.499095	7.995012	100.593735	87.987452	9.209290	7.621402	0.10
min	12.170000	9.983000	68.645000	0.000000	0.000000	339.168000	284.633000	1.18
25%	18.382500	13.358000	81.823000	1193.215250	213.527000	350.241250	321.420000	1.43
50%	20.845000	14.308000	86.739000	1273.138500	271.792000	356.843000	325.669000	1.49
75%	23.032500	15.517000	92.372000	1289.196000	321.680000	362.242250	329.175000	1.5€
max	27.600000	16.958000	121.717000	1351.240000	419.014000	399.135000	337.012000	1.74

• Handling Duplicate Rows:

Checking for duplicated rows:
data.duplicated()

0	False
1	False
2	False
3	False
4	False
319	True
320	True
321	True
322	True
323	True

Length: 324, dtype: bool

[#] Sorting the DataFrame by a specific column:
data.sort_values(by='BF-CMratio', ascending=True)

	Observation	Y-Kappa	ChipRate	BF-CMratio	BlowFlow	ChipLevel4	T-upperExt-2	T-lowerExt-2
31	1-06:00	16.10	16.292	68.645	1094.883	245.962	344.478	321.061
112	4-15:00	18.80	16.455	71.679	1191.663	195.725	359.473	326.503
30	1-05:00	24.15	14.533	72.015	1173.241	267.057	347.734	317.653
63	2-14:00	24.70	16.183	72.611	1231.362	332.497	350.938	322.801
15	31-14:00	25.40	16.425	72.924	1197.775	118.821	350.765	329.799
•••								
207	8-14:00	16.05	13.114	NaN	NaN	0.000	NaN	NaN
208	8-15:00	17.75	11.283	NaN	NaN	41.607	362.545	331.240
314	8-13:00	16.70	NaN	NaN	NaN	0.000	373.867	324.184
315	8-14:00	16.05	13.114	NaN	NaN	0.000	NaN	NaN
316	8-15:00	17.75	11.283	NaN	NaN	41.607	362.545	331.240

Counting 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	

Checking for null values in the DataFrame:
data.isnull()

	Observation	Y-Kappa	ChipRate	BF-CMratio	BlowFlow	ChipLevel4	T-upperExt-2	T-lowerExt-2
0	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False
•••								•••
319	False	False	False	False	False	False	False	False
320	False	False	False	False	False	False	False	False
321	False	False	False	False	False	False	False	False
322	False	False	False	False	False	False	False	False
323	False	False	False	False	False	False	False	False

Counting null values in each column:
data.isnull().sum()

Observation	0
Y-Kappa	0
ChipRate	5
BF-CMratio	17
BlowFlow	16
ChipLevel4	1
T-upperExt-2	2
T-lowerExt-2	2
UCZAA	25
WhiteFlow-4	1
AAWhiteSt-4	151
AA-Wood-4	1
ChipMoisture-4	1
SteamFlow-4	1
Lower-HeatT-3	2
Upper-HeatT-3	2
ChipMass-4	1
WeakLiquorF	1
BlackFlow-2	2
WeakWashF	1
SteamHeatF-3	2
T-Top-Chips-4	1
SulphidityL-4	151
dtype: int64	

Counting total null values in the DataFrame:
data.isnull().sum().sum()

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• Handling Missing Values

```
# Filling null values with 0:
data1 = data.fillna(value=0)
data1
```

	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
•••								•••
319	10-16:00	23.75	12.667	93.450	1178.252	276.955	347.286	310.970
320	9-19:00	19.80	12.558	94.352	1184.119	297.071	399.135	319.576
321	9-20:00	23.01	12.550	90.842	1188.517	289.826	373.633	314.591
322	9-21:00	24.32	13.083	88.910	1192.879	318.006	364.081	308.559
323	9-22:00	25.75	13.417	85.451	1186.342	248.312	356.289	310.482

Counting null values in each column after filling null values with 0
data1.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	

Removing duplicate rows from the DataFrame:
data2 = data.drop_duplicates()
data2

	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
•••								
298	12-09:00	20.90	15.167	84.640	1283.706	339.440	354.803	311.041
299	12-10:00	24.98	NaN	85.034	1278.345	368.564	357.723	321.387
300	12-11:00	21.00	NaN	88.013	1307.722	278.842	357.438	323.757
301	12-12:00	21.40	NaN	85.490	1255.986	273.484	361.365	322.689
307	31-05:00	20.89	14.308	94.172	1327.832	251.120	351.263	332.485

Remove non-numeric columns
numeric_data = data.select_dtypes(include=['number'])

Fill null values with the mean of each column
data3 = numeric_data.fillna(numeric_data.mean())
data3

	Y-Kappa	ChipRate	BF-CMratio	BlowFlow	ChipLevel4	T-upperExt-2	T-lowerExt-2	UCZAA	Whi
0	23.10	16.520	121.717	1177.607	169.805	358.282	329.545	1.44300	
1	27.60	16.810	79.022	1328.360	341.327	351.050	329.067	1.54900	
2	23.19	16.709	79.562	1329.407	239.161	350.022	329.260	1.60000	
3	23.60	16.478	81.011	1334.877	213.527	350.938	331.142	1.60400	
4	22.90	15.618	93.244	1334.168	243.131	351.640	332.709	1.49201	
•••									
319	23.75	12.667	93.450	1178.252	276.955	347.286	310.970	1.52300	
320	19.80	12.558	94.352	1184.119	297.071	399.135	319.576	1.45100	
321	23.01	12.550	90.842	1188.517	289.826	373.633	314.591	1.45700	
322	24.32	13.083	88.910	1192.879	318.006	364.081	308.559	1.52300	
323	25.75	13.417	85.451	1186.342	248.312	356.289	310.482	1.47400	

324 rows × 22 columns

Filling null values using backward fill (bfill):
data4 = data.fillna(method = 'bfill')
data4

	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
•••								•••
319	10-16:00	23.75	12.667	93.450	1178.252	276.955	347.286	310.970
320	9-19:00	19.80	12.558	94.352	1184.119	297.071	399.135	319.576
321	9-20:00	23.01	12.550	90.842	1188.517	289.826	373.633	314.591
322	9-21:00	24.32	13.083	88.910	1192.879	318.006	364.081	308.559
323	9-22:00	25.75	13.417	85.451	1186.342	248.312	356.289	310.482

Filling null values using forward fill (pad):
data5 = data.fillna(method = 'pad')
data5

	Observation	Ү-Карра	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
•••							•••	
319	10-16:00	23.75	12.667	93.450	1178.252	276.955	347.286	310.970
320	9-19:00	19.80	12.558	94.352	1184.119	297.071	399.135	319.576
321	9-20:00	23.01	12.550	90.842	1188.517	289.826	373.633	314.591
322	9-21:00	24.32	13.083	88.910	1192.879	318.006	364.081	308.559
323	9-22:00	25.75	13.417	85.451	1186.342	248.312	356.289	310.482

324 rows × 23 columns

• Outlier Detection and Removal:

import numpy as np

data1.columns

```
'WhiteFlow-4', 'AAWhiteSt-4', 'AA-Wood-4', 'ChipMoisture-4', 'SteamFlow-4', 'Lower-HeatT-3', 'Upper-HeatT-3', 'ChipMass-4', 'WeakLiquorF', 'BlackFlow-2', 'WeakWashF', 'SteamHeatF-3', 'T-Top-Chips-4', 'SulphidityL-4'],
                 dtype='object')
# Dropping the 'Observation' column:
data1.drop(['Observation'],axis = 1, inplace = True)
data1.columns
       'AAWhiteSt-4', 'AA-Wood-4', 'ChipMoisture-4', 'SteamFlow-4', 'Lower-HeatT-3', 'Upper-HeatT-3', 'ChipMass-4', 'WeakLiquorF', 'BlackFlow-2', 'WeakWashF', 'SteamHeatF-3', 'T-Top-Chips-4', 'SulphidityL-4'],
                 dtype='object')
# Calculate the first quartile (Q1)
Quartile1 = data1.quantile(0.25)
# Calculate the third quartile (Q3)
Quartile3 = data1.quantile(0.75)
# Calculate the interquartile range (IQR)
iqr = Quartile3 - Quartile1
print(iqr)
       Y-Kappa 4.65000
ChipRate 2.25625
BF-CMratio 11.11225
BlowFlow 98.43375
ChipLevel4 107.92275
T-upperExt-2 11.96500
T-lowerExt-2 7.82875
UC7AA 0.12025
                                      7.82875
0.13925
       UCZAA
       WhiteFlow-4 98.59525
AAWhiteSt-4 6.14000
AA-Wood-4 1.45900
```

AA-Wood-4 1.45900 ChipMoisture-4 2.22000 SteamFlow-4 9.04675 Lower-HeatT-3 8.46750 Upper-HeatT-3 7.77050 ChipMass-4 19.70375 WeakLiquorF 174.05550 BlackFlow-2 276.51675 WeakWashF 271.44325

271.44325

WeakWashF