TC2-DS- Experiment 1

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AIM - Write a python program to open Comma Separated Value (CSV) and perform given statistical operations.

• THEORY:

What is CSV?

A CSV file, often known as a comma separated values file, enables the saving of data in plain text in a table-structured format. Every line in the file represents a data record, and each record is made up of one or more fields that are separated by commas. CSVs resemble spreadsheets but have a ".csv" extension.

They differ from other spreadsheet file types in that you can only have a single sheet in a file, that cell, column, or row formatting cannot be saved, and that formulae cannot be saved.

Data is the value you record in your data sheet and is a specific measurement of a variable.

Data typically falls into one of two categories:

- 1. Categorical data represents groupings/classifications.
- 2. **Numerical/Quantitative data** represents quantities/amounts.

Variables with quantitative data are considered to be **quantitative variables**, while variables with categorical data are considered to be **categorical variables**. Each of these variable kinds can be further subdivided into other types.

Categorical Variables :

It represent groupings of some kind. They are sometimes recorded as numbers, but the numbers represent categories rather than actual amounts of things.

There are three types of categorical variables: binary, nominal, and ordinal variables.

- Binary Variable: A binary variable is a categorical variable that can only
 have one of two possible values. These variables are typically expressed as
 Boolean variables (True or False) or integer variables (0 or 1). Example:
 Heads/tails in a coin flip
- **Nominal Variable**: A nominal variable is a categorical variable that groups with no rank or order between them. Example: Colors.
- **Ordinal Variable**: An ordinal variable is a categorical variable that ranks groups in a particular order. Example: Finishing place in a race.

• Numerical Variables:

The numbers you record when collecting quantitative data represent real amounts that can be added, subtracted, divided, and so on.

There are two types of numerical variables: discrete, continuous variables.

- Discrete Variable: A variable is considered discrete if the data it represents is a count of individual items or values. Example: Number of students in a class.
- **Continuous Variable**: A variable is considered continuous if the data it represents is continuous or non-finite. Example: Age, Distance.

```
The Dataset is as Follows:
     Id SepalLengthCm SepalWidthCm PetalLengthCm PetalWidthCm \
      1
                 5.1
                             3.5
                                           1.4
1
     2
                 4.9
                             3.0
                                           1.4
                                                       0.2
2
     3
                 4.7
                            3.2
                                                       0.2
                                           1.3
3
     4
                 4.6
                             3.1
                                           1.5
                                                       0.2
4
     5
                5.0
                             3.6
                                          1.4
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145 146
                 6.7
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                                           5.2
                                                       2.3
                                                       1.9
146 147
                             2.5
                6.3
                                          5.0
147 148
                6.5
                            3.0
                                          5.2
                                                       2.0
148 149
                 6.2
                             3.4
                                           5.4
                                                       2.3
149 150
                 5.9
                             3.0
                                           5.1
                                                       1.8
          Species
      Iris-setosa
      Iris-setosa
2
      Iris-setosa
      Iris-setosa
4
     Iris-setosa
145 Iris-virginica
146 Iris-virginica
147 Iris-virginica
148 Iris-virginica
149 Iris-virginica
[150 rows x 6 columns]
```

- Classifications of data:

Number of categorical variables: 1

Index(['Species'], dtype='object')

```
Categorical_columns = Dataset_df.dtypes[Dataset_df.dtypes == 'object'].index
print('Number of categorical variables:', len(Categorical_columns), '\n')
print(Categorical_columns)
```

```
Numerical_columns = Dataset_df.dtypes[Dataset_df.dtypes != 'object'].index
print('Number of numerical variables:', len(Numerical_columns), '\n')
print(Numerical_columns)

Number of numerical variables: 5

Index(['Id', 'SepalLengthCm', 'SepalWidthCm', 'PetalLengthCm', 'PetalWidthCm'], dtyp
```

Contingency of the table:

```
print("Contingency of the table:")
print(pd.crosstab(Dataset_df.Species, Dataset_df.SepalLengthCm, margins=True), '\n')
     Contingency of the table:
                     4.3 4.4 4.5 4.6 4.7 4.8 4.9 5.0 5.1 5.2
     SepalLengthCm
                                                                             6.9 \
     Species
     Iris-setosa
                        1
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                                  1
                                       4
                                            2
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     Iris-versicolor
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                                       0
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                                                           2
                        0
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                                                                     1
                                                                               1
     Iris-virginica
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                                       0
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                                                      1
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                                                                               3
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    All
                                       4
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    SepalLengthCm
                     7.0 7.1 7.2 7.3 7.4 7.6 7.7 7.9
     Species
     Iris-setosa
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                                            0
                                                 0
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                                                           0
                                                               50
     Iris-versicolor
                        1
                             0
                                  0
                                       0
                                            0
                                                 0
                                                      0
                                                           0
                                                               50
     Iris-virginica
                             1
                                            1
                                                 1
                                                      4
                                                               50
    All
                                  3
                                                     4
                                                           1 150
```

[4 rows x 36 columns]

Statistical Operations/Calculations:

1. Mean:

```
# Function that Calculates Mean:
def mean(n):
    return sum(n) / len(n)

# Function that Calculates Median:
def median(n):
    length = len(n)
    if length % 2 == 0:
        median1 = n[length // 2]
        median2 = n[length // 2 - 1]
        median = (median1 + median2) / 2
    else:
```

```
# Function that Calculates Mode:
def mode(n):
   length = len(n)
   data = collections.Counter(n)
   d_list = dict(data)
   max_value = max(list(data.values()))
   mode_val = [num for num, freq in d_list.items() if freq == max_value]
   if len(mode val) == length:
        return
   else:
        return ', '.join(map(str, mode_val))
# Function that Calculates Variance:
def variance(n):
   ans = sum((x - mean(n)) ** 2 for x in n) / len(n)
    return ans
# Function that Calculates Standard Deviation:
def standard deviation(n):
   Stan Deviation = variance(n) ** 0.5
    return Stan_Deviation
# Function that Calculates Quartile Range:
def inter_quartile_range(arr):
    arr = sorted(arr, reverse=False)
    if len(arr) % 2 != 0:
        mid_i = int(len(arr) / 2)
        arr_1 = arr[:mid_i]
        arr 2 = arr[mid i + 1:]
        arr_1_mid = arr_1[int(len(arr_1) / 2)]
        arr 2 mid = arr 2[int(len(arr 2) / 2)]
        return arr_2_mid - arr_1_mid
    else:
        first_i = int(len(arr) / 2)
        second i = int(len(arr) / 2)
        arr_1 = arr[:first_i]
        arr_1_mid_1 = arr_1[int(len(arr_1) / 2) - 1]
        arr_1_mid_2 = arr_1[int(len(arr_1) / 2)]
        arr_1_mid = sum([arr_1_mid_1, arr_1_mid_2]) / 2
        arr_2 = arr[second_i:]
        arr 2 mid 1 = arr 2[int(len(arr 2) / 2) - 1]
        arr 2 mid 2 = arr 2[int(len(arr 2) / 2)]
        arr_2_mid = sum([arr_2_mid_1, arr_2_mid_2]) / 2
        return arr_2_mid - arr_1_mid
print("Mean is: ", mean(Dataset_df["SepalLengthCm"]))
print("Median is: ", median(Dataset_df["SepalLengthCm"]))
```

median = n[length // 2]

return median

```
print("Mode is: ", mode(Dataset_df["SepalLengthCm"]))
print("Variance is: ", variance(Dataset_df["SepalLengthCm"]))
print("Standard Deviation is: ", standard_deviation(Dataset_df["SepalLengthCm"]))
print("Quartile Range is: ", inter_quartile_range(Dataset_df["SepalLengthCm"]))

Mean is: 5.84333333333335
Median is: 6.5
Mode is: 5.0
Variance is: 0.6811222222222222
Standard Deviation is: 0.8253012917851409
Quartile Range is: 1.3000000000000007
```

Determination of type of categorical variable:

```
#Function that Calculates Nominal:
def Nominal(n):
    if len(set(n)) > len(n) // 5:
        return
    elif len(set(n)) == len(n):
        print("The Data is Nominal.")
    elif len(set(n)) <= 10:</pre>
        print("The Data is Nominal.")
def binary(n):
    if len(set(n)) == 2:
        print("The Data is Binary. ")
    else:
        return
def Ordinal(n):
    if len(set(n)) >= len(n) // 5:
        print("The Data is Ordinal.")
    else:
        return
def Category(n):
    binary(n)
    Nominal(n)
    Ordinal(n)
Category(Dataset_df["Species"])
     The Data is Nominal.
```

Converting to Nominal data

Converting to Ordinal data

```
print("Ordinal Data is given as: ")
data_ordinal = Dataset_df.copy()
# define ordinal encoding
encoder2 = OrdinalEncoder()
# transform data
data_ordinal = encoder2.fit_transform(data_ordinal)
print(data_ordinal)
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                                   1.]
      [ 99.
             14.
                   7. 17.
                             9.
                                   1.]
      [100.
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                                   2.]
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      [101.
             15.
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      [102.
             28.
                   9. 35.
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      [103.
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```

```
- Graphs:
```

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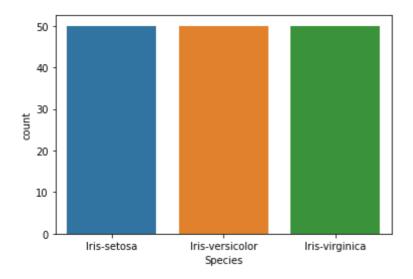
2.]

2.]

2.]]

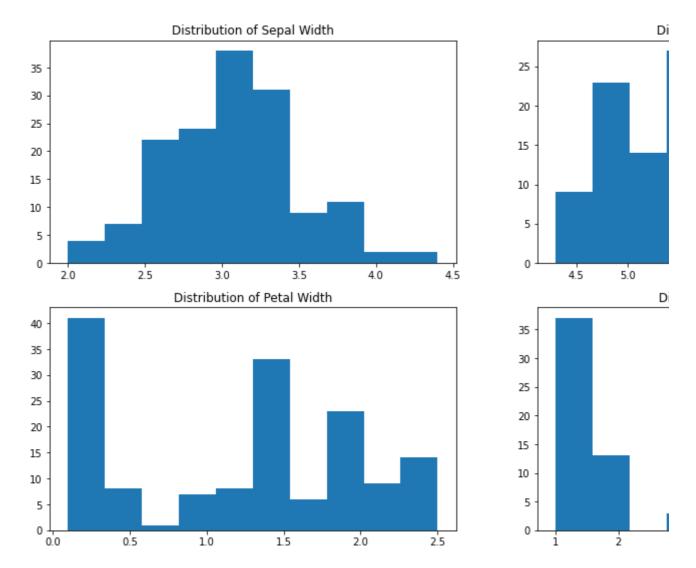
1. Count of Species

```
# Graph for Species Count
sns.countplot(x='Species', data=Dataset_df)
plt.show()
```



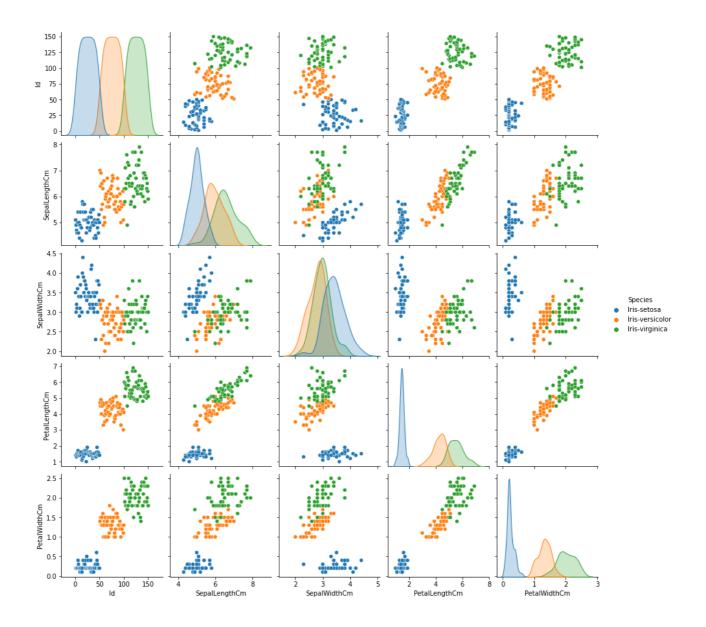
2. Histogram for Sepal Length & Width, Petal Length & Width

```
fig, axes = plt.subplots(2, 2, figsize=(16, 9))
axes[0, 0].set_title("Distribution of Sepal Width")
axes[0, 0].hist(Dataset_df['SepalWidthCm'], bins=10)
axes[0, 1].set_title("Distribution of Sepal Length")
axes[0, 1].hist(Dataset_df['SepalLengthCm'], bins=10)
axes[1, 0].set_title("Distribution of Petal Width")
axes[1, 0].hist(Dataset_df['PetalWidthCm'], bins=10)
axes[1, 1].set_title("Distribution of Petal Length")
axes[1, 1].hist(Dataset_df['PetalLengthCm'], bins=10)
plt.show()
```



3. Graph displaying a multivariate set of observations such as species and sepal and petal lengths and widths

```
sns.pairplot(Dataset_df, hue='Species')
plt.show()
```



Conclusion:

Hence, we performed operations on the given dataset, and executed data type classification, contingency, and statistical functions.

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