# Data\_Analysis

#### March 15, 2022

#### Data Fields and Types

```
attrname attrtype factororresponse
O SepalLength numeric
                                 factor
  SepalWidth numeric
                                 factor
2 PetalLength numeric
                                 factor
3
  PetalWidth numeric
                                 factor
4
    IrisClass nominal
                               response
attribute_datas None
Getting a Response Column
response_attributes: IrisClass
Getting a Non-Factor Column
non_factor_attributesname:
                           []
Getting a Factor Column
factor_attributesname: ['SepalLength' 'SepalWidth' 'PetalLength' 'PetalWidth']
Getting a Nominal Type Column
nominal_attributes: ['IrisClass']
Getting a Numeric Type Column
numeric_attributes: ['SepalLength' 'SepalWidth' 'PetalLength' 'PetalWidth']
Getting a Binary Type Column
binary_attributes: []
Sample Data
  SepalLength SepalWidth PetalLength PetalWidth
                                                      IrisClass
0
          5.1
                                   1.4
                      3.5
                                               0.2 Iris-setosa
          4.9
                      3.0
1
                                   1.4
                                               0.2 Iris-setosa
2
          4.7
                      3.2
                                   1.3
                                               0.2 Iris-setosa
```

data head None

4.6

5.0

3.1

3.6

Data Columns

3

1.5

1.4

0.2 Iris-setosa

0.2 Iris-setosa

```
data columns: Index(['SepalLength', 'SepalWidth', 'PetalLength', 'PetalWidth',
'IrisClass'], dtype='object')
```

#### Summary Statistics

Data Describe

	count	mean	std	min	25%	50%	75%	max
SepalLength	150.0	5.843333	0.828066	4.3	5.1	5.80	6.4	7.9
SepalWidth	150.0	3.054000	0.433594	2.0	2.8	3.00	3.3	4.4
PetalLength	150.0	3.758667	1.764420	1.0	1.6	4.35	5.1	6.9
PetalWidth	150.0	1.198667	0.763161	0.1	0.3	1.30	1.8	2.5

data.describe None

#### Data Count and Values

data count:

SepalLength 150 SepalWidth 150 PetalLength 150 PetalWidth 150 IrisClass 150

dtype: int64

Data Shape

data shape (150, 5)

Displaying Counted Unique Values of Nominal Attributes

Column > IrisClass < unique values</pre>

['Iris-setosa' 'Iris-versicolor' 'Iris-virginica']

Column > IrisClass < unique values count:3</pre>

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

data head:

	SepalLength	SepalWidth	PetalLength	PetalWidth	IrisClass
0	5.1	3.5	1.4	0.2	Iris-setosa
1	4.9	3.0	1.4	0.2	Iris-setosa
2	4.7	3.2	1.3	0.2	Iris-setosa
3	4.6	3.1	1.5	0.2	Iris-setosa
4	5.0	3.6	1.4	0.2	Iris-setosa

None

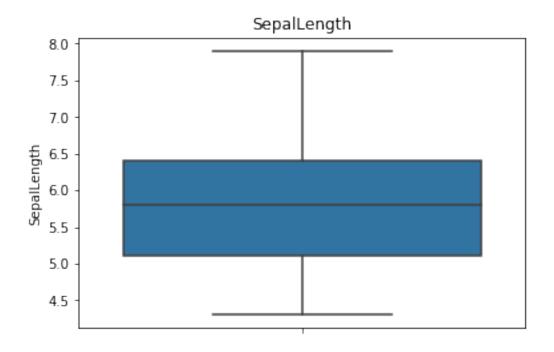
## Columns to be Dropped

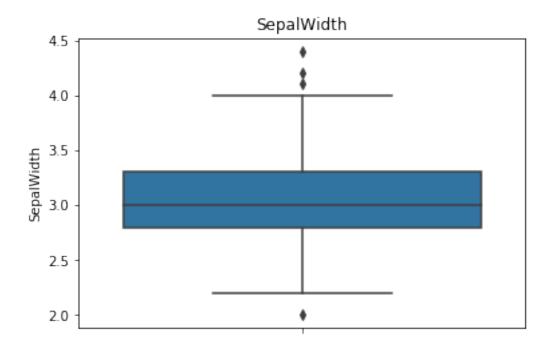
dropping\_columns : []

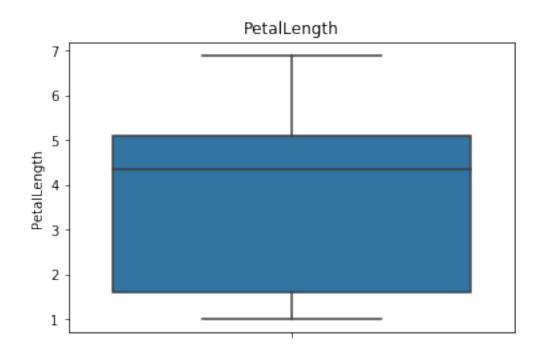
processing\_columns : ['SepalLength', 'SepalWidth', 'PetalLength', 'PetalWidth']

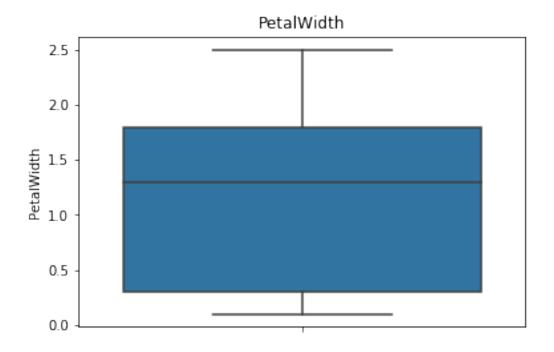
Univariate Analysis

# Box Plots for Numeric Attributes

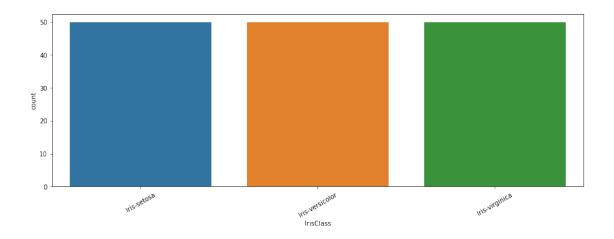








Frequency Charts for Nominal(Categorical) Attributes



IrisClass :

Iris-setosa 50 Iris-versicolor 50 Iris-virginica 50

Name: IrisClass, dtype: int64

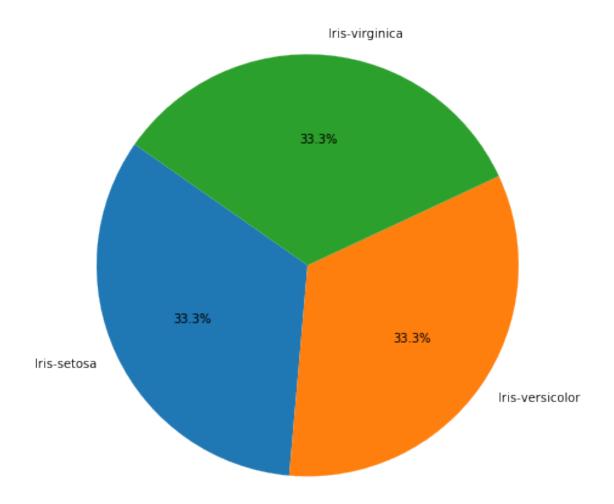
A number of Class : 3

Displaying Binary Variables

Pie Charts for Nominal(Categorical) Attributes

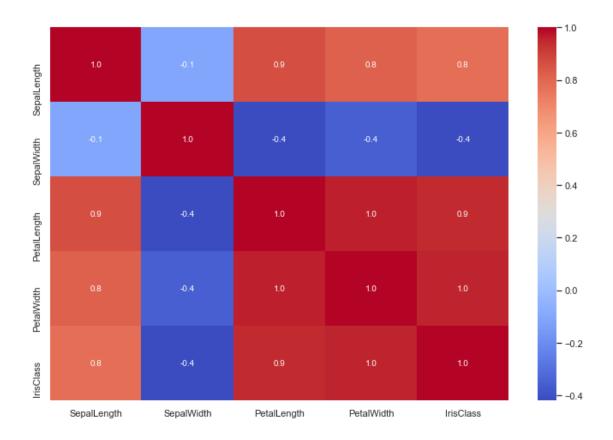
Display Nominal Features Value Counts on pie graph

Displaying IrisClass in pie graph.



Correlation Between Numeric Column

numeric\_values\_for\_correlation: ['SepalLength' 'SepalWidth' 'PetalLength'
'PetalWidth']



Matrix that involves correlation values between numeric columns

	SepalLength	SepalWidth	PetalLength	PetalWidth	IrisClass
SepalLength	1.000000	-0.109369	0.871754	0.817954	0.782561
SepalWidth	-0.109369	1.000000	-0.420516	-0.356544	-0.419446
PetalLength	0.871754	-0.420516	1.000000	0.962757	0.949043
PetalWidth	0.817954	-0.356544	0.962757	1.000000	0.956464
IrisClass	0.782561	-0.419446	0.949043	0.956464	1.000000

corr table: None

Analysis for Values of Nominal Attributes

Nominal Features: IrisClass - Mean Of Numeric Features (First 10)

	IrisClass	SepalLength	SepalWidth	PetalLength	PetalWidth
0	0	5.006	3.418	1.464	0.244
1	1	5.936	2.770	4.260	1.326
2	2	6.588	2.974	5.552	2.026

None

Nominal Features: IrisClass - Count Of Numeric Features (First 10)

IrisClass SepalLength SepalWidth PetalLength PetalWidth 0 0 50 50 50 50

1	1	50	50	50	50
2	2	50	50	50	50

None

Frequency Charts for Nominal(Categorical) Attributes

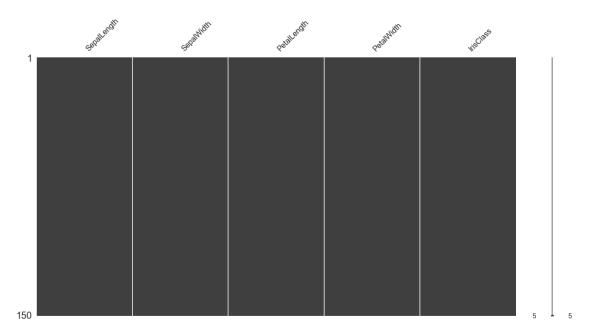
Analysis of Missing(Null) Values

Check null values and processing on data

data null sum each attribute:

SepalLength 0
SepalWidth 0
PetalLength 0
PetalWidth 0
IrisClass 0
dtype: int64

data null sum : 0

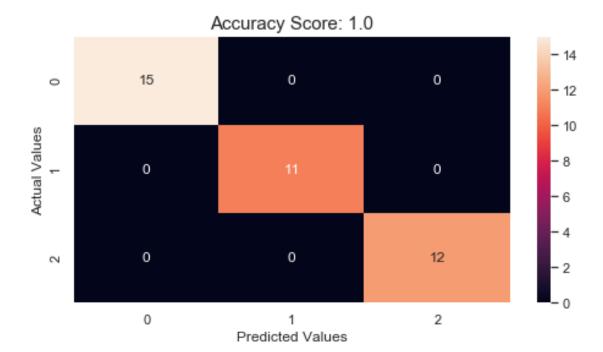


	Null	Values	Sum	%	Value
SepalLength			0		0.0
SepalWidth			0		0.0
PetalLength			0		0.0
PetalWidth			0		0.0
IrisClass			0		0.0

data shape : (150, 5)

```
data null sum:
SepalLength
                0
SepalWidth
               0
PetalLength
               0
PetalWidth
               0
IrisClass
               0
dtype: int64
Encoding Nominal Attributes
data columns Index(['SepalLength', 'SepalWidth', 'PetalLength', 'PetalWidth',
'IrisClass'], dtype='object')
data columns after dropping Index(['SepalLength', 'SepalWidth', 'PetalLength',
'PetalWidth', 'IrisClass'], dtype='object')
Splitting our data as test and train
   SepalLength SepalWidth PetalLength PetalWidth
      0.22222
0
                  0.625000
                               0.067797
                                           0.041667
1
      0.166667
                  0.416667
                               0.067797
                                           0.041667
2
      0.111111
                  0.500000
                               0.050847
                                           0.041667
3
     0.083333
                  0.458333
                               0.084746
                                           0.041667
4
      0.194444
                  0.666667
                               0.067797
                                           0.041667
train data head table: None
Decision Tree Classification
decision_tree_regression_model DecisionTreeClassifier(ccp_alpha=0.0,
class_weight=None, criterion='gini',
                       max_depth=None, max_features=None, max_leaf_nodes=None,
                       min_impurity_decrease=0.0, min_impurity_split=None,
                       min_samples_leaf=1, min_samples_split=2,
                       min_weight_fraction_leaf=0.0, presort='deprecated',
                       random_state=None, splitter='best')
[[15 0 0]
[ 0 11 0]
[ 0 0 12]]
```

accuracy\_score 1.0



		precision	recall	f1-score	support
	0	1.00	1.00	1.00	15
	1	1.00	1.00	1.00	11
	2	1.00	1.00	1.00	12
accui	racy			1.00	38
macro	avg	1.00	1.00	1.00	38
weighted	avg	1.00	1.00	1.00	38

Fitting 10 folds for each of 30 candidates, totalling 300 fits

```
[Parallel(n_jobs=-1)]: Using backend LokyBackend with 8 concurrent workers.
```

[Parallel(n\_jobs=-1)]: Done 25 tasks | elapsed: 3.9s

[Parallel(n\_jobs=-1)]: Done 300 out of 300 | elapsed: 4.6s finished

model\_cv\_name: best\_score : 0.9371212121212119

model\_cv\_name: best\_params : {'max\_depth': 3, 'min\_samples\_split': 10}
model\_cv\_name: best\_estimator : DecisionTreeClassifier(ccp\_alpha=0.0,

class\_weight=None, criterion='gini',

max\_depth=3, max\_features=None, max\_leaf\_nodes=None,
min\_impurity\_decrease=0.0, min\_impurity\_split=None,
min\_samples\_leaf=1, min\_samples\_split=10,
min\_weight\_fraction\_leaf=0.0, presort='deprecated',
random\_state=None, splitter='best')

accuracy\_score 1.0

## Decision tree rules

X[3] <= 0.292 gini = 0.666 samples = 112 value = [35, 39, 38]

gini = 0.0 samples = 35 value = [35, 0, 0] X[2] <= 0.636 gini = 0.5 samples = 77 value = [0, 39, 38]

X[3] <= 0.646 gini = 0.056 samples = 35 value = [0, 34, 1] X[3] <= 0.688 gini = 0.21 samples = 42 value = [0, 5, 37]

gini = 0.0 samples = 34 value = [0, 34, 0] gini = 0.0 samples = 1 value = [0, 0, 1] X[2] <= 0.669 gini = 0.5 samples = 8 value = [0, 4, 4]

X[2] <= 0.653 gini = 0.057 samples = 34 value = [0, 1, 33]

gini = 0.0 samples = 2 value = [0, 2, 0] X[3] <= 0.604 gini = 0.444 samples = 6 value = [0, 2, 4] X[0] <= 0.458 gini = 0.444 samples = 3 value = [0, 1, 2]

gini = 0.0 samples = 31 value = [0, 0, 31]

gini = 0.0 samples = 3 value = [0, 0, 3] X[2] <= 0.754 gini = 0.444 samples = 3 value = [0, 2, 1]

gini = 0.0 samples = 1 value = [0, 1, 0]

gini = 0.0 samples = 2 value = [0, 0, 2]

gini = 0.0 samples = 2 value = [0, 2, 0] gini = 0.0 samples = 1 value = [0, 0, 1]

0