```
from google.colab import drive
drive.mount('/content/drive')
Mounted at /content/drive
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from sklearn.linear model import LogisticRegression
from sklearn.model selection import train test split
from sklearn import metrics
df = pd.read csv('/content/drive/MyDrive/MLDS-LAB/Iris.csv')
df.head()
{"summary":"{\n \"name\": \"df\",\n \"rows\": 150,\n \"fields\": [\
n {\n \"column\": \"Id\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 43,\n
                                                    \"min\": 1,\n
\"max\": 150,\n \"num_unique_values\": 150,\n
\"samples\": [\n
                    74,\n 19,\n
                                                      119\
        ],\n \"semantic_type\": \"\",\n
\"column\":
\"SepalLengthCm\",\n \"properties\": {\n
                                                   \"dtype\":
\"number\",\n \"std\": 0.828066127977863,\n
                                                       \"min\":
4.3,\n \"max\": 7.9,\n \"num unique values\": 35,\n
                        6.2,\n
\"samples\": [\n
                                       4.5,\n
                                                       5.6\n
],\n \"semantic_type\": \"\",\n \"descript
}\n },\n {\n \"column\": \"SepalWidthCm\",\n
                                            \"description\": \"\"\n
                         \"dtype\": \"number\",\n
\"properties\": {\n \"dtype\": \"number\",\n \"std\": 0.4335943113621737,\n \"min\": 2.0,\n \"max\": 4.4,\n
                                  \"samples\": [\n
\"num unique values\": 23,\n
                                                           2.3, n
4.0,\n
                         ],\n
                                       \"semantic type\": \"\",\n
         3.5\n
\"description\": \"\"\n
                           }\n },\n {\n \"column\":
                       \"properties\": {\n
\"PetalLengthCm\",\n
                                                  \"dtype\":
\"number\",\n \"std\": 1.7644204199522626,\n \"min\":
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1.0, n
\"samples\": [\n
                                        3.8,\n
                         6.7,\n
                                                       3.7\n
],\n \"semantic_type\": \"\",\n \"description\": \"\"\n \\n \\n \"column\": \"PetalWidthCm\",\n \"properties\": \\n \"dtype\": \"number\",\n \"std\": 0.7631607417008411,\n \"min\": 0.1,\n \"max\": 2.5,\n
\"num_unique_values\": 22,\n \"samples\": [\n 0.2,\n
\"semantic_type\": \"\",\n
\"Species\",\n \"properties\": {\n \"dtype\":
\"category\",\n \"num_unique_values\": 3,\n
                                                       \"samples\":
[\n \"Iris-setosa\",\n \"Iris-versicolor\",\n
                                   \"semantic_type\": \"\",\n
\"Iris-virginica\"\n
                           ],\n
```

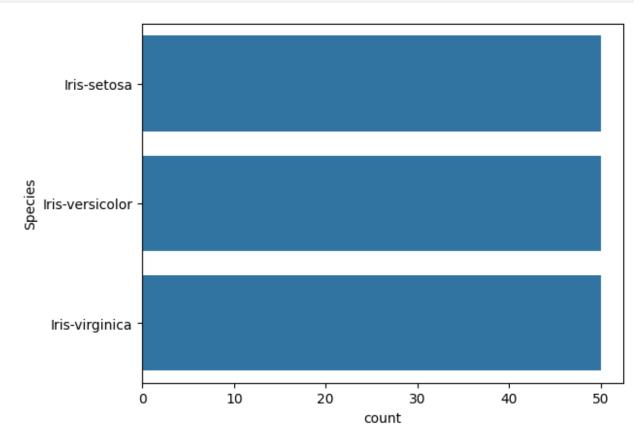
```
n}","type":"dataframe","variable_name":"df"}
df.tail()
{"summary":"{\n \"name\": \"df\",\n \"rows\": 5,\n \"fields\": [\n \]}
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\"semantic_type\": \"\",\n \"description\": \"\"\n
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\"num unique values\": 5,\n
                                        \"samples\": [\n
                                                                      6.3, n
                 \"semantic type\": \"\",\n
\"description\": \"\"\n
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2.5,\n \"max\": 3.4,\n \"num_unique_values\": 3,\n \"samples\": [\n 3.0,\n 2.5,\n 3.4\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n }\n },\n {\n \"column\": \"PetalLengthCm\",\n
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                                                            \"max\": 5.4,\n
\"PetalWidthCm\",\n\\"properties\": {\n\\"dtype\": \"number\",\n\\"std\": 0.23021728866442667,\n\\"min\":
1.8,\n \"max\": 2.3,\n \"num_unique_values\": 4,\n \"samples\": [\n 1.9,\n 1.8,\n 2.3\n
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\"num_unique_values\": 1,\n \"samples\": [\n \"Iris
virginica\"\n ],\n \"semantic_type\": \"\",\n
\"description\": \"\"\n }\n ]\n}","type":"dataframe"}
                                                                      \"Iris-
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 150 entries, 0 to 149
Data columns (total 6 columns):
 #
     Column
                      Non-Null Count
                                         Dtvpe
- - -
      -----
 0
                       150 non-null
     Ιd
                                         int64
     SepalLengthCm 150 non-null
 1
                                         float64
 2
     SepalWidthCm
                      150 non-null
                                         float64
     PetalLengthCm 150 non-null
 3
                                         float64
```

```
PetalWidthCm
                                    150 non-null
                                                                   float64
 5
         Species
                                    150 non-null
                                                                   object
dtypes: float64(4), int64(1), object(1)
memory usage: 7.2+ KB
df.describe()
{"summary":"{\n \"name\": \"df\",\n \"rows\": 8,\n \"fields\": [\n \]}
{\n \"column\": \"Id\",\n \"properties\": {\n
\"dtype\": \"number\",\n \"std\": 53.756293020494844,\n
\"min\": 1.0,\n \"max\": 150.0,\n \"num unique values\":
6,\n \"samples\": [\n 150.0,\n
                                                                                                      75.5,\n
                           ],\n \"semantic_type\": \"\",\n
112.75\n
\"number\",\n\\"std\": 51.24711349471842,\n
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                                                       5.8,\n
                                                                                     150.0\n
                                                                                                                 ],\n
                                                               \"description\": \"\"\n
                                                                                                                }\
         },\n {\n \"column\": \"SepalWidthCm\",\n
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                                                                                                          3.0, n
150.0\n
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                          ],\n
                                                   \"description\": \"\"\n
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\"PetalLengthCm\",\n
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1.0,\n \"max\": 150.0,\n \"num unique values\": 8,\n
\"samples\": [\n 3.75866666666666, \n
                                                                                              4.35,\n
                                                \"semantic_type\": \"\",\n
                          ],\n
150.0\n
                                                              },\n {\n \"column\":
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                                                   }\n
\"PetalWidthCm\",\n \"properties\": {\n
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\"number\",\n
                                       \"std\": 52.636634243409915,\n
                                                                                                             \"min\":
               \"max\": 150.0,\n \"num_unique_values\": 8,\n
0.1, n
\"description\": \"\"\n
                                                 }\n }\n ]\n}","type":"dataframe"}
df.isnull().sum()
Id
                               0
SepalLengthCm
                                0
SepalWidthCm
                                0
PetalLengthCm
                               0
PetalWidthCm
                               0
                               0
Species
dtype: int64
```

```
print(df['Species'].value_counts())
sns.countplot(df['Species'])

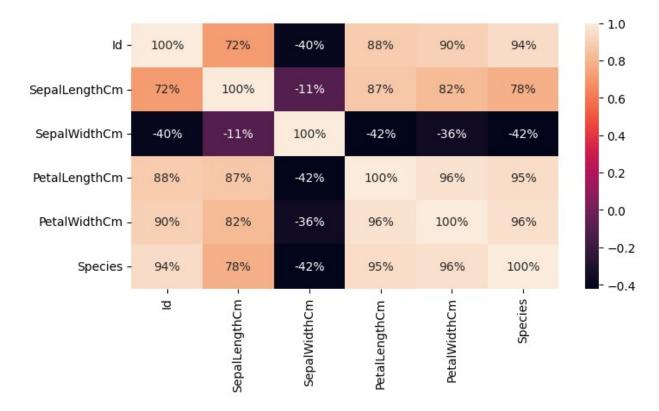
Species
Iris-setosa 50
Iris-versicolor 50
Iris-virginica 50
Name: count, dtype: int64

<Axes: xlabel='count', ylabel='Species'>
```



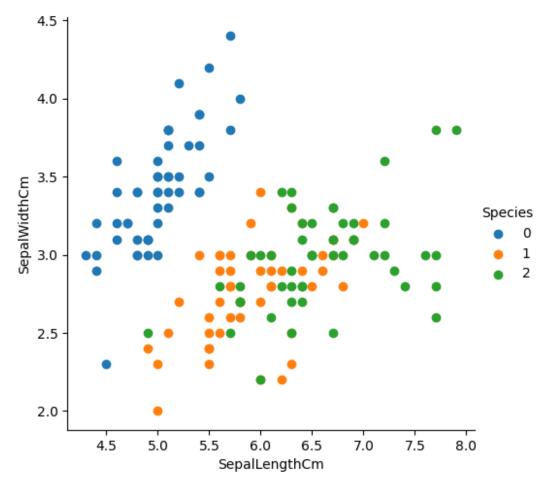
```
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
df['Species'] = le.fit_transform(df['Species'])

plt.figure(figsize=(8,4))
sns.heatmap(df.corr(), annot=True, fmt=".0%")
plt.show()
```

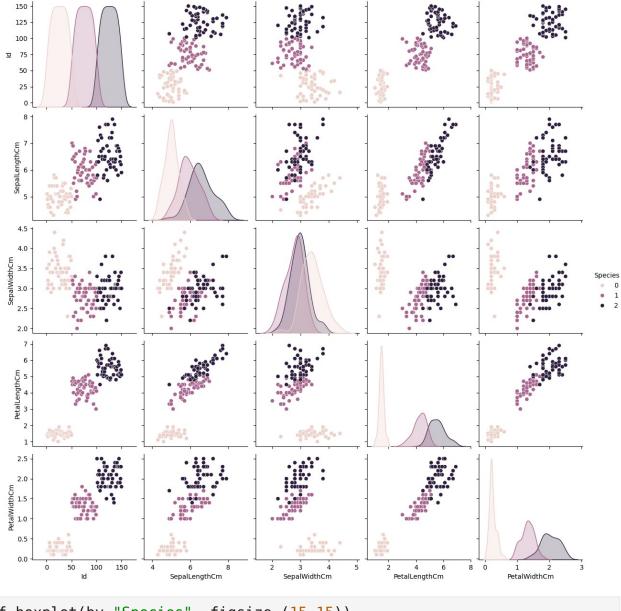


```
sns.FacetGrid(df, hue="Species",
height=5).map(plt.scatter, "SepalLengthCm",
"SepalWidthCm").add_legend()
```

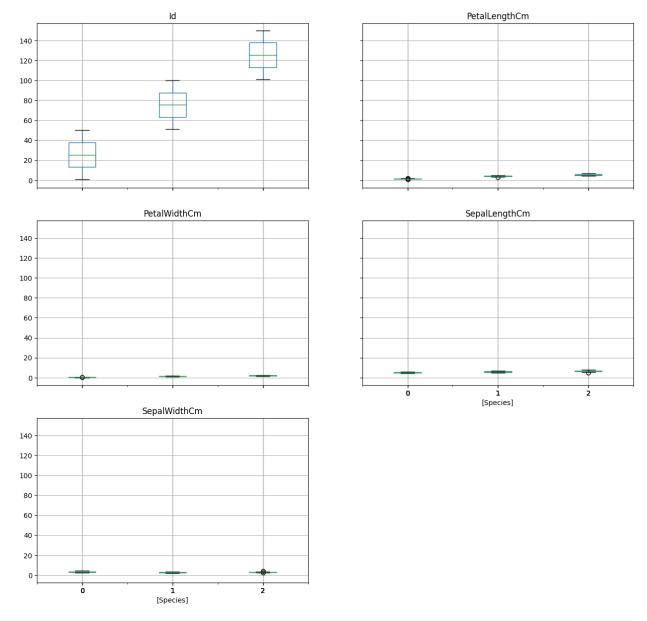
<seaborn.axisgrid.FacetGrid at 0x7a92858ae9b0>



sns.pairplot(df.iloc[:,:],hue='Species')
<seaborn.axisgrid.PairGrid at 0x7a92846a7eb0>



Boxplot grouped by Species



```
df
{"summary":"{\n \"name\": \"df\",\n \"rows\": 150,\n \"fields\": [\
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\"dtype\": \"number\",\n \"std\": 43,\n \"min\": 1,\n
\"max\": 150,\n \"num_unique_values\": 150,\n
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n ],\n \"semantic_type\": \"\",\n
\"description\": \"\"\n }\n {\n \"column\": \"SepalLengthCm\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 0.828066127977863,\n \"min\":
```

```
4.3,\n \"max\": 7.9,\n \"num_unique_values\": 35,\n \"samples\": [\n 6.2,\n 4.5,\n 5.6\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n
}\n },\n {\n \"column\": \"SepalWidthCm\",\n
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\"num_unique_values\": 23,\n \"samples\": [\n 2.3,\n
\"number\",\n \"std\": 1.7644204199522626,\n \"min\":
1.0,\n \"max\": 6.9,\n \"num_unique_values\": 43,\n \"samples\": [\n 6.7,\n 3.8,\n 3.7\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n
}\n    },\n    {\n     \"column\": \"PetalWidthCm\",\n
\"properties\": {\n     \"dtype\": \"number\",\n     \"std\":
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\"num_unique_values\": 22,\n \"samples\": [\n 0.2,\n
n}","type":"dataframe","variable_name":"df"}
df.shape
(150, 6)
X = df.iloc[:,0:4]
y = df['Species']
Χ
{"summary":"{\n \"name\": \"X\",\n \"rows\": 150,\n \"fields\": [\n
{\n \"column\": \"Id\",\n \"properties\": {\n \"dtype\": \"number\",\n \"std\": 43,\n \"
                                                                         \"min\": 1,\n
\"max\": 150,\n \"num_unique_values\": 150,\n \"samples\": [\n 74,\n 19,\n n ],\n \"semantic_type\": \"\",\n
                                                                            119\
4.3,\n \"max\": 7.9,\n \"num_unique_values\": 35,\n \"samples\": [\n 6.2,\n 4.5,\n 5.6\n ],\n \"semantic_type\": \"\",\n \"description\": \"\"\n \\n \\"num_unique_values\": 35,\n \\"samples\": [\n 6.2,\n 4.5,\n 5.6\n \\" description\": \"\"\n \\" olumn\": \"SepalWidthCm\",\n \\" properties\": \\"\" dtype\": \"number\",\n \"std\":
```

```
0.4335943113621737,\n
                            \"min\": 2.0,\n
                                                  \"max\": 4.4,\n
\"num unique values\": 23,\n \"samples\": [\n
                                                            2.3, n
4.0,\n
               3.5\n
                            ],\n
                                       \"semantic_type\": \"\",\n
\"description\": \"\"\n
                                          {\n
                                                \"column\":
                            }\n
                                   },\n
\"PetalLengthCm\",\n
                         \"properties\": {\n
                                                   \"dtype\":
                    \"number\",\n
                                                         \"min\":
             \max: 6.9,\n
                                   \"num unique values\": 43,\n
1.0, n
\"samples\": [\n
                         6.7,\n
                                        3.8,\n
                                                        3.7\n
           \"semantic type\": \"\",\n
                                            \"description\": \"\"\n
],\n
      }\n ]\n}","type":"dataframe","variable name":"X"}
}\n
У
      0
0
1
      0
2
      0
3
      0
4
      0
145
      2
146
      2
      2
147
      2
148
      2
149
Name: Species, Length: 150, dtype: int64
X train, X test, y train, y test = train test split(X, y, test size =
0.2, random state=42)
X train.shape
(120, 4)
X_test.shape
(30, 4)
y train.shape
(120,)
y test.shape
(30,)
log = LogisticRegression()
log.fit(X train, y train)
/usr/local/lib/python3.10/dist-packages/sklearn/linear model/
_logistic.py:460: ConvergenceWarning: lbfgs failed to converge
(status=1):
STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
```

```
Increase the number of iterations (max_iter) or scale the data as
shown in:
    https://scikit-learn.org/stable/modules/preprocessing.html
Please also refer to the documentation for alternative solver options:
https://scikit-learn.org/stable/modules/linear_model.html#logistic-
regression
    n_iter_i = _check_optimize_result(
LogisticRegression()
prediction = log.predict(X_test)
metrics.accuracy_score(prediction,y_test)
1.0
```