

this problem is a multi class classification task of predicting the species of the plant based on 4 features (sepal len,wid and petal len,wid)

LOADING DATASET

```
import pandas as pd

#loading the dataset as a pandas dataframe
df = pd.read_csv("iris.csv")

#prints first 5 cols of the dataset
df.head()
```

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa

```
#shows no of rows and cols in the dataset
df.shape

(150, 5)

#summary statistics of the dataset(only the numerical columns)
df.describe()
```

	sepal_length	sepal_width	petal_length	petal_width
count	150.000000	150.000000	150.000000	150.000000
mean	5.843333	3.054000	3.758667	1.198667
std	0.828066	0.433594	1.764420	0.763161
min	4.300000	2.000000	1.000000	0.100000
25%	5.100000	2.800000	1.600000	0.300000
50%	5.800000	3.000000	4.350000	1.300000
75%	6.400000	3.300000	5.100000	1.800000
max	7.900000	4.400000	6.900000	2.500000

To know how many classes are present in the target variable and also what is the individual no of examples for each class, we can use group by method

```
df.groupby('species').size()

species
setosa      50
versicolor  50
virginica   50
dtype: int64
```

DATA PREPROCESSING:

```
#checking if dataset has null values
df.isnull().sum()

sepal_length    0
sepal_width     0
petal_length    0
petal_width     0
species         0
dtype: int64
```

as there is no null values in the dataset, we can go for further preprocessing steps

```
#splitting dataset into features(x) and target col(y)

X = df.drop(['species'],axis=1).values
y = df['species'].values
```

as y has 3 classes, we need to encode it using label encoder

```
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
y = le.fit_transform(y)
y

array([0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
       0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0,
       0, 0, 0, 0, 0, 0, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
       1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1,
       1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 1, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2,
       2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2,
       2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2, 2])
```

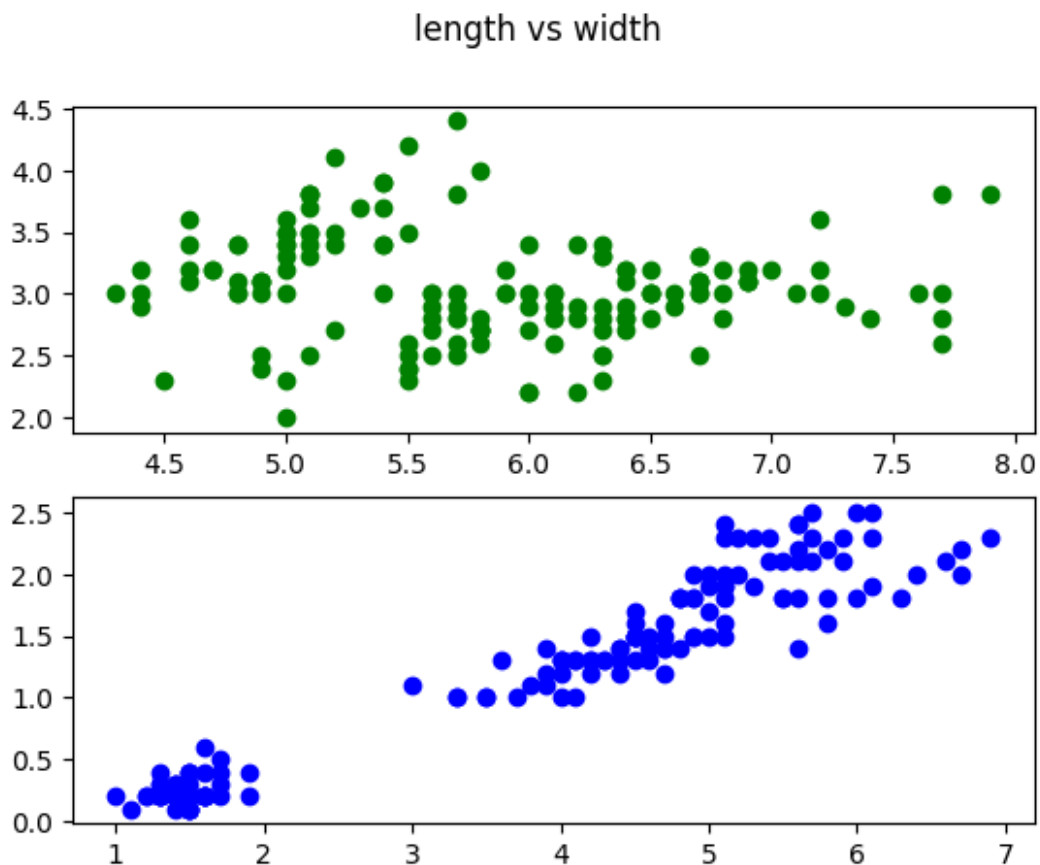
next the dataset is split into train and test sets

```
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size =
0.3, random_state = 0)
```

DATA VISUALISATION

```
#subplots for sepal len vs width and petal len vs width
import matplotlib.pyplot as plt
```

```
fig, axs = plt.subplots(2)
fig.suptitle('length vs width')
axs[0].plot(df.sepal_length,
df.sepal_width,ls='',marker='o',color='g',label='Sepal')
axs[1].plot(df.petal_length,
df.petal_width,ls='',marker='o',color='b',label='Petal')
[<matplotlib.lines.Line2D at 0x7fa9f522e4c0>]
```



MODEL TRAINING

```
#Fitting classifier to the Training set
from sklearn.neighbors import KNeighborsClassifier
from sklearn.model_selection import cross_val_score

#Instantiate model (k = 3)
classifier = KNeighborsClassifier(n_neighbors=3)

#Fitting the model
classifier.fit(X_train, y_train)
```

```
#Predicting the Test set results
y_pred = classifier.predict(X_test)
```

MODEL EVALUATION

```
#confusion matrix
```

```
from sklearn.metrics import confusion_matrix
cm = confusion_matrix(y_test, y_pred)
cm
```

```
array([[16,  0,  0],
       [ 0, 17,  1],
       [ 0,  0, 11]])
```

```
#classification report, it shows the precision, recall and F1 score of
all the 3 classes separately and also the overall accuracy
```

```
from sklearn.metrics import classification_report
print(classification_report(y_test, y_pred))
```

	precision	recall	f1-score	support
0	1.00	1.00	1.00	16
1	1.00	0.94	0.97	18
2	0.92	1.00	0.96	11
accuracy			0.98	45
macro avg	0.97	0.98	0.98	45
weighted avg	0.98	0.98	0.98	45

MODEL INTERPRETATION

=> In class 0, all the exaples in the test set belonging to this class has been correctly predicted

=> In class 1, only 94% of the samples actually belonging to this class have been perdicted correctly (recall=0.94)

=> In class 2, only 92% of the samples predicted as this class actually belong to class 2 (precision=0.92) y

=> In 98% of the cases, the model has predicted the class correctl