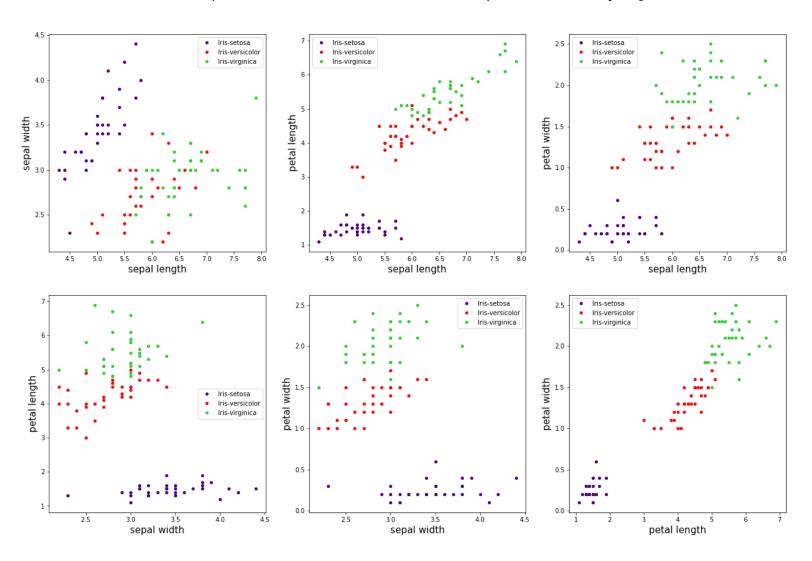
ML Lab 5 Report

Question 1:

1. Preprocessing and Visualisation:

- There were no NULL values or unuseful data. So nothing to drop.
- Then plotted distributions of all the features pair-wise for analysing them.



2. Training QDA:

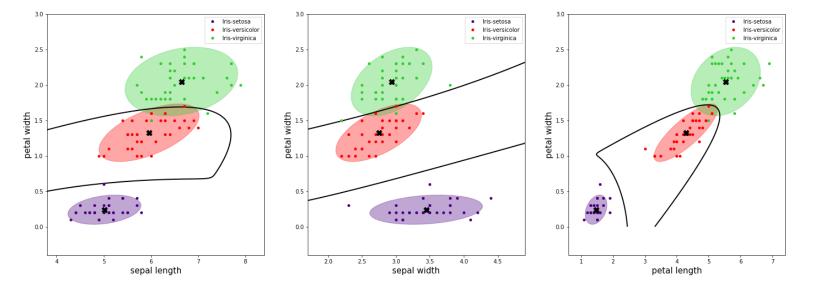
 After visualising the plots, selected 3 pairs which gave the most separation among the classes and used them for training the QDA model.

3. Mean and Covariance Matrix for QDAs:

- Calculated Mean and Covariance matrix for the selected pairs according to the classes.
- One of the 3 pairs is shown here:

4. Decision Boundary:

• From the obtained means, covariance matrices, I plotted the elliptical distributions and then using the predicted probabilities from QDA models, plotted the decision boundary.



5. Class Predictions:

- Using the trained QDA models, predicted the classes for the testing data.
- Found that petal length and petal width pair gives maximum accuracy and minimum error rate. (err_rate = 1-accuracy_score)

```
We got following Error Rates for the following pairs:

For [sepal length , petal width] : 0.0667

For [sepal width , petal width] : 0.0667

For [petal length , petal width] : 0.0444

Hence the best model is obtained from the pair [petal width , petal length]

Minimum Error Rate : 0.0444

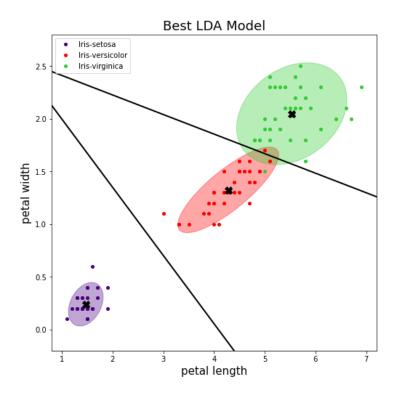
Maximum Accuracy : 95.56 %
```

6. LDA model:

• Trained a LDA model for the best pair using training data.

7. Decision boundary by LDA:

 Predicted probabilities by the LDA and used them for plotting the decision boundaries for all the classes.

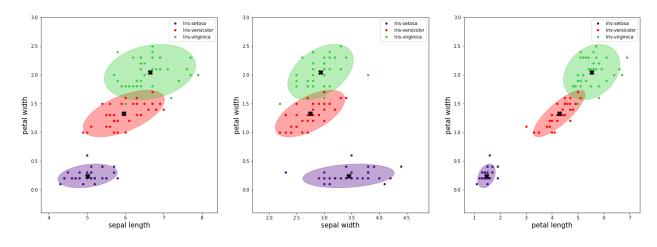


8. Error Rate for LDA:

• Predicted the classes of the testing data and calculated error rates of the same. Found that LDA performs worse than QDA as expected. (LDA underfits the given dataset)

```
For best LDA Model, following are the metrics :
Error Rate : 0.0889
Accuracy score : 91.11 %
```

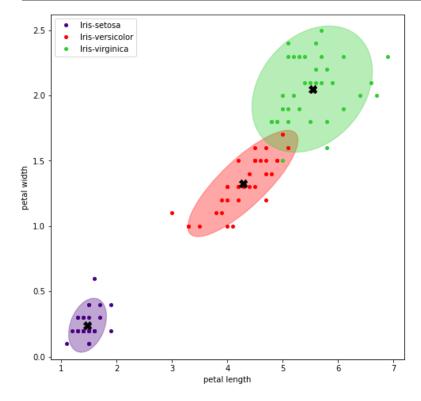
9. Visualisation of Distributions:



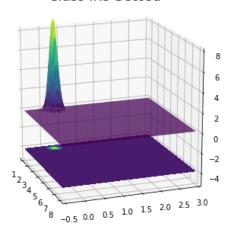
Question 2:

A. Visualising Gaussian Distributions [petal length, petal width]:

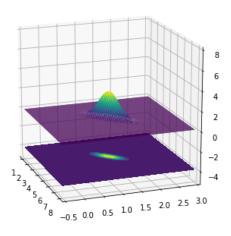
- Calculated Mean and Covariance Matrices and used them to plot ellipses for the class distributions.
- Then plotted gaussian distribution for all the classes in 3 dimensions for better visualisation.



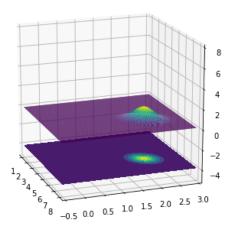
Class Iris-Setosa



Class Iris-Versicolor



Class Iris-Verginica



B. Compute Likelihood:

• Made a function compute_likelihood() for calculating the likelihoods keeping multivariate distribution into account.

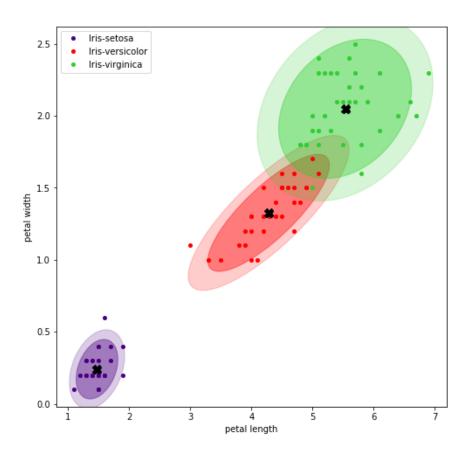
$$p(x|\mu, \Sigma) = \frac{1}{(2\pi)^{n/2} |\Sigma|^{1/2}} \exp\left\{-\frac{1}{2} (x - \mu)^T \Sigma^{-1} (x - \mu)\right\}$$

C. Maximum Likelihood Estimation:

Performed MLE by calculating the class-wise mean and covariance matrix and using them while calculating the likelihoods.

D. Visualising the Distributions of (A and C):

From mean and covariances obtained in part A and C, plotted ellipses for all the classes in the same plot as said. The ellipse represents the contour of the distribution at a certain height.



E. Making Predictions:

Predicted the classes for the testing data points and compared the accuracy score with QDA's accuracy.

We find that QDA performs better than the Naive Bayesian model.

```
Accuracy obtained by Naive Bayes Classifier :
Accuracy : 91.11 %

Accuracy obtained by QDA :
Accuracy : 95.56 %
```

Question 3:

1. Likelihood Calculation:

• Calculated the likelihoods for all the data points in the training dataset.

2. Laplace Smoothing:

• For the probability to remain non-zero for the certain dataset entries, we use laplace smoothing to make the probability a small positive value.

$$p_{\lambda}(C_k) = p_{\lambda}(Y = C_k) = \frac{\sum_{i=1}^{N} I(y_i = C_k) + \lambda}{N + K\lambda}$$

3. Naive Bayes Classification:

- Classified the testing data to the corresponding classes and measured the accuracy scores.
- Also compared with the QDA accuracy, and we see that QDA performs better than the Naive Bayes Classifier.

Accuracy obtained by Naive Bayes Classifier : Accuracy : 91.67 %

Accuracy obtained by QDA:

Accuracy : 95.56 %