

Using the wine dataset:

- Train 2 SVM classifiers to predict the type of wine using a subset of the other 13 variables. You may choose the subset based on previous analysis. One using a linear kernel and another of your choice.

- Use tune.svm to find the optimum C and gamma values.

```
> cat("\n[Linear SVM] Best C =", svm.linear$cost, "\n")
[Linear SVM] Best C = 0.0625

> cat("\n[RBF SVM] Best C =", svm.rbf$cost, " | Best gamma =", svm.rbf$gamma, "\n")
[RBF SVM] Best C = 0.5 | Best gamma = 0.125
>
===== TRAIN =====
> cat("\nLinear SVM Confusion Matrix:\n"); print(res.lin.tr$cm); print(res.lin.tr$metrics)

Linear SVM Confusion Matrix:
   Predicted
Actual  1  2  3
  1 40  2  0
  2  0 50  0
  3  0  1 33
      class precision recall      f1
1        1  1.0000 0.9524 0.9756
2        2  0.9434 1.0000 0.9709
3        3  1.0000 0.9706 0.9851
4 macro_avg  0.9811 0.9743 0.9772
> cat("\nRBF SVM Confusion Matrix:\n");   print(res.rbf.tr$cm); print(res.rbf.tr$metrics)

> cat("\nRBF SVM Confusion Matrix:\n");   print(res.rbf.tr$cm); print(res.rbf.tr$metrics)

RBF SVM Confusion Matrix:
   Predicted
Actual  1  2  3
  1 41  1  0
  2  0 49  1
  3  0  1 33
      class precision recall      f1
1        1  1.0000 0.9762 0.9880
2        2  0.9608 0.9800 0.9703
3        3  0.9706 0.9706 0.9706
4 macro_avg  0.9771 0.9756 0.9763
```

- Choose another classification method (kNN, Random Forest, etc.) and train a classifier based on the same features.

```
> cat("\nRandom Forest Confusion Matrix:\n"); print(res.rf.tr$cm); print(res.rf.tr$metrics)

Random Forest Confusion Matrix:
  Predicted
Actual   1   2   3
  1 42  0  0
  2  0 50  0
  3  0  0 34
  class precision recall f1
1          1        1     1  1
2          2        1     1  1
3          3        1     1  1
4 macro_avg      1     1  1

```

- Compare the performance of the 2 models (Precision, Recall, F1)

==== Test Macro-Average (Precision / Recall / F1) ====

```
> print(cmp)
            precision recall    f1
SVM-linear    0.9512 0.9365 0.9416
SVM-RBF       0.9815 0.9841 0.9823
RandomForest   0.9815 0.9841 0.9823
```

The RBF SVM and Random Forest achieved comparable performance (macro F1 ≈ 0.98), outperforming the Linear SVM (F1 ≈ 0.94). This suggests nonlinear models better capture the complex relationships among selected wine features.