ML Practical List

1. The advertising dataset captures the sales revenue generated with respect to advertisement costs across multiple channels like Radio, TV, and Newspapers.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Build Regression model to predict the sales w.r.t Radio features.
- iii. Also evaluate the model using scores RMSE
- 2. The advertising dataset captures the sales revenue generated with respect to advertisement costs across multiple channels like radio, tv, and newspapers.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Build Regression model to predict the sales w.r.t attribute tv.
- iii. Also evaluate the model using scores RMSE
- 3. The advertising dataset captures the sales revenue generated with respect to advertisement costs across multiple channels like radio, tv, and newspapers.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Build Regression model to predict the sales w.r.t attribute newspaper.
- iii. Also evaluate the model using scores RMSE
- 4. The advertising dataset captures the sales revenue generated with respect to advertisement costs across multiple channels like radio, tv, and newspapers.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Build Regression model to predict the sales w.r.t Radio and TV
- iii. Also evaluate the model using scores RMSE
- 5. The advertising dataset captures the sales revenue generated with respect to advertisement costs across multiple channels like radio, tv, and newspapers.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Build Regression model to predict the sales w.r.t Newspaper and TV
- iii. Also evaluate the model using scores RMSE
- 6. The advertising dataset captures the sales revenue generated with respect to advertisement costs across multiple channels like radio, tv, and newspapers.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Build Regression model to predict the sales w.r.t Newspaper and Radio
- iii. Also evaluate the model using scores RMSE
- 7. The car dataset captures the selling price of the used cars with respect to features like year_bought, km_driven, transmission and owner.

- i. Understand the Dataset & cleanup (if required).
- ii. Build Regression model to predict the selling prices w.r.t year brought
- iii. Also evaluate the model using scores RMSE

8. The car dataset captures the selling price of the used cars with respect to features like year_bought, km_driven, transmission and owner.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Build Regression model to predict the selling prices w.r.t km_driven
- iii. Also evaluate the model using scores RMSE
- 9. The car dataset captures the selling price of the used cars with respect to features like year_bought, km driven, transmission and owner.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Build Regression model to predict the selling prices w.r.t transmission
- iii. Also evaluate the model using scores RMSE
- 10. The car dataset captures the selling price of the used cars with respect to features like year_bought, km_driven, transmission and owner.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Build Regression model to predict the selling prices w.r.t owner
- iii. Also evaluate the model using scores RMSE
- 11. The car dataset captures the selling price of the used cars with respect to features like year_bought, km_driven, transmission and owner.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Build Regression model to predict the selling prices w.r.t year brought and km_driven
- iii. Also evaluate the model using scores RMSE
- 12. The car dataset captures the selling price of the used cars with respect to features like year_bought, km_driven, transmission and owner.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Build Regression model to predict the selling prices w.r.t year brought and transmission
- iii. Also evaluate the model using scores RMSE
- 13. The car dataset captures the selling price of the used cars with respect to features like year_bought, km_driven, transmission and owner.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Build Regression model to predict the selling prices w.r.t year brought and owner
- iii. Also evaluate the model using scores RMSE
- 14. The car dataset captures the selling price of the used cars with respect to features like year_bought, km_driven, transmission and owner.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Build Regression model to predict the selling prices w.r.t year brought and owner
- iii. Also evaluate the model using scores RMSE
- 15. The car dataset captures the selling price of the used cars with respect to features like year_bought, km driven, transmission and owner.

- i. Understand the Dataset & cleanup (if required).
- ii. Build Regression models to predict the selling prices w.r.t km driven and transmission
- iii. Also evaluate the model using scores RMSE

16. The car dataset captures the selling price of the used cars with respect to features like year_bought, km driven, transmission and owner.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Build Regression models to predict the selling prices w.r.t km driven and owner
- iii. Also evaluate the model using scores RMSE
- 17. The car dataset captures the selling price of the used cars with respect to features like year_bought, km driven, transmission and owner.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Build Regression models to predict the selling prices w.r.t transmission and owner
- iii. Also evaluate the model using scores RMSE
- 18. The car dataset captures the selling price of the used cars with respect to features like year_bought, km driven, transmission and owner.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Build Regression models to predict the selling prices w.r.t transmission and owner
- iii. Also evaluate the model using scores RMSE
- 19. A social network dataset is a categorical dataset to determine whether a user purchased a particular product based on gender, age and estimated salary.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Use a SVM to classify whether a user purchased a car or not? (Use Linear Kernel)
- iii. Create a confusion matrix and evaluate the model using accuracy
- 20. A social network dataset is a categorical dataset to determine whether a user purchased a particular product based on gender, age and estimated salary.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Use a SVM to classify whether a user purchased a car or not? (Use Linear Kernel)
- iii. Create a confusion matrix and evaluate the model using Recall.
- 21. A social network dataset is a categorical dataset to determine whether a user purchased a particular product based on gender, age and estimated salary.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Use a SVM to classify whether a user purchased a car or not? (Use Linear Kernel)
- iii. Create a confusion matrix and evaluate the model using Precision
- 22. A social network dataset is a categorical dataset to determine whether a user purchased a particular product based on gender, age and estimated salary.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Use a SVM to classify whether a user purchased a car or not? (Use RBF Kernel)
- iii. Create a confusion matrix and evaluate the model using F1-Measure
- 23. A social network dataset is a categorical dataset to determine whether a user purchased a particular product based on gender, age and estimated salary.

- i. Understand the Dataset & cleanup (if required).
- ii. Use a SVM to classify whether a user purchased a car or not? (Use RBF Kernel)
- iii. Create a confusion matrix and evaluate the model using accuracy

24. A social network dataset is a categorical dataset to determine whether a user purchased a particular product based on gender, age and estimated salary.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Use a SVM to classify whether a user purchased a car or not? (Use RBF Kernel)
- iii. Create a confusion matrix and evaluate the model using Recall.
- 25. A social network dataset is a categorical dataset to determine whether a user purchased a particular product based on gender, age and estimated salary.

Objectives:

- i. Understand the Dataset & cleanup (if required).
- ii. Use a SVM to classify whether a user purchased a car or not? (Use RBF Kernel)
- iii. Create a confusion matrix and evaluate the model using Precision
- 26. A social network dataset is a categorical dataset to determine whether a user purchased a particular product based on gender, age and estimated salary.

- i. Understand the Dataset & cleanup (if required).
- ii. Use a SVM to classify whether a user purchased a car or not? (Use RBF Kernel)
- iii. Create a confusion matrix and evaluate the model using F1-Measure
- 27. The Iris data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant. One class is linearly separable from the other 2; the latter are NOT linearly separable from each other.
 - i. Understand the Dataset & cleanup (if required).
 - ii. Use a SVM to predict the iris plant category? (Use Linear Kernel)
 - iii. Create a confusion matrix and evaluate the model using Precision.
- 28. The Iris data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant. One class is linearly separable from the other 2; the latter are NOT linearly separable from each other.
 - i. Understand the Dataset & cleanup (if required).
 - ii. Use a SVM to predict the iris plant category? (Use Linear Kernel)
- iii. Create a confusion matrix and evaluate the model using Recall.
- 29. The Iris data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant. One class is linearly separable from the other 2; the latter are NOT linearly separable from each other.
 - i. Understand the Dataset & cleanup (if required).
 - ii. Use a SVM to predict the iris plant category? (Use Linear Kernel)
 - iii. Create a confusion matrix and evaluate the model using Accuracy.
- 30. The Iris data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant. One class is linearly separable from the other 2; the latter are NOT linearly separable from each other.
 - i. Understand the Dataset & cleanup (if required).
 - ii. Use a SVM to predict the iris plant category? (Use Linear Kernel)
 - **iii.** Create a confusion matrix and evaluate the model using F1-Measure.
- 31. The Iris data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant. One class is linearly separable from the other 2; the latter are NOT linearly separable from each other.
 - i. Understand the Dataset & cleanup (if required).
 - ii. Use a SVM to predict the iris plant category? (Use RBF Kernel)
- iii. Create a confusion matrix and evaluate the model using Precision.
- 32. The Iris data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant. One class is linearly separable from the other 2; the latter are NOT linearly separable from each other.
 - i. Understand the Dataset & cleanup (if required).
 - ii. Use a SVM to predict the iris plant category? (Use RBF Kernel)

- iii. Create a confusion matrix and evaluate the model using Recall.
- 33. The Iris data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant. One class is linearly separable from the other 2; the latter are NOT linearly separable from each other.
 - i. Understand the Dataset & cleanup (if required).
 - ii. Use a SVM to predict the iris plant category? (Use RBF Kernel)
- iii. Create a confusion matrix and evaluate the model using Accuracy.
- 34. The Iris data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant. One class is linearly separable from the other 2; the latter are NOT linearly separable from each other.
 - i. Understand the Dataset & cleanup (if required).
 - ii. Use a SVM to predict the iris plant category? (Use RBF Kernel)
 - **iii.** Create a confusion matrix and evaluate the model using F1-Measure.
- 35. The Iris data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant.
 - i. Understand the Dataset & cleanup (if required).
 - ii. Build a decision tree classifier to predict the iris plant category? (Use GINI INDEX criteria, use max_depth=4, min_samples_split=2)
- iii. Evaluate the model using Accuracy.
- 36. The Iris data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant.
 - i. Understand the Dataset & cleanup (if required).
 - ii. Build a decision tree classifier to predict the iris plant category? (Use Entropy criteria, use max_depth=4, min_samples_split=2)
- iii. Evaluate the model using Accuracy.
- 37. The Iris data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant.
 - i. Understand the Dataset & cleanup (if required).
 - ii. Build a decision tree classifier to predict the iris plant category? (Use log loss criteria, use max depth=4, min samples split=2)
- iii. Evaluate the model using Accuracy.
- 38. The Iris data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant.
 - i. Understand the Dataset & cleanup (if required).
 - ii. Build a logistic Regression classifier to predict the iris plant category?
 - iii. Evaluate the model using Accuracy.
- 39. The Iris data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant.
- i. Understand the Dataset & cleanup (if required).
- ii. Build a **Bagging Classifier** model to predict the iris plant category?
- iii. Evaluate the model using Accuracy.
- 40. The Iris data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant.
 - i. Understand the Dataset & cleanup (if required).
 - ii. Build a Random Forest Classifier model to predict the iris plant category?
- iii. Evaluate the model using Accuracy.
- 41. The Iris data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant.
- i. Understand the Dataset & cleanup (if required).
- ii. Build a Gradient Boost Classifier model to predict the iris plant category?
- iii. Evaluate the model using Accuracy
- 42. The Iris data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant.
- i. Understand the Dataset & cleanup (if required).
- ii. Build a AdaBoost Classifier model to predict the iris plant category?

- iii. Evaluate the model using Accuracy
- 43. The Iris data set contains 3 classes of 50 instances each, where each class refers to a type of iris plant. Apply Principal Component Analysis on the Iris dataset.
 - i. Understand the Dataset & cleanup (if required).
- ii. Build any one classifier of your choice on the dataset to predict the Iris plant category. (Without Applying PCA)
- iii. Apply **PCA** technique for dimensionality reduction. And build the same classifier that you have chosen in step ii to predict the Iris plant category. Compare the results.