Thyroid Disease Prediction (thyroid.ipynb)

- **Problem Statement**: The project aims to predict whether a patient has thyroid disease (target classes likely 'negative', 'hyperthyroid', 'hypothyroid', etc., making it a multi-class classification problem).
- Dataset Used

• The dataset was loaded from hypothyroid.csv.

• The target variable is 'binaryClass' (P/N), making this a binary classification problem as implemented, although thyroid disease can have more states. The original target in the CSV seems to be the last column which might be a more detailed class, but the notebook focuses on 'binaryClass'.

Methodology and Approach

Data Preprocessing

• Numerous columns with excessive missing values (more than 200 NaNs) were dropped.

• Remaining rows with any missing values were dropped using dropna().

- Categorical features (like 'sex', 'on_thyroxine', 'query_on_thyroxine', etc., and the target 'binaryClass') were converted to numerical representations using LabelEncoder.
- Numerical features were scaled using StandardScaler.

Model Training

• The data was split into training (75%) and testing (25%) sets.

• The following classification models were implemented:

• Logistic Regression (Logistic Regression)

K-Nearest Neighbors (KNeighborsClassifier)

• Support Vector Classifier (SVC)

• Decision Tree Classifier (DecisionTreeClassifier)

Random Forest Classifier (RandomForestClassifier)

• Gaussian Naive Bayes (GaussianNB)

Model Evaluation

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• Models were evaluated using accuracy scores, confusion matrices, and classification reports.

• Results and Conclusion

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- Logistic Regression: Accuracy 97.65%.
- K-Nearest Neighbors: Accuracy 97.52%.
- Support Vector Classifier: Accuracy 97.65%.
- Decision Tree: Accuracy 97.78%.
- Random Forest: Accuracy 98.16%.
- Gaussian Naive Bayes: Accuracy 94.74%.
- The Random Forest Classifier achieved the highest accuracy.
- The Random Forest model was saved to thyroid_model.pkl.