

DSAA 5002 - Data Mining and Knowledge Discovery in Data Science**Final Exam Report – Q5 Smoke Status Recognition****50015940 Jiaxiang Gao****1. Data Preprocessing:**

- a) The 'id' column is removed from the training dataset as it's not relevant for prediction.
- b) Missing values in various columns are handled:
 - For 'height(cm)' and 'waist(cm)', missing values are filled with the mean of the respective columns.
 - For 'eyesight(left)' and 'eyesight(right)', missing values in one eye are replaced with values from the other eye.
 - For 'hearing(right)', missing values are replaced with values from 'hearing(left)'.
 - Missing values in 'Urine protein' are filled with the mean.
- c) Duplicate rows in the training dataset are identified and removed.

2. Model Training and Validation:

- a) The data is scaled using StandardScaler to normalize feature values.
- b) The K-Fold cross-validation approach (with 10 splits) is applied to validate the model's performance.
- c) Two types of models are trained and validated:
 - i. LGBMClassifier: A Light Gradient Boosting Machine classifier.
 - ii. CatBoostClassifier: A classifier from the CatBoost framework.
- d) In each fold of the cross-validation:
 - The model is trained on the training subset.
 - The model's performance is evaluated on the validation subset using the ROC AUC score.
 - The best model is updated if the current model's ROC AUC score is higher than the previously recorded best score.

3. Final Model Prediction:

The model with the highest ROC AUC score from the cross-validation process is selected as the best model. This best model is used to predict the 'smoking' variable on the test dataset.

5.2. LightGBM

```

In 11 1 for train_index, val_index in kf.split(X):
      2     X_train, X_val = X[train_index], X[val_index]
      3     y_train, y_val = y.iloc[train_index], y.iloc[val_index]
      4
      5     model = LGBMClassifier(random_state=50015940, verbose=0)
      6     model.fit(X_train, y_train)
      7
      8     val_predictions = model.predict_proba(X_val)[: , 1]
      9     val_auc = roc_auc_score(y_val, val_predictions)
     10     print(f'Validation ROC Score: {val_auc}')
     11     print(f'BEST ROC Score: {best_auc}')
     12
     13     if val_auc > best_auc:
     14         best_auc = val_auc
     15         best_model = model

```

BEST ROC Score: 0.862899295908313
 Validation ROC Score: 0.8624086168194984
 BEST ROC Score: 0.862899295908313
 Validation ROC Score: 0.8700809969733871
 BEST ROC Score: 0.862899295908313
 Validation ROC Score: 0.8658729199198454
 BEST ROC Score: 0.8700809969733871
 Validation ROC Score: 0.8640661489840153
 BEST ROC Score: 0.8700809969733871
 Validation ROC Score: 0.8654416798100414
 BEST ROC Score: 0.8700809969733871
 Validation ROC Score: 0.8632586726895297
 BEST ROC Score: 0.8700809969733871
 Validation ROC Score: 0.8620072700299117
 BEST ROC Score: 0.8700809969733871
 Validation ROC Score: 0.863367368299281
 BEST ROC Score: 0.8700809969733871

5.3. CatBoost

```

In 12 1 for train_index, val_index in kf.split(X):
      2     X_train, X_val = X[train_index], X[val_index]
      3     y_train, y_val = y.iloc[train_index], y.iloc[val_index]
      4
      5     model = CatBoostClassifier(random_state=50015940, verbose=0)
      6     model.fit(X_train, y_train)
      7
      8     val_predictions = model.predict_proba(X_val)[: , 1]
      9     val_auc = roc_auc_score(y_val, val_predictions)
     10     print(f'Validation ROC Score: {val_auc}')
     11     print(f'BEST ROC Score: {best_auc}')
     12
     13     if val_auc > best_auc:
     14         best_auc = val_auc
     15         best_model = model

```

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BEST ROC Score: 0.8700809969733871
 Validation ROC Score: 0.8668537580837085
 BEST ROC Score: 0.8700809969733871
 Validation ROC Score: 0.8755228561916721
 BEST ROC Score: 0.8700809969733871
 Validation ROC Score: 0.8694767722727569
 BEST ROC Score: 0.8755228561916721
 Validation ROC Score: 0.8682589421248335
 BEST ROC Score: 0.8755228561916721
 Validation ROC Score: 0.8697600122299483
 BEST ROC Score: 0.8755228561916721
 Validation ROC Score: 0.8674647067677526
 BEST ROC Score: 0.8755228561916721
 Validation ROC Score: 0.8670172165483092
 BEST ROC Score: 0.8755228561916721
 Validation ROC Score: 0.8668720515748672
 BEST ROC Score: 0.8755228561916721