

Module 1 - Task 1

Load the train data.

- Import Pandas and alias it as 'pd'.
- Read the CSV file movies train.csv into a Pandas DataFrame named 'train'.
- To import the 'train.csv' file, which is located in the root path of your project, you should use the following path: './train.csv'.
- Inspect the data by calling the variable 'train'.

Helpful Links:

To complete this task, knowledge of Machine Learning is required. You can learn Python and ML using the following resources:

- [Machine Learning Course](#)
- [Python Interview QnA](#)
- [Python Coding Practice](#)

Note: Complete the code, and confirm success with "Run Test".

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Module 1 - Task 2

Load the test data.

- Read the CSV file movies test.csv into a Pandas DataFrame named 'test'.
- To import the 'test.csv' file, which is located in the root path of your project, you should use the following path: './test.csv'.
- Inspect the data by calling the variable 'df'.

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Module 1 - Task 3

Counting Null Values in train data.

- Calculate the count of null (missing) values in a DataFrame named 'train'.
- Use the isnull() method on the 'train' DataFrame to create a new DataFrame where each element is either True (if the corresponding element in 'train' is null) or False (if it's not null).
- Calculate the sum of True values for each column of the new DataFrame, effectively counting the number of null values in each column.
- Store the count of null values for each column in a Pandas Series named 'null_values_train'.

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Module 1 - Task 4

Counting Null Values in test data.

- Calculate the count of null (missing) values in a Pandas DataFrame called 'test'.
- Use the `.isnull()` method on the 'test' DataFrame to create a new DataFrame of the same shape. Each element in this new DataFrame will be either True if the corresponding element in 'test' is null or False if it's not null.
- Calculate the sum of True values for each column of the new DataFrame. This effectively counts the number of null values in each column of the 'test' DataFrame.
- Store the count of null values for each column in a Pandas Series named 'null_values_test'.

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Module 1 - Task 5

Handling Missing Values in train Data.

- Replace missing values (NA) in the 'Bed Grade' and 'City_Code_Patient' columns of the 'train' DataFrame using `.fillna()`.
- Impute missing values with the mode (most frequent value) of the 'Bed Grade' and 'City_Code_Patient' columns.
- Apply the replacement directly to the 'train' DataFrame with `inplace=True` for both steps.

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Module 1 - Task 6

Handling Missing Values in test Data.

- Replace missing values (NA) in the 'Bed Grade' and 'City_Code_Patient' columns of the 'test' DataFrame using .fillna().
- Impute missing values with the mode (most frequent value) of the 'Bed Grade' and 'City_Code_Patient' columns.
- Apply the replacement directly to the 'test' DataFrame with inplace=True for both steps.

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Module 2 - Task 1

Encoding 'Stay' Column in train Data.

- Import the LabelEncoder class from the sklearn.preprocessing module.
- Create an instance of the LabelEncoder class and assign it to the variable 'le'.
- Use the fit_transform method of the LabelEncoder to encode the values in the 'Stay' column of the 'train' DataFrame. The .astype('str') ensures that the values are treated as strings.
- Update the 'Stay' column in the 'train' DataFrame with the transformed values. The original categorical values have now been replaced with numerical labels.

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Module 2 - Task 2

Setting a Default Value.

- In the 'test' DataFrame, assign the value -1 to the 'Stay' column for all rows.
- This operation sets the 'Stay' column for all records in the 'test' DataFrame to have a constant value of -1.

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Module 2 - Task 3

Merging Train and Test Data.

- Create a new DataFrame named 'df' using the `pd.concat` function.
- Concatenate the 'train' and 'test' DataFrames along their rows (vertically).
- Use the `ignore_index=True` parameter to reset the index of the combined DataFrame, ensuring a continuous, new index.
- The resulting 'df' DataFrame now contains all the data from 'train' and 'test' DataFrames in a single, merged DataFrame with a unified index.

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Module 2 - Task 4

Categorical Data Label Encoding.

- Iterate through a list of categorical column names in the 'df' DataFrame: 'Hospital_type_code', 'Hospital_region_code', 'Department', 'Ward_Type', 'Ward_Facility_Code', 'Type of Admission', 'Severity of Illness', and 'Age'.
- For each column in the list, create a new instance of the LabelEncoder class and assign it to the variable 'le'.
- Use the fit_transform method of the LabelEncoder to encode the values in the current categorical column from the 'df' DataFrame. The .astype(str) ensures that the values are treated as strings.
- Update the column in the 'df' DataFrame with the transformed numerical labels, effectively replacing the original categorical values with encoded numerical values.
- This loop applies label encoding to convert categorical columns into numerical labels for all columns in the provided list.

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Module 3 - Task 1

Filtering the Training Dataset.

- Create a new DataFrame called 'train' by filtering the existing DataFrame 'df'.
- The filter is based on a condition:
- This condition selects rows in the 'df' DataFrame where the value in the 'Stay' column is not equal to -1.
- The resulting 'train' DataFrame now contains only the rows where 'Stay' is not equal to -1.

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Module 3 - Task 2

Filtering the Testing Dataset.

- Create a new DataFrame called 'test' by filtering the existing DataFrame 'df'.
- The filter is based on a condition:
- This condition selects rows in the 'df' DataFrame where the value in the 'Stay' column is equal to -1.
- The resulting 'test' DataFrame now contains only the rows where 'Stay' is equal to -1.

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Module 3 - Task 3

Column Removal in the Test DataFrame.

- Create a new DataFrame called 'test1' as a subset of the 'test' DataFrame.
- Use the drop method to remove specific columns from 'test' based on the specified labels.
- The columns being dropped are: 'Stay', 'patientid', 'Hospital_region_code', and 'Ward_Facility_Code'.
- The axis=1 parameter indicates that columns (not rows) should be dropped.
- The resulting 'test1' DataFrame is a modified version of 'test' with the specified columns removed.

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Module 3 - Task 4

Column Removal in the Train DataFrame.

- Create a new DataFrame named 'train1' as a modified version of the 'train' DataFrame.
- Utilize the drop method to eliminate specific columns from the 'train' DataFrame based on the provided column labels.
- The columns being dropped are: 'case_id', 'patientid', 'Hospital_region_code', and 'Ward_Facility_Code'.
- The axis=1 parameter specifies that columns (not rows) should be removed from the DataFrame.

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Module 4 - Task 1

Splitting Data into Training and Testing.

- Import the necessary module, `train_test_split`, from scikit-learn's `model_selection` library for splitting data into training and testing sets.
- Create feature and target variables:
- `X1` represents the feature variables by excluding the 'Stay' column. The 'drop' method is used with `axis=1` to remove the 'Stay' column.
- `y1` represents the target variable and is set to the 'Stay' column.
- Split the data into training and testing sets using `train_test_split`:
- The feature variables are denoted as `X1`, and the target variable is denoted as `y1`.
- Use the `test_size` parameter to specify the proportion of data to be allocated for testing. In this case, it's set to 20% of the data.
- The `random_state` parameter(`random_state = 100`) sets a seed for randomization, ensuring reproducibility of the split.
- After the split, the variables `X_train`, `X_test`, `y_train`, and `y_test` hold the feature and target data for both the training and testing sets.

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Module 4 - Task 2

Training an XGBoost Classifier and Evaluating Accuracy.

- Import the necessary libraries and modules, including xgboost for XGBoost, and accuracy_score from scikit-learn for evaluating the model's accuracy.
- Create an XGBoost classifier instance named 'classifier_xgb' with specified hyperparameters:
- max_depth=4: Maximum depth of the tree.
- learning_rate=0.1: Learning rate for gradient boosting.
- n_estimators=800: The number of boosting rounds (trees).
- objective='multi:softmax': Objective function for multi-class classification.
- reg_alpha=0.5 and reg_lambda=1.5: L1 and L2 regularization terms.
- booster='gbtree': Type of boosting model.
- n_jobs=4: Number of CPU cores used for parallel processing.
- min_child_weight=2: Minimum sum of instance weight (hessian) needed in a child.
- base_score=0.75: Initial prediction made by the model.
- Fit the 'classifier_xgb' model to the training data ('X_train' and 'y_train') and assign it to 'model_xgb'.
- Use the trained 'model_xgb' to make predictions on the test data ('X_test') and store the predictions in 'prediction_xgb'.
- Calculate the accuracy of the XGBoost model's predictions by comparing 'prediction_xgb' with the true test labels ('y_test') using the accuracy_score function. The accuracy score is stored in 'acc_score_xgb'.
- Round the accuracy score to two decimal places.

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Module 4 - Task 3

Post-processing Model Predictions.

- Use the trained 'classifier_xgb' model to make predictions on the 'test1' DataFrame. The 'predict' method is used to obtain predictions, and these predictions are stored in 'pred_xgb'.
- Exclude the first column(case_id) from the 'test1' when making predictions.
- Create a new DataFrame named 'result_xgb' to organize the prediction results. The 'pred_xgb' values are added to this DataFrame with the column name 'Stay'.
- Assign the 'case_id' column from 'test1' to the 'case_id' column in the 'result_xgb' DataFrame.
- Reorder the columns in 'result_xgb' to have 'case_id' as the first column and 'Stay' as the second column.
- Replace the numeric labels in the 'Stay' column of 'result_xgb' with meaningful categories using the .replace() method. The numeric labels are mapped to their corresponding 'Stay' categories, making the results more interpretable.
- The 'result_xgb' DataFrame now contains the 'case_id' and 'Stay' columns with predictions converted into human-readable stay categories.

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Module 4 - Task 4

Counting Unique Cases per Stay Category.

- Utilize the 'result_xgb' DataFrame to perform a grouping operation based on the 'Stay' column.
- Group the data by unique 'Stay' values using the groupby method.
- Calculate the number of unique 'case_id' values in each 'Stay' group by applying the .nunique() method to the 'case_id' column.
- The result is stored in the variable 'result', which is a Pandas Series containing the count of unique 'case_id' values for each 'Stay' category.

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