pip install ucimlrepo

```
→ Collecting ucimlrepo

       Downloading ucimlrepo-0.0.7-py3-none-any.whl.metadata (5.5 kB)
     Requirement already satisfied: pandas>=1.0.0 in /usr/local/lib/python3.11/dist-packages (from ucimlrepo) (2.2.2)
     Requirement already satisfied: certifi>=2020.12.5 in /usr/local/lib/python3.11/dist-packages (from ucimlrepo) (2025.1.31)
     Requirement already satisfied: numpy>=1.23.2 in /usr/local/lib/python3.11/dist-packages (from pandas>=1.0.0->ucimlrepo) (2.0.2)
     Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.11/dist-packages (from pandas>=1.0.0->ucimlrepo) (2.8.2)
     Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.11/dist-packages (from pandas>=1.0.0->ucimlrepo) (2025.1)
     Requirement already satisfied: tzdata>=2022.7 in /usr/local/lib/python3.11/dist-packages (from pandas>=1.0.0->ucimlrepo) (2025.1)
     Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.11/dist-packages (from python-dateutil>=2.8.2->pandas>=1.0.0->ucimlrep
     Downloading ucimlrepo-0.0.7-py3-none-any.whl (8.0 kB)
     Installing collected packages: ucimlrepo
     Successfully installed ucimlrepo-0.0.7
#First two steps are to bring in the data from the referenced website
from ucimlrepo import fetch_ucirepo
national_health_and_nutrition_health_survey_2013_2014_nhanes_age_prediction_subset = fetch_ucirepo(id=887)
X = national_health_and_nutrition_health_survey_2013_2014_nhanes_age_prediction_subset.data.features
y = {\tt national\_health\_and\_nutrition\_health\_survey\_2013\_2014\_nhanes\_age\_prediction\_subset.data.targets}
print(national_health_and_nutrition_health_survey_2013_2014_nhanes_age_prediction_subset.metadata)
print(national_health_and_nutrition_health_survey_2013_2014_nhanes_age_prediction_subset.variables)
🚁 {'uci_id': 887, 'name': 'National Health and Nutrition Health Survey 2013-2014 (NHANES) Age Prediction Subset', 'repository_url': 'https
             name
                      role
                                   type demographic \
     0
             SEQN
                       ID
                             Continuous
                                               None
                    Target Categorical
     1
        age group
                                                Age
         RIDAGEYR
                             Continuous
     2
                     0ther
                                                Age
     3
         RIAGENDR
                   Feature
                             Continuous
                                             Gender
     4
           PAQ605
                             Continuous
                   Feature
                                               None
     5
           BMXBMI
                  Feature
                             Continuous
                                               None
     6
           LBXGLU Feature
                             Continuous
                                               None
     7
           DI0010 Feature
                             Continuous
                                               None
     8
           LBXGLT
                   Feature
                             Continuous
                                               None
     9
            LBXIN Feature
                             Continuous
                                               None
                                              description units missing values
     0
                               Respondent Sequence Number None
     1
               Respondent's Age Group (senior/non-senior)
                                                           None
                                                                             no
                                         Respondent's Age None
                                                                            no
                                      Respondent's Gender None
     3
                                                                            nο
        If the respondent engages in moderate or vigor...
     4
                                                           None
                                                                             no
                             Respondent's Body Mass Index None
                                                                            no
     6
                 Respondent's Blood Glucose after fasting None
                                                                             no
                            If the Respondent is diabetic
                                                           None
                                                                             no
                                       Respondent's Oral
     8
     9
                        Respondent's Blood Insulin Levels None
                                                                             no
     4
import pandas as pd
import numpy as np
#Create a dataframe with Features and Targets
national = national_health_and_nutrition_health_survey_2013_2014_nhanes_age_prediction_subset
df = pd.DataFrame(national.data.features)
df = df.assign(age_group=national.data.targets)
df = df.assign(RIDAGEYR=national.data.targets)
#Index the insulin levels to create a new column with assigned labels based off the value in the insulin column
df['Category'] = pd.cut(df['LBXIN'],
                        bins=[0, 10, 15, 30, 110],
                        labels=['Low', 'Medium', 'High', 'Very High'],
                        right=False.
                        include_lowest=True)
```

df.head()

```
<del>_</del>
         RIAGENDR PAQ605 BMXBMI LBXGLU DIQ010 LBXGLT LBXIN age_group RIDAGEYR Category
                                                                                                        \blacksquare
      0
               2.0
                                                              14.91
                       2.0
                               35.7
                                      110.0
                                                 2.0
                                                       150.0
                                                                          Adult
                                                                                     Adult
                                                                                             Medium
                                                                                                        th
      1
               2.0
                       2.0
                               20.3
                                       89.0
                                                 2.0
                                                        80.0
                                                               3.85
                                                                          Adult
                                                                                     Adult
                                                                                                Low
      2
               1.0
                       2.0
                               23.2
                                       89.0
                                                 2.0
                                                        68.0
                                                               6.14
                                                                          Adult
                                                                                     Adult
                                                                                                Low
      3
               1.0
                       2.0
                               28.9
                                      104.0
                                                 2.0
                                                        84.0
                                                              16.15
                                                                          Adult
                                                                                     Adult
                                                                                                High
               2.0
                       1.0
                               35.9
                                      103.0
                                                 2.0
                                                        81.0
                                                              10.92
                                                                          Adult
                                                                                     Adult
                                                                                             Medium
              Generate code with df
                                      View recommended plots
                                                                    New interactive sheet
 Next steps:
#Using Label encoding to change Adult and Senior into 1 and 2 for the Cross Validation
d = {'Adult': 1, 'Senior': 2}
df['age_group'] = df['age_group'].map(d)
df['RIDAGEYR'] = df['RIDAGEYR'].map(d)
print(df)
RIAGENDR
                      PAQ605
                               BMXBMI
                                       LBXGLU
                                                DIQ010
                                                        LBXGLT LBXIN
                                                                        age_group
     0
                                 35.7
                                        110.0
                                                         150.0
                 2.0
                         2.0
                                                   2.0
                                                                 14.91
                                                                                 1
     1
                 2.0
                          2.0
                                 20.3
                                         89.0
                                                   2.0
                                                          80.0
                                                                  3.85
                                                                                 1
     2
                 1.0
                          2.0
                                 23.2
                                         89.0
                                                   2.0
                                                           68.0
                                                                  6.14
                                                                                 1
     3
                 1.0
                          2.0
                                 28.9
                                        104.0
                                                   2.0
                                                           84.0
                                                                 16.15
                                                                                 1
     4
                 2.0
                         1.0
                                 35.9
                                        103.0
                                                   2.0
                                                          81.0
                                                                 10.92
                                                                                 1
     2273
                 2.0
                          2.0
                                 33.5
                                        100.0
                                                   2.0
                                                           73.0
                                                                  6.53
                                                                                 1
     2274
                                                          208.0
                 1.0
                          2.0
                                 30.0
                                         93.0
                                                   2.0
                                                                 13.02
                                                                                 1
     2275
                 1.0
                          2.0
                                 23.7
                                        103.0
                                                   2.0
                                                          124.0
                                                                 21.41
                                                                                 1
     2276
                 2.0
                          2.0
                                 27.4
                                         90.0
                                                          108.0
                                                                  4.99
                                                   2.0
                                                                                 1
     2277
                 1.0
                          2.0
                                 24.5
                                        108.0
                                                   2.0
                                                         108.0
                                                                  3.76
                                                                                 1
           RIDAGEYR Category
     0
                       Medium
                   1
     1
                   1
                          Low
     2
                   1
                          Low
     3
                   1
                         High
     4
                   1
                       Medium
     2273
                   1
                           Low
     2274
                   1
                       Medium
                         High
     2275
                   1
     2276
                   1
                          Low
                   1
                          Low
     [2278 rows x 10 columns]
import pandas
from sklearn import tree
from sklearn.tree import DecisionTreeClassifier
import matplotlib.pyplot as plt
#To give more familar names to the column headers
df.rename(columns={'RIAGENDR': 'Gender', 'PAQ605': 'Exercise', 'BMXBMI': 'BMI', 'LBXGLU': 'Glucose', 'DIQ010': 'Diabetic', 'LBXGLT': 'Oral_6
print(df)
<del>_</del>
                               BMI
           Gender
                   Exercise
                                     Glucose Diabetic Oral Glucose Insulin \
     0
               2.0
                         2.0
                               35.7
                                       110.0
                                                    2.0
                                                                 150.0
                                                                           14.91
               2.0
                          2.0
                              20.3
                                        89.0
                                                    2.0
                                                                  80.0
                                                                            3.85
     1
                                                                  68.0
     2
              1.0
                          2.0
                               23.2
                                        89.0
                                                    2.0
                                                                            6.14
     3
              1.0
                          2.0
                               28.9
                                       104.0
                                                    2.0
                                                                  84.0
                                                                           16.15
               2.0
                         1.0
                              35.9
                                       103.0
                                                    2.0
                                                                  81.0
                                                                           10.92
                         2.0
     2273
               2.0
                              33.5
                                       100.0
                                                    2.0
                                                                  73.0
                                                                            6.53
     2274
                          2.0
                              30.0
                                        93.0
                                                                 208.0
                                                                           13.02
               1.0
                                                    2.0
     2275
               1.0
                          2.0
                               23.7
                                       103.0
                                                    2.0
                                                                 124.0
                                                                           21.41
     2276
              2.0
                         2.0
                              27.4
                                        90.0
                                                    2.0
                                                                 108.0
                                                                            4.99
     2277
               1.0
                         2.0 24.5
                                       108.0
                                                    2.0
                                                                 108.0
                                                                            3.76
                       Age Insulin_Category
            age_group
     0
                    1
                         1
                                       Medium
     1
                    1
                         1
                                         Low
     2
                    1
                         1
                                         Low
                         1
                                        High
```

```
4
                   1
                       1
                                    Medium
     2273
                                       Low
     2274
                   1
                                    Medium
                        1
     2275
                   1
                        1
                                      High
     2276
                   1
     2277
                                       Low
     [2278 rows x 10 columns]
from sklearn.model_selection import cross_val_score, KFold
from sklearn.svm import SVC
#Run the cross validation with 10 folds
X = df.drop(columns=['Insulin_Category', 'Age'])
Y = df['Insulin_Category']
svm_classifier = SVC(kernel='linear')
num_folds = 10
kf = KFold(n_splits=num_folds, shuffle=True, random_state=42)
cross val results = cross val score(svm classifier, X, y, cv=kf)
🚁 /usr/local/lib/python3.11/dist-packages/sklearn/utils/validation.py:1408: DataConversionWarning: A column-vector y was passed when a 1d
       y = column or 1d(y, warn=True)
     /usr/local/lib/python3.11/dist-packages/sklearn/utils/validation.py:1408: DataConversionWarning: A column-vector y was passed when a 1d
      y = column_or_1d(y, warn=True)
     /usr/local/lib/python3.11/dist-packages/sklearn/utils/validation.py:1408: DataConversionWarning: A column-vector y was passed when a 1d
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     /usr/local/lib/python3.11/dist-packages/sklearn/utils/validation.py:1408: DataConversionWarning: A column-vector y was passed when a 1d
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     /usr/local/lib/python3.11/dist-packages/sklearn/utils/validation.py:1408: DataConversionWarning: A column-vector y was passed when a 1d
       y = column or 1d(y, warn=True)
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       y = column_or_1d(y, warn=True)
     /usr/local/lib/python3.11/dist-packages/sklearn/utils/validation.py:1408: DataConversionWarning: A column-vector y was passed when a 1d
       y = column_or_1d(y, warn=True)
print("Cross-Validation Results (Accuracy):")
for i, result in enumerate(cross_val_results, 1):
    print(f" Fold {i}: {result * 100:.2f}%")
print(f'Mean Accuracy: {cross_val_results.mean()* 100:.2f}%')
→ Cross-Validation Results (Accuracy):
       Fold 1: 100.00%
       Fold 2: 100.00%
       Fold 3: 100.00%
       Fold 4: 100.00%
       Fold 5: 100.00%
       Fold 6: 100.00%
       Fold 7: 100.00%
       Fold 8: 100.00%
       Fold 9: 100.00%
       Fold 10: 100.00%
     Mean Accuracy: 100.00%
```