```
In [221...
         # ! pip install -q -U keras-tuner
In [222...
         import pandas as pd
          import numpy as np
          import seaborn as sns
          import matplotlib.pyplot as plt
          import matplotlib.image as mpimg
          from datetime import time
          # preprocessing and pipeline
          from sklearn.model_selection import train_test_split, cross_val_score, GridSearc
          from sklearn.compose import ColumnTransformer
          from sklearn.preprocessing import OneHotEncoder, LabelEncoder, StandardScaler
          from sklearn.pipeline import Pipeline
          from outlier_cleaner import OutlierCleaner
          # sklearn libraries
          from sklearn.linear_model import LinearRegression, Ridge, Lasso
          from sklearn.svm import SVC, SVR
          from sklearn.tree import DecisionTreeRegressor, DecisionTreeClassifier
          from sklearn.neighbors import KNeighborsRegressor, KNeighborsClassifier
          from sklearn.ensemble import (
              RandomForestRegressor, GradientBoostingRegressor, AdaBoostRegressor,
              VotingRegressor, StackingRegressor, RandomForestClassifier, VotingClassifier
          # sklearn metrics
          from sklearn.metrics import r2_score, mean_squared_error, accuracy_score, confus
          from extended_sklearn_metrics import evaluate_model_with_cross_validation
          # extra model
          import xgboost as xgb
          import joblib
          # deep Learning
          from imblearn.over_sampling import SMOTE
          import tensorflow as tf
          from tensorflow.keras import layers, models
          import keras tuner as kt
          from tensorflow import keras
In [223...
         df = pd.read csv('heart attack prediction indonesia.csv', nrows=30000)
          df1 = df.copy()
```

Feature Engineering (higher values = more risk)

```
In [224... df['alcohol_consumption'] = df['alcohol_consumption'].fillna('None')
stress_mapping = {
    'Low': 1,
    'Moderate': 2,
    'High': 3
}
df['stress_level'] = df['stress_level'].map(stress_mapping)
smoking_mapping = {
```

```
'Never': 1,
     'Past': 2,
     'Current': 3
}
df['smoking_status'] = df['smoking_status'].map(smoking_mapping)
alcohol consumption mapping = {
     'None': 1,
     'Moderate': 2,
     'High': 3
df['alcohol_consumption'] = df['alcohol_consumption'].map(alcohol_consumption_ma
physical_activity_mapping = {
     'Low': 1,
     'Moderate': 2,
     'High': 3
}
df['physical_activity'] = df['physical_activity'].map(physical_activity_mapping)
dietary_habits_mapping = {
     'Unhealthy': 1,
     'Healthy': 2,
df['dietary_habits'] = df['dietary_habits'].map(dietary_habits_mapping)
air_pollution_mapping = {
     'Low': 1,
     'Moderate': 2,
     'High': 3
df['air_pollution_exposure'] = df['air_pollution_exposure'].map(air_pollution_ma
income_level_mapping = {
    'Low': 3,
     'Middle': 2,
     'High': 1
df['income_level'] = df['income_level'].map(income_level_mapping)
df['age group'] = np.where(
     df['age'] < 40, 1, np.where(</pre>
         df['age'] < 55, 2, np.where(
             df['age'] < 65, 3, 4
    )
df.age_group.unique()
array([3, 2, 4, 1])
```

```
Out[224...
```

```
df.columns.tolist()
In [225...
```

```
Out[225...
           ['age',
            'gender',
            'region',
            'income_level',
            'hypertension',
            'diabetes',
            'cholesterol level',
            'obesity',
            'waist circumference',
            'family_history',
            'smoking_status',
            'alcohol_consumption',
            'physical_activity',
            'dietary_habits',
            'air_pollution_exposure',
            'stress level',
            'sleep_hours',
            'blood_pressure_systolic',
            'blood_pressure_diastolic',
            'fasting_blood_sugar',
            'cholesterol hdl',
            'cholesterol_ldl',
            'triglycerides',
            'EKG_results',
            'previous_heart_disease',
            'medication_usage',
            'participated in free screening',
            'heart_attack',
            'age_group']
In [226...
          df['health_risk_score'] = (
              df['hypertension'] +
              df['diabetes'] +
              df['obesity'] +
              df['family history'] +
              df['smoking_status'] +
              df['alcohol_consumption'] +
              (1 - df['physical_activity']) +
              (1 - df['dietary_habits']) +
              df['air pollution exposure'] +
              df['stress_level'] +
              df.income level
          )
          # df['cholesterol ratio'] = df['cholesterol hdl']/df['cholesterol ldl']
          # df = df.drop(labels=['cholesterol hdl', 'cholesterol ldl'], axis=1)
          df['obesity_risk_score'] = np.where(df['gender'] == 'Male' , df['obesity'] * 1 +
          # df = df.drop(labels=['obesity', 'waist_circumference'], axis=1)
          # df['pulse_pressure'] = df['blood_pressure_systolic'] - df['blood_pressure_dias
          # df = df.drop(labels=['blood_pressure_systolic', 'blood_pressure_diastolic'], a
          df['stress to sleep ratio'] = df['stress level'] / df['sleep hours']
          # df = df.drop(columns=['stress_level', 'sleep_hours'])
          df['mean arterial pressure'] = (2* df['blood pressure systolic'] + df['blood pre
          # df = df.drop(columns=['blood_pressure_systolic', 'blood_pressure_diastolic'])
          df['triglyceride-hdl-ratio'] = df['triglycerides'] / df['cholesterol_hdl']
          # df = df.drop(columns=['triglycerides', 'cholesterol_hdl'])
In [227...
          # df risk score = df[['health risk score', 'hypertension', 'diabetes', 'obesity r
          # df_risk_score.corr().sort_values(by='health_risk_score', ascending=False).styl
```

```
In [228...
          # df = df.drop(columns=['physical_activity', 'dietary_habits'])
          df.isna().any()
In [229...
Out[229...
                                              False
           gender
                                              False
           region
                                              False
           income_level
                                              False
           hypertension
                                              False
           diabetes
                                              False
           cholesterol level
                                              False
           obesity
                                              False
           waist_circumference
                                              False
           family_history
                                              False
           smoking_status
                                              False
           alcohol_consumption
                                              False
           physical_activity
                                             False
                                             False
           dietary_habits
           air_pollution_exposure
                                             False
           stress_level
                                              False
           sleep_hours
                                             False
           blood_pressure_systolic
                                             False
           blood_pressure_diastolic
                                             False
           fasting_blood_sugar
                                              False
           cholesterol_hdl
                                             False
           cholesterol_ldl
                                              False
                                              False
           triglycerides
           EKG_results
                                              False
           previous_heart_disease
                                              False
           medication_usage
                                              False
           participated_in_free_screening
                                              False
           heart_attack
                                              False
           age_group
                                              False
           health_risk_score
                                              False
           obesity_risk_score
                                              False
           stress_to_sleep_ratio
                                              False
           mean_arterial_pressure
                                              False
           triglyceride-hdl-ratio
                                              False
           dtype: bool
In [230...
          # df.health risk score.unique()
```

Encoding

```
In [231... ## Encoding
    cat_df = df.select_dtypes(include='object')
    num_df = df.select_dtypes(exclude='object')
    encoder = LabelEncoder()
    for cols in cat_df:
        cat_df[cols+'_encoded'] = encoder.fit_transform(cat_df[cols])
    cat_df = cat_df.select_dtypes(exclude='object')
    df = pd.concat([cat_df, num_df], axis=1)
```

Cleaning Outliers

```
In [232... cleaner = OutlierCleaner(df, preserve_index=True)
    cleaned_df, info = cleaner.clean_columns(
        method='zscore',
        show_progress=True
)
    df = cleaned_df

Cleaning columns: 100%| 34/34 [00:00<00:00, 188.79it/s]</pre>
```

Visualization of correlation of features with heart attack

```
In [233...
           df.waist_circumference.min(), df.waist_circumference.max()
           # df.waist_circumference.unique()
Out[233...
          (45, 142)
In [234...
          df.columns
Out[234... Index(['gender_encoded', 'region_encoded', 'EKG_results_encoded', 'age',
                   'income_level', 'hypertension', 'diabetes', 'cholesterol_level',
                   'obesity', 'waist_circumference', 'family_history', 'smoking_status',
                   'alcohol_consumption', 'physical_activity', 'dietary_habits',
                   'air_pollution_exposure', 'stress_level', 'sleep_hours',
'blood_pressure_systolic', 'blood_pressure_diastolic',
                   'fasting_blood_sugar', 'cholesterol_hdl', 'cholesterol_ldl',
                   'triglycerides', 'previous_heart_disease', 'medication_usage',
                   'participated_in_free_screening', 'heart_attack', 'age_group',
                   'health_risk_score', 'obesity_risk_score', 'stress_to_sleep_ratio',
                   'mean_arterial_pressure', 'triglyceride-hdl-ratio'],
                  dtype='object')
           results = []
In [235...
           for i in df1.columns:
               results.append({
                    f'{i}': df1[i].unique()
               })
           results
```

```
[{'age': array([60, 53, 62, 73, 52, 64, 49, 61, 57, 32, 34, 48, 42, 58, 44, 38,
72,
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 {'physical_activity': array(['High', 'Moderate', 'Low'], dtype=object)},
 {'dietary_habits': array(['Unhealthy', 'Healthy'], dtype=object)},
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 {'stress_level': array(['Moderate', 'High', 'Low'], dtype=object)},
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          83,
```

```
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```

```
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           {'medication_usage': array([0, 1], dtype=int64)},
           {'participated_in_free_screening': array([0, 1], dtype=int64)},
           {'heart_attack': array([0, 1], dtype=int64)}]
In [236...
          results = []
          for i in df.columns:
              results.append({
                  f'{i}': df[i].unique()
              })
          results
```

```
[{'gender_encoded': array([1, 0])},
{'region_encoded': array([0, 1])},
 {'EKG_results_encoded': array([1, 0])},
 {'age': array([60, 53, 62, 73, 52, 64, 49, 61, 57, 32, 34, 48, 42, 58, 38, 72,
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```
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                                 69, 206, 209, 63,
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        169, 223, 107, 275, 165, 80, 179, 111, 244, 127, 144, 130, 181,
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        135, 105, 72, 73, 174, 142, 70, 242, 121, 166, 177, 214, 118,
        206, 158, 240, 97, 122, 238, 57, 160, 98, 100, 131, 128,
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```

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 {'medication_usage': array([0, 1], dtype=int64)},
 {'participated_in_free_screening': array([0, 1], dtype=int64)},
 {'heart_attack': array([0, 1], dtype=int64)},
 {'age_group': array([3, 2, 4, 1])},
 {'health_risk_score': array([ 7, 11, 9, 12, 10, 14, 8, 13, 6, 5, 4, 15],
dtype=int64)},
 {'obesity_risk_score': array([0., 0.5, 1.5, 1.])},
 {'stress_to_sleep_ratio': array([0.33497453, 0.53155551, 0.15782338, ..., 0.29
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        113.66666667, 93. , 118.33333333, 109.
        120.66666667, 121. , 101. , 102.66666667,
        115. , 104.66666667, 111.33333333, 88.66666667,
        132.3333333, 83.66666667, 107.33333333, 91.33333333,
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        110.
                  , 107. , 103.66666667, 108.
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        130. , 127. , 93.66666667, 128.66666667,
                 , 127.66666667, 128. , 114.
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94.3333333
        128.33333333, 104.
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 {'triglyceride-hdl-ratio': array([2.10416667, 2.37931034, 2.47826087, ..., 1.0
```

Out[237...

heart_attack

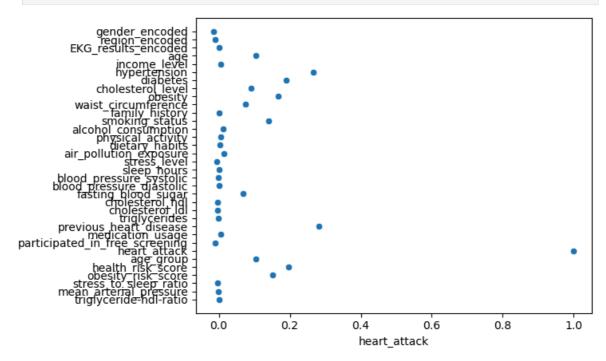
	neart_attack
heart_attack	1.000000
previous_heart_disease	0.281527
hypertension	0.266837
health_risk_score	0.197037
diabetes	0.188859
obesity	0.166300
obesity_risk_score	0.152269
smoking_status	0.141086
age	0.103794
age_group	0.103686
cholesterol_level	0.090968
waist_circumference	0.074709
fasting_blood_sugar	0.068222
air_pollution_exposure	0.013178
alcohol_consumption	0.011258
income_level	0.005869
medication_usage	0.004564
physical_activity	0.004182
dietary_habits	0.003435
triglyceride-hdl-ratio	0.001366
family_history	0.001147
blood_pressure_diastolic	0.000394
EKG_results_encoded	0.000035
sleep_hours	-0.000019
mean_arterial_pressure	-0.001625
blood_pressure_systolic	-0.001841
triglycerides	-0.002392
cholesterol_ldl	-0.004101
cholesterol_hdl	-0.004503
stress_to_sleep_ratio	-0.004751
stress_level	-0.006581
participated_in_free_screening	-0.009895
region_encoded	-0.010467

heart attack

gender_encoded

-0.014083

```
In [238... sns.scatterplot(data=df.corr(), x='heart_attack', y=df.columns.tolist())
plt.show()
```



Modelling

```
In [239...
          x = df.drop('heart_attack', axis=1)
          y = df['heart_attack']
          X_train, X_test, y_train, y_test = train_test_split(x, y, test_size=0.3, random_
          # scaler = StandardScaler()
          # X_train_scaled = scaler.fit_transform(X_train)
          # X_test_scaled = scaler.fit_transform(X_test)
          # X train scaled, X train scaled[0]
          SEED = 42
          rf = RandomForestClassifier(
              criterion='entropy',
              max depth=5,
              max_features='sqrt',
              min samples leaf=1,
              min_samples_split=3,
          # param_grid = {
                'classifier__criterion': ['gini', 'entropy', 'log_loss'],
                 'classifier__max_depth': [2,4,5],
          #
                 'classifier__min_samples_split': [2, 3, 4],
                'classifier min samples leaf': [1, 2, 3],
                 'classifier__max_features': ['sqrt', 'log2'],
          rf_pipeline = Pipeline([
              ('scaler', StandardScaler()),
              ('classifier', rf),
          1)
          # grid_search = GridSearchCV(
```

Out[239...

In [240...

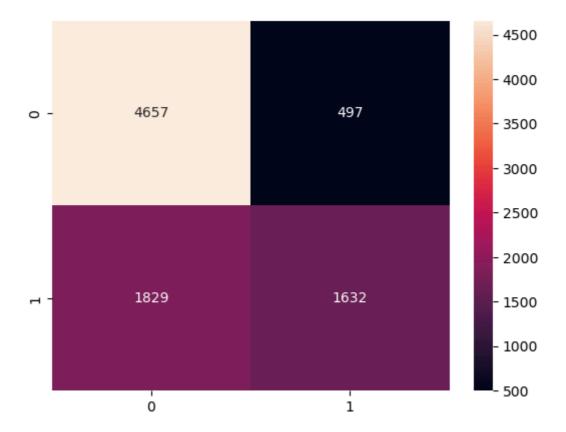
In [242...

```
#
                 estimator = rf_pipeline,
           #
                 param_grid = param_grid,
           #
                 scoring = 'accuracy',
           #
                 n_jobs = None,
           #
                 cv = 5,
           #
                 verbose = 2,
           # )
           # grid_search.fit(X_train, y_train)
           # scaler = StandardScaler()
           # X_train = scaler.fit_transform(X_train)
           # X_test = scaler.transform(X_test)
           # rf.fit(X_train, y_train)
           rf_pipeline.fit(X_train, y_train)
                        Pipeline
                      StandardScaler
                 RandomForestClassifier
           x.columns
           Index(['gender_encoded', 'region_encoded', 'EKG_results_encoded', 'age',
Out[240...
                   'income_level', 'hypertension', 'diabetes', 'cholesterol_level',
                   'obesity', 'waist_circumference', 'family_history', 'smoking_status',
                   'alcohol_consumption', 'physical_activity', 'dietary_habits',
                   'air_pollution_exposure', 'stress_level', 'sleep_hours',
                   'blood_pressure_systolic', 'blood_pressure_diastolic',
                   'fasting_blood_sugar', 'cholesterol_hdl', 'cholesterol_ldl', 'triglycerides', 'previous_heart_disease', 'medication_usage',
                   'participated_in_free_screening', 'age_group', 'health_risk_score',
                   'obesity risk score', 'stress to sleep ratio', 'mean arterial pressure',
                   'triglyceride-hdl-ratio'],
                  dtype='object')
In [241...
           y_pred = rf_pipeline.predict(X_test)
           print(classification_report(y_test,y_pred))
                        precision
                                      recall f1-score
                                                           support
                              0.72
                                         0.90
                     0
                                                   0.80
                                                              5154
                     1
                              0.77
                                         0.47
                                                   0.58
                                                              3461
                                                   0.73
                                                              8615
              accuracy
                              0.74
                                         0.69
                                                   0.69
                                                              8615
             macro avg
         weighted avg
                              0.74
                                        0.73
                                                   0.71
                                                              8615
```

sns.heatmap(confusion matrix(y test,y pred),annot=True, fmt='d')

```
localhost:8888/lab/tree/notebook-dataset/heart_attack_prediction.ipynb
```

plt.show()



Saving model to joblib file

In [243... # joblib.dump(rf_pipeline, 'heart_attack_prediction_model.joblib')

Save clean df as CSV file

```
In [244... df.to_csv('clean_hap.csv', index=False)
In []:
```