heart attack

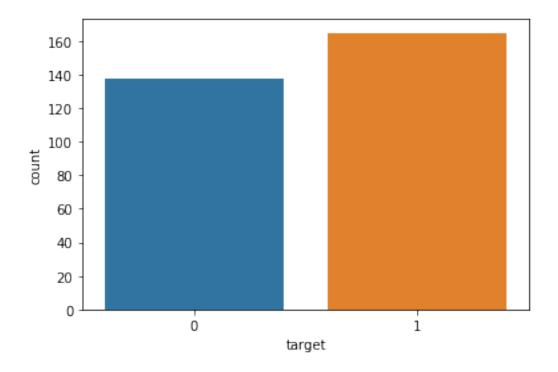
August 6, 2020

```
[1]: import pandas as pd
     import numpy as np
     import matplotlib.pyplot as plt
     import seaborn as sns
 [4]: df = pd.read_csv('datasets_737503_1278636_heart.csv')
     df.head()
 [4]:
        age
              sex
                   ср
                       trestbps
                                   chol
                                         fbs
                                              restecg
                                                        thalach
                                                                  exang
                                                                          oldpeak
                                                                                    slope
     0
         63
                    3
                                   233
                                                     0
                                                                              2.3
                                                                                        0
                1
                             145
                                           1
                                                             150
                                                                      0
     1
         37
                1
                    2
                             130
                                   250
                                           0
                                                     1
                                                             187
                                                                      0
                                                                              3.5
                                                                                        0
     2
                                                                              1.4
                                                                                        2
         41
                             130
                                   204
                                           0
                                                     0
                                                             172
                                                                      0
                0
                    1
                                                                                        2
     3
         56
                    1
                             120
                                   236
                                           0
                                                     1
                                                             178
                                                                              0.8
                1
                                                                      0
     4
         57
                0
                    0
                                   354
                                                     1
                                                                              0.6
                             120
                                           0
                                                             163
            thal
                   target
        ca
     0
         0
                1
                         1
         0
                2
                         1
     1
     2
                2
                         1
         0
     3
         0
                2
                         1
     4
         0
                         1
[14]: df.shape
[14]: (303, 14)
 [8]: df.info()
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 303 entries, 0 to 302
    Data columns (total 14 columns):
                 303 non-null int64
    age
                 303 non-null int64
    sex
                 303 non-null int64
    ср
    trestbps
                 303 non-null int64
                 303 non-null int64
    chol
    fbs
                 303 non-null int64
    restecg
                 303 non-null int64
                 303 non-null int64
    thalach
```

```
exang 303 non-null int64
oldpeak 303 non-null float64
slope 303 non-null int64
ca 303 non-null int64
thal 303 non-null int64
target 303 non-null int64
dtypes: float64(1), int64(13)
memory usage: 33.2 KB
```

[6]: sns.countplot(df['target'])

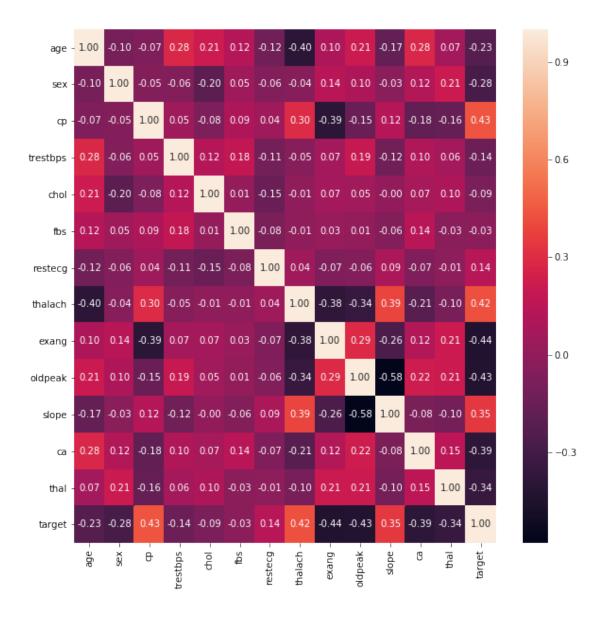
[6]: <matplotlib.axes._subplots.AxesSubplot at 0x7f8f583b9518>



```
[7]: ### data looks good and clean

[10]: plt.figure(figsize = (10, 10))
sns.heatmap(df.corr(), annot = True, fmt = '0.2f')
```

[10]: <matplotlib.axes._subplots.AxesSubplot at 0x7f8f5a6c5898>



```
[11]: ### not obvious correlation
[12]: y = df['target']
X = df.drop('target', axis=1)
from sklearn.model_selection import train_test_split

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.20, \_\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{\text{
```

```
X_test = scaler.transform(X_test)
[16]: from sklearn.model_selection import RepeatedStratifiedKFold
     from sklearn.model_selection import GridSearchCV
     from sklearn.linear_model import LogisticRegression
     model = LogisticRegression()
     c_{values} = [100, 10, 1, 0.1, 0.01]
     penalty = ['12']
     solver = ['newton-cg', 'lbfgs', 'liblinear']
     param = dict(solver = solver, penalty = penalty, C = c_values)
     lg_grid = GridSearchCV(estimator=model, param_grid=param, n_jobs=-1, cv=10,__
      ⇔scoring='accuracy',error_score=0)
     lg_grid.fit(X_train, y_train)
[16]: GridSearchCV(cv=10, error_score=0, estimator=LogisticRegression(), n_jobs=-1,
                  param_grid={'C': [100, 10, 1, 0.1, 0.01], 'penalty': ['12'],
                              'solver': ['newton-cg', 'lbfgs', 'liblinear']},
                  scoring='accuracy')
[17]: | lg_grid.best_estimator_
[17]: LogisticRegression(C=0.01, solver='liblinear')
[18]: log = LogisticRegression(C=0.01, solver='liblinear', penalty = '12')
     log.fit(X_train, y_train)
     y_pred = log.predict(X_test)
[19]: from sklearn.metrics import
      →confusion_matrix,accuracy_score,classification_report
     from sklearn.metrics import roc_auc_score,roc_curve,scorer, auc
     from sklearn.metrics import precision score, recall score, f1 score
    /Users/shijiecai/anaconda3/lib/python3.7/site-
    packages/sklearn/utils/deprecation.py:143: FutureWarning: The
    sklearn.metrics.scorer module is deprecated in version 0.22 and will be removed
    in version 0.24. The corresponding classes / functions should instead be
    imported from sklearn.metrics. Anything that cannot be imported from
    sklearn.metrics is now part of the private API.
      warnings.warn(message, FutureWarning)
[20]: print(classification_report(y_test, y_pred))
                  precision
                               recall f1-score
                                                   support
               0
                       0.84
                                 0.78
                                            0.81
                                                        27
               1
                       0.83
                                 0.88
                                            0.86
                                                        34
```

```
accuracy 0.84 61
macro avg 0.84 0.83 0.83 61
weighted avg 0.84 0.84 0.84 61
```

```
[21]: ### since target is boolean, try bernoulli naive bayes
from sklearn.naive_bayes import BernoulliNB

clf = BernoulliNB()
clf.fit(X_train, y_train)
y_pred = clf.predict(X_test)

print(classification_report(y_test, y_pred))
```

```
precision
                            recall f1-score
                                                 support
           0
                    0.85
                               0.85
                                         0.85
                                                      27
           1
                    0.88
                               0.88
                                         0.88
                                                      34
                                         0.87
                                                      61
    accuracy
                                                      61
   macro avg
                    0.87
                               0.87
                                         0.87
weighted avg
                    0.87
                               0.87
                                         0.87
                                                      61
```

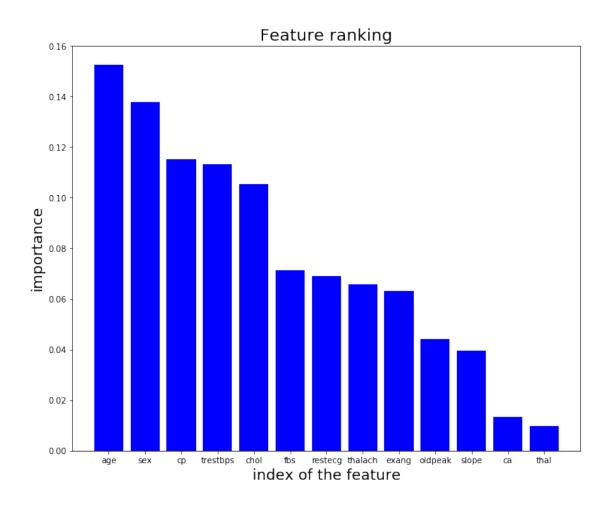
[24]: rf_grid.best_params_

```
precision
                            recall f1-score
                                                support
           0
                    0.88
                              0.78
                                         0.82
                                                      27
           1
                    0.84
                              0.91
                                         0.87
                                                      34
                                         0.85
                                                      61
    accuracy
                              0.84
                                         0.85
                                                      61
   macro avg
                    0.86
weighted avg
                    0.85
                              0.85
                                         0.85
                                                      61
```

```
[27]: importance = rf.feature_importances_
   indices = np.argsort(importance)[::-1]
   feature_names = X.columns

f, ax = plt.subplots(figsize=(11, 9))
   plt.title("Feature ranking", fontsize = 20)
   plt.bar(range(X.shape[1]), importance[indices],
        color="b",
        align="center")
   plt.xticks(range(X.shape[1]), feature_names)
   plt.xlim([-1, X.shape[1]])
   plt.ylabel("importance", fontsize = 18)
   plt.xlabel("index of the feature", fontsize = 18)
```

[27]: Text(0.5, 0, 'index of the feature')



```
[35]: import eli5
     from eli5.sklearn import PermutationImportance
     model = RandomForestClassifier(**param).fit(X_train, y_train)
     perm = PermutationImportance(model, random_state=1).fit(X_test, y_test)
     eli5.show_weights(perm, feature_names = X.columns.tolist())
[35]: <IPython.core.display.HTML object>
[37]: # how come 2 methods show us largely difference result 'thal' and 'age'
[38]:
    df.head()
[38]:
                       trestbps
                                                        thalach
                                                                         oldpeak
        age
                   ср
                                  chol
                                         fbs
                                              restecg
                                                                  exang
                                                                                   slope
             sex
     0
         63
                1
                    3
                             145
                                   233
                                           1
                                                     0
                                                            150
                                                                      0
                                                                              2.3
                                                                                       0
         37
                    2
     1
                1
                             130
                                   250
                                           0
                                                     1
                                                            187
                                                                      0
                                                                              3.5
                                                                                       0
                                                                                       2
     2
         41
                0
                    1
                             130
                                   204
                                           0
                                                     0
                                                            172
                                                                      0
                                                                              1.4
     3
                             120
                                   236
                                           0
                                                     1
                                                            178
                                                                              0.8
                                                                                        2
         56
                1
                    1
                                                                      0
                                                     1
                                                                                        2
         57
                0
                    0
                             120
                                   354
                                           0
                                                            163
                                                                      1
                                                                              0.6
        ca
            thal
                   target
         0
                1
     0
```

```
1
         0
               2
                        1
     2
               2
         0
                        1
     3
         0
                2
                        1
     4
                2
                        1
[39]: | ## Guessing: variables such as cp, restecq, exang, thal, etc, should be
      →categorical variables not numeric values.
[41]: df.columns = ['age', 'sex', 'chest_pain_type', 'resting_blood_pressure', __
      _{\rightarrow}'cholesterol', 'fasting_blood_sugar', 'rest_ecg', 'max_heart_rate_achieved',
             'exercise_induced_angina', 'st_depression', 'st_slope',
      →'num_major_vessels', 'thalassemia', 'target']
     df.head()
[41]:
                   chest_pain_type resting_blood_pressure
                                                              cholesterol
        age
             sex
         63
                                                         145
                                                                       233
               1
                                 3
         37
                                  2
                                                         130
                                                                       250
     1
               1
     2
         41
               0
                                 1
                                                         130
                                                                       204
     3
         56
               1
                                  1
                                                         120
                                                                       236
     4
         57
               0
                                                         120
                                  0
                                                                       354
        fasting_blood_sugar rest_ecg max_heart_rate_achieved \
     0
                           1
                                      0
                                                               150
     1
                           0
                                      1
                                                               187
                           0
                                      0
     2
                                                               172
     3
                           0
                                      1
                                                               178
                           0
                                      1
                                                              163
        exercise_induced_angina st_depression st_slope
                                                             num_major_vessels
     0
                               0
                                             2.3
                                                          0
                               0
                                             3.5
                                                          0
                                                                              0
     1
                                                          2
     2
                                0
                                             1.4
                                                                               0
                                                          2
     3
                                0
                                             0.8
                                                                               0
     4
                                1
                                             0.6
                                                          2
                                                                               0
        thalassemia target
     0
                   1
                           1
                   2
     1
                           1
     2
                   2
                           1
                   2
     3
                           1
                   2
                           1
[42]: df['sex'][df['sex'] == 0] = 'female'
     df['sex'][df['sex'] == 1] = 'male'
     df['chest_pain_type'][df['chest_pain_type'] == 0] = 'typical angina'
     df['chest_pain_type'][df['chest_pain_type'] == 1] = 'atypical angina'
     df['chest_pain_type'][df['chest_pain_type'] == 2] = 'non-anginal pain'
     df['chest_pain_type'][df['chest_pain_type'] == 3] = 'asymptomatic'
```

```
df['fasting blood_sugar'][df['fasting_blood_sugar'] == 0] = 'lower than 120mg/
df['fasting_blood_sugar'][df['fasting_blood_sugar'] == 1] = 'greater than 120mg/
df['rest_ecg'][df['rest_ecg'] == 0] = 'normal'
df['rest_ecg'][df['rest_ecg'] == 1] = 'ST-T wave abnormality'
df['rest_ecg'][df['rest_ecg'] == 2] = 'left ventricular hypertrophy'
df['exercise_induced_angina'][df['exercise_induced_angina'] == 0] = 'no'
df['exercise induced angina'][df['exercise induced angina'] == 1] = 'yes'
df['st_slope'][df['st_slope'] == 0] = 'upsloping'
df['st_slope'][df['st_slope'] == 1] = 'flat'
df['st_slope'][df['st_slope'] == 2] = 'downsloping'
df['thalassemia'][df['thalassemia'] == 0] = 'normal'
df['thalassemia'][df['thalassemia'] == 1] = 'fixed defect'
df['thalassemia'][df['thalassemia'] == 2] = 'reversable defect'
/Users/shijiecai/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:1:
SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: http://pandas.pydata.org/pandas-
docs/stable/indexing.html#indexing-view-versus-copy
  """Entry point for launching an IPython kernel.
/Users/shijiecai/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:4:
SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: http://pandas.pydata.org/pandas-
docs/stable/indexing.html#indexing-view-versus-copy
  after removing the cwd from sys.path.
/Users/shijiecai/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:9:
SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: http://pandas.pydata.org/pandas-
docs/stable/indexing.html#indexing-view-versus-copy
  if __name__ == '__main__':
/Users/shijiecai/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:10:
SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: http://pandas.pydata.org/pandas-
```

docs/stable/indexing.html#indexing-view-versus-copy

Remove the CWD from sys.path while we load stuff.

/Users/shijiecai/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:12: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy

if sys.path[0] == '':

/Users/shijiecai/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:16: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy

app.launch_new_instance()

/Users/shijiecai/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:17: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy

/Users/shijiecai/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:19: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy

/Users/shijiecai/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:23: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy

/Users/shijiecai/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:24: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy

/Users/shijiecai/anaconda3/lib/python3.7/site-packages/ipykernel_launcher.py:25: SettingWithCopyWarning:

A value is trying to be set on a copy of a slice from a DataFrame

See the caveats in the documentation: http://pandas.pydata.org/pandas-docs/stable/indexing.html#indexing-view-versus-copy

```
[43]: df.head()
[43]:
        age
                 sex
                       chest_pain_type resting_blood_pressure
                                                                   cholesterol
     0
         63
               male
                          asymptomatic
                                                              145
                                                                           233
     1
         37
               male
                     non-anginal pain
                                                             130
                                                                           250
     2
         41
                       atypical angina
                                                                           204
             female
                                                             130
     3
         56
                       atypical angina
                                                             120
                                                                           236
               male
                        typical angina
     4
         57
             female
                                                             120
                                                                           354
          fasting_blood_sugar
                                                        max_heart_rate_achieved
                                              rest_ecg
        greater than 120mg/ml
                                                normal
                                                                              150
     0
                                                                              187
          lower than 120mg/ml
     1
                                ST-T wave abnormality
     2
          lower than 120mg/ml
                                                 normal
                                                                              172
     3
          lower than 120mg/ml
                                ST-T wave abnormality
                                                                              178
     4
          lower than 120mg/ml
                                ST-T wave abnormality
                                                                              163
       exercise_induced_angina
                                  st_depression
                                                     st_slope
                                                               num_major_vessels
     0
                                            2.3
                                                    upsloping
                                            3.5
                                                                                0
     1
                             nο
                                                    upsloping
     2
                                            1.4 downsloping
                                                                                0
                             no
     3
                                            0.8
                                                  downsloping
                                                                                0
                             no
     4
                                                                                 0
                                                  downsloping
                                            0.6
                            yes
              thalassemia
                           target
             fixed defect
     0
                                  1
       reversable defect
     1
                                  1
     2 reversable defect
                                  1
     3 reversable defect
                                  1
     4 reversable defect
                                  1
[44]: df.dtypes
[44]: age
                                    int64
                                   object
     sex
     chest_pain_type
                                   object
     resting_blood_pressure
                                    int64
     cholesterol
                                    int64
     fasting_blood_sugar
                                   object
     rest_ecg
                                   object
     max_heart_rate_achieved
                                    int64
     exercise_induced_angina
                                   object
     st_depression
                                  float64
     st_slope
                                   object
     num_major_vessels
                                    int64
     thalassemia
                                   object
     target
                                    int64
     dtype: object
```

```
[46]: df = pd.get_dummies(df, drop_first=True)
     df.head()
[46]:
        age
             resting_blood_pressure cholesterol max_heart_rate_achieved
         63
                                  145
                                                233
                                                                           150
         37
                                                250
     1
                                  130
                                                                           187
     2
         41
                                  130
                                                204
                                                                           172
                                                236
     3
         56
                                  120
                                                                           178
         57
                                                354
                                  120
                                                                           163
        st_depression num_major_vessels target
                                                     sex_male
     0
                   2.3
                                                  1
     1
                   3.5
                                          0
                                                  1
                                                             1
     2
                   1.4
                                          0
                                                  1
                                                             0
     3
                   0.8
                                          0
                                                  1
                                                             1
                   0.6
     4
                                                             0
        chest_pain_type_atypical angina chest_pain_type_non-anginal pain
     0
                                        0
                                                                             1
     1
     2
                                        1
                                                                             0
     3
                                         1
                                                                             0
     4
                                         0
                                                                             0
        chest_pain_type_typical angina fasting_blood_sugar_lower than 120mg/ml
     0
                                       0
     1
                                                                                    1
     2
                                       0
                                                                                    1
     3
                                       0
                                                                                    1
     4
                                                                                    1
        rest_ecg_left ventricular hypertrophy rest_ecg_normal
     0
                                               0
                                                                 1
                                               0
                                                                 0
     1
     2
                                               0
                                                                 1
     3
                                               0
                                                                 0
                                               0
        exercise_induced_angina_yes
                                       st_slope_flat st_slope_upsloping
     0
     1
                                    0
                                                    0
                                                                          1
     2
                                    0
                                                    0
                                                                          0
     3
                                    0
                                                    0
                                                                          0
     4
                                                    0
                                                                          0
                                                         thalassemia_reversable defect
        thalassemia_fixed defect thalassemia_normal
     0
```

```
1
                                0
                                                    0
                                                                                    1
     2
                                0
                                                    0
                                                                                     1
     3
                                0
                                                    0
                                                                                     1
     4
                                0
                                                                                     1
[47]: y = df['target']
     X = df.drop('target', axis=1)
     X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.20, u)
     →random_state = 0)
     scaler = StandardScaler()
     X_train = scaler.fit_transform(X_train)
     X_test = scaler.transform(X_test)
[48]: model = LogisticRegression()
     c_values = [100, 10, 1, 0.1, 0.01]
     penalty = ['12']
     solver = ['newton-cg', 'lbfgs', 'liblinear']
     param = dict(solver = solver, penalty = penalty, C = c_values)
     lg_grid = GridSearchCV(estimator=model, param_grid=param, n_jobs=-1, cv=10, ___

→scoring='accuracy',error_score=0)
     lg_grid.fit(X_train, y_train)
[48]: GridSearchCV(cv=10, error_score=0, estimator=LogisticRegression(), n_jobs=-1,
                  param_grid={'C': [100, 10, 1, 0.1, 0.01], 'penalty': ['12'],
                               'solver': ['newton-cg', 'lbfgs', 'liblinear']},
                  scoring='accuracy')
[49]: lg_grid.best_estimator_
[49]: LogisticRegression(C=0.01, solver='newton-cg')
[50]: log = LogisticRegression(C = 0.01, solver = 'newton-cg', penalty = '12')
     log.fit(X_train, y_train)
     y_pred = log.predict(X_test)
     print(classification_report(y_test, y_pred))
                  precision
                                recall f1-score
                                                    support
               0
                        0.84
                                  0.78
                                            0.81
                                                         27
               1
                        0.83
                                  0.88
                                            0.86
                                                         34
                                            0.84
                                                         61
        accuracy
       macro avg
                        0.84
                                  0.83
                                            0.83
                                                         61
                        0.84
                                  0.84
                                            0.84
                                                         61
    weighted avg
```

```
[51]: clf = BernoulliNB()
     clf.fit(X_train, y_train)
     y_pred = clf.predict(X_test)
    print(classification_report(y_test, y_pred))
                  precision
                               recall f1-score
                                                   support
               0
                       0.85
                                 0.85
                                            0.85
                                                        27
                       0.88
                                 0.88
                                            0.88
                                                        34
                                           0.87
                                                        61
        accuracy
       macro avg
                       0.87
                                 0.87
                                           0.87
                                                        61
    weighted avg
                       0.87
                                 0.87
                                            0.87
                                                        61
[52]: rf = RandomForestClassifier()
     n_{estimators} = [100, 200, 300, 400]
     max_depth = [4, 5, 6, 7]
    min_samples_split = [2, 3, 4, 5]
     max_features = ['auto', 'sqrt', 'log2']
     params = dict(n_estimators = n_estimators, max_depth=max_depth,__

-max_features=max_features, min_samples_split=min_samples_split)

     rf_grid = GridSearchCV(estimator=rf, param_grid=params, n_jobs=-1, cv=5,__
     →scoring = 'accuracy')
     rf_grid.fit(X_train, y_train)
[52]: GridSearchCV(cv=5, estimator=RandomForestClassifier(), n_jobs=-1,
                  param_grid={'max_depth': [4, 5, 6, 7],
                              'max_features': ['auto', 'sqrt', 'log2'],
                              'min_samples_split': [2, 3, 4, 5],
                              'n_estimators': [100, 200, 300, 400]},
                  scoring='accuracy')
[53]: rf_grid.best_params_
[53]: {'max_depth': 6,
      'max features': 'log2',
      'min_samples_split': 2,
      'n estimators': 100}
[54]: param = {'n_estimators':100, 'max_depth':6, 'max_features':'log2', |
     rf = RandomForestClassifier(**param)
     rf.fit(X_train, y_train)
     y_pred = rf.predict(X_test)
```

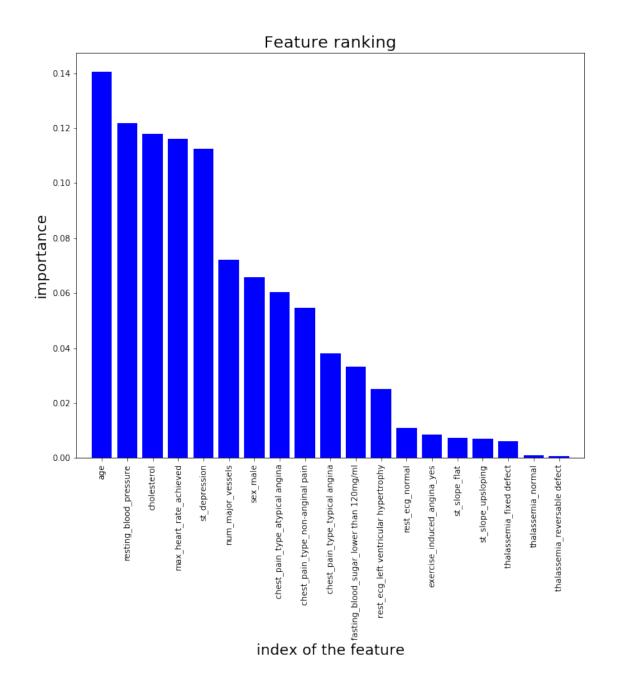
```
print(classification_report(y_test, y_pred))
```

```
precision
                           recall f1-score
                                               support
                   0.84
                             0.78
           0
                                        0.81
                                                    27
           1
                   0.83
                             0.88
                                        0.86
                                                    34
                                        0.84
    accuracy
                                                    61
  macro avg
                   0.84
                             0.83
                                        0.83
                                                    61
weighted avg
                   0.84
                             0.84
                                        0.84
                                                    61
```

```
[57]: importance = rf.feature_importances_
   indices = np.argsort(importance)[::-1]
   feature_names = X.columns

f, ax = plt.subplots(figsize=(11, 9))
   plt.title("Feature ranking", fontsize = 20)
   plt.bar(range(X.shape[1]), importance[indices],
        color="b",
        align="center")
   plt.xticks(range(X.shape[1]), feature_names, rotation = 90)
   plt.xlim([-1, X.shape[1]])
   plt.ylabel("importance", fontsize = 18)
   plt.xlabel("index of the feature", fontsize = 18)
```

[57]: Text(0.5, 0, 'index of the feature')



```
[58]: model = RandomForestClassifier(**param).fit(X_train, y_train)
perm = PermutationImportance(model, random_state=1).fit(X_test, y_test)
eli5.show_weights(perm, feature_names = X.columns.tolist())
```

[58]: <IPython.core.display.HTML object>