```
In [29]:
         import numpy as np
         import pandas as pd
         data=pd.read_csv('har_activity1.csv')
         print(data.head())
         print(data.info())
              user gender age how_tall_in_meters weight body_mass_index x1
                                                                                    y1
         \
            debora
                    Woman
                             46
                                               1.62
                                                         75
                                                                         28.6
                                                                                    97
            debora
                    Woman
                             46
                                               1.62
                                                         75
                                                                         28.6
                                                                                1
                                                                                   100
         2
            debora
                    Woman
                             46
                                               1.62
                                                         75
                                                                         28.6
                                                                               -2
                                                                                   102
            debora
                    Woman
                             46
                                               1.62
                                                         75
                                                                         28.6
                                                                                   102
           debora
                    Woman
                             46
                                               1.62
                                                         75
                                                                         28.6 -3
                                                                                    97
                                                  у4
                                                              class
            z1 x2
                    y2 z2 x3
                                  y3 z3
                                           х4
                                                       z4
         0 -61 -14
                    19 -16 -13
                                 105 -89 -148 -102.0 -146
                                                           sitting
         1 -63 -10
                    22 -13 -14
                                 104 -91 -177
                                               -99.0 -133
                                                            sitting
                                               -99.0 -150
         2 -73 -13
                    19 -17 -13
                                 104 -90 -158
                                                            sitting
         3 -67 -10
                    21 -13 -13
                                 104 -91 -156 -101.0 -147
                                                            sitting
                    18 -17 -13
         4 -66 -15
                                 104 -90 -157
                                               -99.0 -148
                                                            sitting
         <class 'pandas.core.frame.DataFrame'>
         RangeIndex: 9998 entries, 0 to 9997
         Data columns (total 19 columns):
                                9998 non-null object
         user
                                9998 non-null object
         gender
                                9998 non-null int64
         age
         how_tall_in_meters
                                9998 non-null float64
         weight
                                9998 non-null int64
         body mass index
                                9998 non-null float64
                                9998 non-null int64
         x1
         у1
                                9998 non-null int64
         z1
                                9998 non-null int64
         x2
                                9998 non-null int64
         y2
                                9998 non-null int64
                                9998 non-null int64
         z2
                                9998 non-null int64
         х3
                                9998 non-null int64
         у3
                                9998 non-null int64
         z3
         х4
                                9998 non-null int64
                                9997 non-null float64
         ٧4
         z4
                                9998 non-null int64
         class
                                9998 non-null object
         dtypes: float64(3), int64(13), object(3)
         memory usage: 1.4+ MB
```

None

```
In [30]:
         from sklearn import preprocessing
         # encode class values as integers
         le = preprocessing.LabelEncoder() #create an encoder instance
         le.fit(data['class'].unique())
                                                     #fit the encoder with two string "n
         o" and "yes"
         data['class']=le.transform(data['class']) #to transform df['driveway'] valu
         e of yes and no to 1 and 0
         data=data.dropna(axis='rows')
         print(data.head())
                           age how_tall_in_meters weight body_mass_index x1
              user gender
                                                                                   у1
         \
            debora
                    Woman
                            46
                                               1.62
                                                         75
                                                                        28.6
                                                                                   97
                                                         75
                                                                        28.6
                                                                               1
            debora
                    Woman
                            46
                                               1.62
                                                                                  100
                                                         75
         2 debora
                    Woman
                            46
                                               1.62
                                                                        28.6
                                                                             -2
                                                                                  102
            debora
                                                         75
                                                                        28.6
                                                                               0
                                                                                  102
         3
                    Woman
                            46
                                               1.62
                                               1.62
                                                         75
                                                                        28.6
                                                                                   97
           debora
                    Woman
                            46
                                                                             - 3
                                                       z4 class
            z1 x2
                    y2 z2 x3
                                 y3 z3
                                          х4
                                                 y4
         0 -61 -14
                    19 -16 -13
                                105 -89 -148 -102.0 -146
                                                               0
         1 -63 -10
                    22 -13 -14
                                104 -91 -177
                                              -99.0 -133
                                                               0
                    19 -17 -13
         2 -73 -13
                                104 -90 -158
                                              -99.0 -150
                                                               0
         3 -67 -10
                    21 -13 -13
                                104 -91 -156 -101.0 -147
                                                               0
         4 -66 -15
                    18 -17 -13
                                104 -90 -157 -99.0 -148
                                                               0
In [31]:
         le2 = preprocessing.LabelEncoder() #create an encoder instance
         le2.fit(data['gender'].unique())
                                                      #fit the encoder with two string
          "no" and "yes"
         data['gender']=le2.transform(data['gender']) #to transform df['driveway'] v
         alue of yes and no to 1 and 0
         print(data.head())
              user
                    gender
                            age how_tall_in_meters weight body_mass_index
                                                                                    у1
                                                                               х1
         ١
            debora
                             46
                                                1.62
                                                          75
                                                                         28.6
                                                                                    97
                         1
            debora
                                                1.62
                                                          75
                                                                         28.6
                                                                                1
                                                                                   100
         1
                         1
                             46
         2
            debora
                         1
                             46
                                                1.62
                                                          75
                                                                         28.6
                                                                               -2
                                                                                   102
         3 debora
                             46
                                                1.62
                                                          75
                                                                         28.6
                                                                                   102
                                                                                    97
                         1
                             46
                                                1.62
                                                          75
                                                                         28.6 -3
           debora
                                                          class
            z1 x2
                    y2 z2 x3
                                 y3 z3
                                          х4
                                                 y4
                                                       z4
         0 -61 -14
                    19 -16 -13
                                105 -89 -148 -102.0 -146
                                                               0
         1 -63 -10
                    22 -13 -14
                                104 -91 -177
                                              -99.0 -133
                                                               0
         2 -73 -13
                    19 -17 -13
                                104 -90 -158
                                              -99.0 -150
                                                               0
                    21 -13 -13
                                104 -91 -156 -101.0 -147
                                                               0
         3 -67 -10
         4 -66 -15
                    18 -17 -13
                                104 - 90 - 157
                                              -99.0 -148
                                                               0
```

```
In [46]:
         import matplotlib.pyplot as plt
         from sklearn.model selection import train test split
         from sklearn.linear model import SGDClassifier
         from sklearn.svm import LinearSVC
         from sklearn.metrics import accuracy score
         #classifier=SGDClassifier(max_iter=1000)
         clf = LinearSVC(random state=42, tol=1e-3)
         #prepare data which is step 3
         Y = data['class'] #price is the target, Y
         X = np.array([data['body_mass_index'],data['x1'],data['y1'],data['z1'],data['x
         2'],data['y2'],data['z2'],data['x3'],data['y3'],data['z3'],data['x4'],data['y
         4'],data['z4']]) # X
         X=X.T
         xtrain,xtest,ytrain,ytest=train test split(X,Y,random state=43,test size=0.2)
         classifier.fit(xtrain,ytrain)
         y_pred=classifier.predict(xtest)
         print(accuracy_score(ytest,y_pred)) #percentage of classification on the test
          data is correct
```

0.789

```
In [47]: from sklearn.metrics import confusion_matrix
import seaborn as sns

mat2=confusion_matrix(ytest,y_pred)
sns.heatmap(mat2,square=True,annot=True,cbar=False,fmt='d')
plt.xlabel('predicted value')
plt.ylabel('true value')
```

Out[47]: Text(91.68,0.5, 'true value')

