

Felix_prototype_V0.2

November 11, 2018

1 Felix prototype

Version 0.2

Date 10/11/2018

Model used : **Random Forest** Classifier on features selected using various techniques including **lasso**

Clustering method used : **Hierarchical clustering** using **ward metric** based on 6 **NOT variable**

```
In [1]: from pathlib import Path
import pandas as pd
import numpy as np
from datetime import datetime
import time
import matplotlib.pyplot as plt
%matplotlib inline
import itertools
import pickle
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.model_selection import cross_val_score, GridSearchCV
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import confusion_matrix, f1_score, precision_score, recall_score
from sklearn.model_selection import StratifiedKFold
from sklearn.utils import resample

In [2]: path_project = Path.home() / Path('Google Drive/Felix')
path_fig = path_project / Path("fig")
path_data = path_project / Path("data")
path_dump = path_project / Path("dump")

In [3]: # loading data
file = path_data / Path("dataset.csv")
with Path.open(file, 'rb') as fp:
    dataset = pd.read_csv(fp, encoding='utf-8', low_memory=False, index_col = 0)
```

1.0.1 Features scope and selection strategy

Features are selected using lasso on the full scope of feature. The 50 more important features (logistic regression coef ranking) are kept regardless of their activability

```
In [4]: # load feature sets
filename = path_dump / Path("dict_features_sets.sav")
with open(filename, 'rb') as fp:
    dict_features_sets = pickle.load(fp)
```

```
In [5]: [k for k in dict_features_sets.keys()]
```

```
Out[5]: ['all_features',
'cdv_features',
'additional_features',
'insee_features',
'insee_demographics_features',
'insee_recreation_features',
'cat_features',
'quant_features',
'non_redundant_cdv_features',
'cdv_actionable_individual_1_features',
'cdv_actionable_individual_2_features',
'cdv_actionable_individual_3_features',
'cdv_actionable_individual_4_features',
'cdv_actionable_admin_1_features',
'cdv_actionable_admin_2_features',
'cdv_actionable_admin_3_features',
'cdv_actionable_admin_4_features',
'cdv_actionable_admin_5_features',
'insee_recreation_actionable_admin_1_features',
'insee_recreation_actionable_admin_3_features',
'insee_recreation_actionable_admin_4_features',
'insee_demographics_actionable_admin_1_features',
'insee_demographics_actionable_admin_2_features',
'insee_demographics_actionable_admin_3_features',
'insee_demographics_actionable_admin_4_features',
'insee_demographics_actionable_admin_5_features',
'insee_environment_score_features',
'insee_recreation_score_features',
'usual_common_scope_features',
'usual_synthetic_scope_features',
'RFE_LinearSVC_100_features',
'RFE_LinearSVC_50_features',
'RFE_LinearSVC_20_features',
'RFE_LinearSVC_10_features',
'RFE_RandomForestClassifier_100_features',
'RFE_RandomForestClassifier_50_features',
'RFE_RandomForestClassifier_20_features',
'RFE_RandomForestClassifier_10_features',
'RFE_LogisticRegression_100_features',
'RFE_LogisticRegression_50_features',
'RFE_LogisticRegression_20_features',
```

```

'RFE_LogisticRegression_10_features',
'SelectFromModel_LinearSCV_features',
'SelectFromModel_LogisticRegression_features']

```

```

In [6]: usual_common_scope_features = dict_features_sets.get('usual_common_scope_features', set())
        # retrieving actionable features from disk
        cdv_actionable_individual_1_features = dict_features_sets.get('cdv_actionable_individual_1_features')
        cdv_actionable_individual_2_features = dict_features_sets.get('cdv_actionable_individual_2_features')
        cdv_actionable_admin_1_features = dict_features_sets.get('cdv_actionable_admin_1_features')
        cdv_actionable_admin_2_features = dict_features_sets.get('cdv_actionable_admin_2_features')
        insee_recreation_actionable_admin_1_features = dict_features_sets.get('insee_recreation_actionable_admin_1_features')
        insee_recreation_actionable_admin_2_features = dict_features_sets.get('insee_recreation_actionable_admin_2_features')
        insee_environment_actionable_admin_1_features = dict_features_sets.get('insee_environment_actionable_admin_1_features')
        insee_environment_actionable_admin_2_features = dict_features_sets.get('insee_environment_actionable_admin_2_features')
        insee_demographics_actionable_admin_1_features = dict_features_sets.get('insee_demographics_actionable_admin_1_features')
        insee_demographics_actionable_admin_2_features = dict_features_sets.get('insee_demographics_actionable_admin_2_features')
        # defining sets of actionable features
        actionable_individual_1_features = cdv_actionable_individual_1_features
        actionable_individual_2_features = cdv_actionable_individual_2_features
        actionable_admin_1_features = cdv_actionable_admin_1_features | insee_recreation_actionable_admin_1_features
        actionable_admin_2_features = cdv_actionable_admin_2_features | insee_recreation_actionable_admin_2_features

        RFE_LogisticRegression_10_features = dict_features_sets['RFE_LogisticRegression_10_features']
        RFE_LogisticRegression_20_features = dict_features_sets['RFE_LogisticRegression_20_features']
        RFE_LogisticRegression_50_features = dict_features_sets['RFE_LogisticRegression_50_features']
        RFE_LogisticRegression_100_features = dict_features_sets['RFE_LogisticRegression_100_features']

        RFE_RandomForestClassifier_100_features = dict_features_sets['RFE_RandomForestClassifier_100_features']
        RFE_RandomForestClassifier_20_features = dict_features_sets['RFE_RandomForestClassifier_20_features']
        RFE_RandomForestClassifier_50_features = dict_features_sets['RFE_RandomForestClassifier_50_features']
        RFE_RandomForestClassifier_10_features = dict_features_sets['RFE_RandomForestClassifier_10_features']

        RFE_LinearSVC_100_features = dict_features_sets['RFE_LinearSVC_100_features']
        RFE_LinearSVC_50_features = dict_features_sets['RFE_LinearSVC_50_features']
        RFE_LinearSVC_20_features = dict_features_sets['RFE_LinearSVC_20_features']
        RFE_LinearSVC_10_features = dict_features_sets['RFE_LinearSVC_10_features']
        SelectFromModel_LinearSCV_features = dict_features_sets['SelectFromModel_LinearSCV_features']
        SelectFromModel_LogisticRegression_features = dict_features_sets['SelectFromModel_LogisticRegression_features']

In [7]: print(f"The {len(SelectFromModel_LogisticRegression_features)} features obtained using lasso")
        print(f"{len(SelectFromModel_LogisticRegression_features & dict_features_sets.get('cdv_features'))} features obtained using CDV")
        print(list(SelectFromModel_LogisticRegression_features))

```

The 56 features obtained using lasso:

56 issues de l'étude CDV, 0 de l'insee, 0 calculées à partir des données insee

['INQRROUT3_Non inquiet', 'RE_ALIM_Oui', 'INQALIM', 'TRANSFST_Oui', 'ETATSAN', 'SITUEMP3_Inactif']

1.0.2 Clustering method - feature used

Hierarchical clustering is used using 6 common "NOT_" variable

```
In [8]: # loading clustering
        file = path_data / Path("clustTest3.csv")
        with Path.open(file, 'rb') as fp:
            clustTest1 = pd.read_csv(fp, encoding='utf-8', low_memory=False, sep=";", index_col=0)

In [9]: cluster_name = ["", "Civiques", "Equilibrés", "Flâneurs", "Domestiques modérés", "Solitaires"]
```

1.1 Option 1 - Lasso only

```
In [10]: # choosing set of features
         scope = SelectFromModel_LogisticRegression_features
         print(f"number of features : {len(scope)} ow actionable")
         A = scope & dict_features_sets.get('cdv_actionable_admin_1_features', set())
         B = scope & dict_features_sets.get('cdv_actionable_individual_1_features', set())
         print(f"- at administrative level 1 : \t{len(A)}\n- at individual level 1 : \t{len(B)}")

number of features : 56 ow actionable
- at administrative level 1 :          27
- at individual level 1 :          25
```

1.1.1 Training set and test set preparation

```
In [11]: df = dataset.loc[:, :]
         # reducing problem to a 2 class classification problem
         df["HEUREUX_CLF"] = 0
         df.loc[df["HEUREUX"]==4, "HEUREUX_CLF"] = 1
         df.loc[df["HEUREUX"]==3, "HEUREUX_CLF"] = 1
         df.loc[df["HEUREUX"]==5, "HEUREUX_CLF"] = None

         scope = scope & set(dataset.columns)
         n_max = 2000

         df = df.loc[:, scope | {"HEUREUX_CLF"} ].dropna()
         features = df.loc[:, scope ].columns

         X = df.loc[:, scope]
         y = df["HEUREUX_CLF"]

         Xs, ys = resample(X, y, random_state=42)

         Xs = Xs.iloc[0:n_max, :]
         ys = ys.iloc[0:n_max]
```

```
X_train, X_test, y_train, y_test = train_test_split(Xs, ys,
                                                    test_size=0.2,
                                                    random_state=42
                                                    )
```

```
scaler = StandardScaler().fit(X_train)
X_train = scaler.transform(X_train)
X_test = scaler.transform(X_test)
```

```
print(f"Number exemple: {y.shape[0]}\n- training set: \
{y_train.shape[0]}\n- test set: {y_test.shape[0]}")
print(f"Number of features: p={X_train.shape[1]}")
print(f"Number of class: {len(np.unique(y))}")
for c in np.unique(y):
    print(f"class {c:0.0f} : {100*np.sum(y==c)/len(y):0.1f}%")
```

```
Number exemple: 10788
- training set: 1600
- test set: 400
Number of features: p=56
Number of class: 2
class 0 : 35.0%
class 1 : 65.0%
```

1.1.2 Learning and model performance evaluation on full dataset (before clustering)

```
In [12]: startTime = time.time()
n_estimators_range = [32,64,128,256,512]
max_depth_range = [4,8,16,32,64]
param_grid = dict(n_estimators=n_estimators_range, max_depth = max_depth_range)

params = {'max_features' : 'sqrt', 'random_state' : 32,
          'min_samples_split' : 2, 'class_weight' : 'balanced'}
clf = RandomForestClassifier(**params)

grid = GridSearchCV(clf, scoring='accuracy', param_grid=param_grid)
grid.fit(X_train, y_train)
print(f"Determination of optimal hyperparameters in {time.time() - startTime:0.1f} s")
print(f"Optimal values are {grid.best_params_} \n\
Accuracy Score of cross validation {100*grid.best_score_:0.2f}%")

# Learning on full training set with optimals hyperparameters and score on test set
params = {'max_features' : 'sqrt', 'random_state' : 32,
          'min_samples_split' : 2, 'class_weight' : 'balanced',
          'n_estimators' : grid.best_params_['n_estimators'],
          'max_depth' : grid.best_params_['max_depth']}
clf = RandomForestClassifier(**params).fit(X_train, y_train)
```

```

clf.fit(X_train, y_train)
y_test_pred = clf.predict(X_test)

print(f"Random Forest, p={X_train.shape[1]}")
accuracy = clf.score(X_test, y_test)
f1 = f1_score(y_test, y_test_pred)
p = precision_score(y_test, y_test_pred)
r = recall_score(y_test, y_test_pred)
print(f"Model score\n- Accuracy : {accuracy*100:0.1f} %")
print(f"- Precision : {p*100:0.1f} % (Happy # positive class)")
print(f"- Recall : {r*100:0.1f} %")
print(f"- F1 score : {f1*100:0.1f} %")
res_full = {
    'f1_score' : f1,
    'accuracy' : accuracy,
    'precision' : p,
    'recall' : r
}

```

Determination of optimal hyperparameters in 43.2 s
 Optimal values are {'max_depth': 16, 'n_estimators': 256}
 Accuracy Score of cross validation 75.56%
 Random Forest, p=56
 Model score
 - Accuracy : 73.5 %
 - Precision : 73.8 % (Happy # positive class)
 - Recall : 90.1 %
 - F1 score : 81.1 %

```

In [13]: importances = clf.feature_importances_
std = np.std([tree.feature_importances_ for tree in clf.estimators_],
             axis=0)
indices = np.argsort(importances)[::-1]
features_name = np.array(features)
features_short_name_sorted = [ name[:15] for name in features_name[indices]]
n_features_max = 25
n_features = min(X.shape[1], n_features_max)
print("Feature ranking:")

# Plot the feature importances of the forest
plt.figure()
plt.gcf().subplots_adjust(bottom=0.4)
plt.title("Feature importances\nHappiness model before clustering")
plt.bar(range(n_features), importances[indices][:n_features],
        color="r", yerr=std[indices[:n_features]], align="center")
plt.xticks(range(n_features), features_short_name_sorted[:n_features], rotation=90)
plt.xlim([-1, n_features])

```

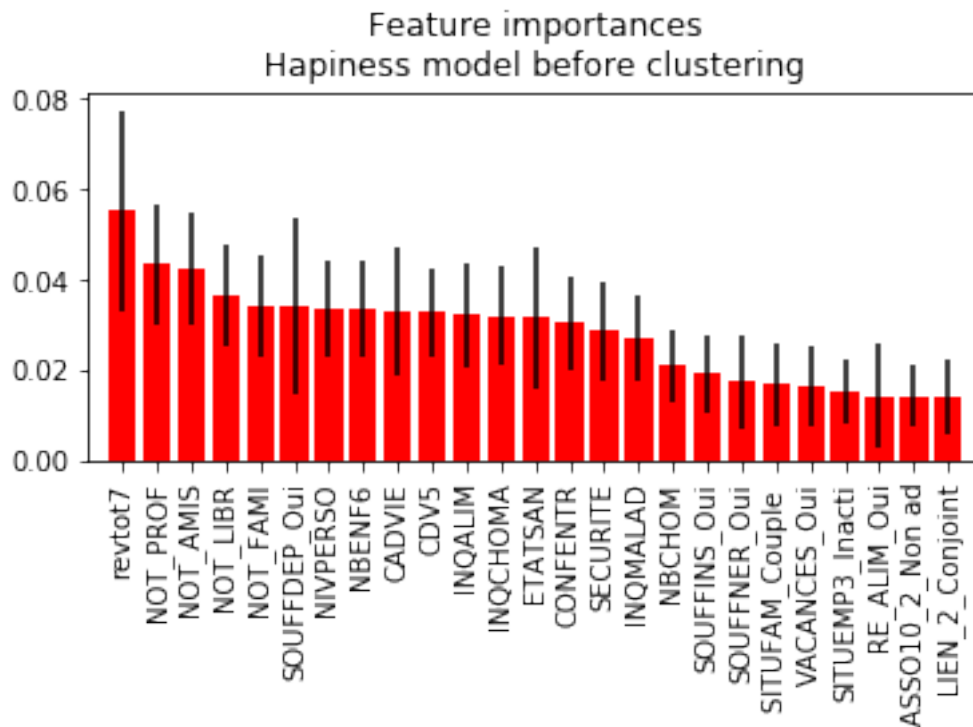
```

filename = path_fig / Path("feature_importance_option1.jpg")
plt.savefig(filename, format='jpg')
plt.show()

for f in range(min(X.shape[1],n_features_max)):
    print("%d. feature %d -%s- (%f)" % (f + 1, indices[f],features_name[indices[f]], im
    if features_name[indices[f]] in actionable_individual_1_features:
        print("\tActionable at individual level (1)")
    if features_name[indices[f]] in actionable_individual_2_features:
        print("\tActionable at individual level (2)")
    if features_name[indices[f]] in actionable_admin_1_features:
        print("\tActionable at administrative level (1)")
    if features_name[indices[f]] in actionable_admin_2_features:
        print("\tActionable at administrative level (2)")

```

Feature ranking:



1. feature 27 -revtot7- (0.055216)
 Actionable at individual level (2)
 Actionable at administrative level (2)
2. feature 36 -NOT_PROF- (0.043599)
 Actionable at individual level (1)
 Actionable at administrative level (2)

3. feature 49 -NOT_AMIS- (0.042551)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
4. feature 34 -NOT_LIBR- (0.036881)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
5. feature 16 -NOT_FAMI- (0.034543)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
6. feature 48 -SOUFFDEP_Oui- (0.034192)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
7. feature 37 -NIVPERSO- (0.033697)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
8. feature 54 -NBENF6- (0.033558)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
9. feature 19 -CADVIE- (0.032957)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
10. feature 8 -CDV5- (0.032856)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
11. feature 2 -INQALIM- (0.032343)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
12. feature 26 -INQCHOMA- (0.032206)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
13. feature 4 -ETATSAN- (0.031754)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
14. feature 28 -CONFENTR- (0.030514)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
15. feature 21 -SECURITE- (0.028726)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
16. feature 44 -INQMALAD- (0.027184)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
17. feature 45 -NBCHOM- (0.021083)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
18. feature 15 -SOUFFINS_Oui- (0.019343)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)


```

19. feature 53 -SOUFFNER_Oui- (0.017693)
    Actionable at individual level (2)
    Actionable at administrative level (1)
20. feature 42 -SITUFAM_Couple sans enfants- (0.016903)
    Actionable at individual level (2)
21. feature 35 -VACANCES_Oui- (0.016708)
    Actionable at individual level (1)
    Actionable at administrative level (1)
22. feature 5 -SITUEMP3_Inactif- (0.015247)
    Actionable at individual level (2)
    Actionable at administrative level (2)
23. feature 1 -RE_ALIM_Oui- (0.014443)
    Actionable at individual level (2)
    Actionable at administrative level (2)
24. feature 46 -ASS010_2_Non adhérent- (0.014405)
    Actionable at individual level (1)
    Actionable at administrative level (1)
25. feature 10 -LIEN_2_Conjoint ou compagnon- (0.014060)

```

```

In [14]: print(f"number of features : {len(scope)} ow actionnable")
        A = scope & dict_features_sets.get('cdv_actionable_admin_1_features',set())
        B = scope & dict_features_sets.get('cdv_actionable_individual_1_features',set())
        print(f"- at administrative level 1 : \t{len(A)}\n- at individual level 1 : \t{len(B)}")
        important_features = set(features_name[indices][:10])
        C = A & important_features
        D = B & important_features
        print(f"- at administrative level 1 in top 10: \t{len(C)}\n- at individual level 1 in top 10: \t{len(D)}")

```

```

number of features : 56 ow actionnable
- at administrative level 1 :          27
- at individual level 1 :          25
- at administrative level 1 in top 10:          4
- at individual level 1 in top 10:          5

```

1.1.3 Learning and model performance evaluation on each clusters

```

In [15]: n_estimators_range = [16,32,64,128]
        max_depth_range = [2,4,8,16,32,64]
        param_grid = dict(n_estimators=n_estimators_range, max_depth = max_depth_range)
        params = {'max_features' : 'sqrt',
                  'random_state' : 32,
                  'min_samples_split' : 2,
                  'class_weight' : 'balanced'
                  }
        #scope = ( SelectFromModel_LogisticRegression_features ) & set(dataset.columns)
        features = df.loc[:,scope].columns

```

```

In [16]: score_clustering_methods = []
         clustering_methods = clustTest1.columns[2:3]

         for method in clustering_methods:
             print("-----")
             print(f"\nAnalysis cluster method {method}")
             cluster_list = clustTest1[method].unique()
             print(f"liste of clusters : {cluster_list}")
             score_cluster = []
             for cluster in cluster_list:
                 index_scope = clustTest1.loc[clustTest1[method]==cluster,:].index
                 print(f"cluster {cluster} '{cluster_name[cluster]}' : {len(index_scope)} elements")

                 Xc = X.loc[index_scope.intersection(X.index),:]
                 yc = y[index_scope.intersection(X.index)]

                 Xs, ys = resample(Xc, yc, random_state=42)

                 Xs = Xs.iloc[0:n_max,:]
                 ys = ys.iloc[0:n_max]

                 X_train, X_test, y_train, y_test = train_test_split(Xs, ys,
                                                                       test_size=0.2,
                                                                       random_state=42)

                 scaler = StandardScaler().fit(X_train)
                 X_train = scaler.transform(X_train)
                 X_test = scaler.transform(X_test)

                 print(f"Number exemple: {ys.shape[0]}\n\
- training set: {y_train.shape[0]}\n\
- test set: {y_test.shape[0]}")
                 print(f"Number of features: p={X_train.shape[1]}")
                 print(f"Number of class: {len(np.unique(y))}")
                 for c in np.unique(y):
                     print(f"class {c:0.0f} : {100*np.sum(yc==c)/len(yc):0.1f}%")

                 startTime = time.time()
                 clf = RandomForestClassifier(**params)
                 grid = GridSearchCV(clf,
                                     scoring='accuracy',
                                     param_grid=param_grid)

                 grid.fit(X_train, y_train)
                 print(f"Optimal values are {grid.best_params_} \n\
cross validation score {100*grid.best_score_:0.2f}%")
                 print()

```

```

# Learning on full training set with optimal hyperparameters and score on test
params_opt = {'max_features' : 'sqrt', 'random_state' : 32,
              'min_samples_split' : 2, 'class_weight' : 'balanced',
              'n_estimators' : grid.best_params_['n_estimators'],
              'max_depth' : grid.best_params_['max_depth']}
clf = RandomForestClassifier(**params_opt).fit(X_train, y_train)

y_test_pred = clf.predict(X_test)
accuracy = clf.score(X_test, y_test)
f1 = f1_score(y_test, y_test_pred)
p = precision_score(y_test, y_test_pred)
r = recall_score(y_test, y_test_pred)

res = {'f1_score' : f1,
       'accuracy' : accuracy,
       'precision' : p,
       'recall' : r}

importances = clf.feature_importances_
std = np.std([tree.feature_importances_ for tree in clf.estimators_],
             axis=0)
indices = np.argsort(importances)[::-1]
features_name = np.array(features)

cl = {'cluster' : cluster,
      'size' : len(index_scope),
      'model' : 'RandomForestClassifier',
      'params' : params_opt,
      'metrics' : res,
      'importances' : importances,
      'std' : std,
      'indices' : indices,
      'features_name' : features_name
      }

score_cluster.append(cl)

d = {'clustering_method' : method,
     'cluster_scores' : score_cluster
     }
score_clustering_methods.append(d)

```

Analysis cluster method clust3
 liste of clusters : [2 4 6 1 3 5]
 cluster 2 'Equilibrés' : 3053 elements
 Number exemple: 2000
 - training set: 1600
 - test set: 400
 Number of features: p=56
 Number of class: 2
 class 0 : 34.7%
 class 1 : 65.3%
 Optimal values are {'max_depth': 16, 'n_estimators': 64}
 cross validation score 80.06%

cluster 4 'Domestiques modérés' : 2359 elements
 Number exemple: 2000
 - training set: 1600
 - test set: 400
 Number of features: p=56
 Number of class: 2
 class 0 : 32.7%
 class 1 : 67.3%
 Optimal values are {'max_depth': 32, 'n_estimators': 128}
 cross validation score 84.38%

cluster 6 'Domestiques stricts' : 2313 elements
 Number exemple: 2000
 - training set: 1600
 - test set: 400
 Number of features: p=56
 Number of class: 2
 class 0 : 32.8%
 class 1 : 67.2%
 Optimal values are {'max_depth': 16, 'n_estimators': 128}
 cross validation score 83.56%

cluster 1 'Civiques' : 528 elements
 Number exemple: 505
 - training set: 404
 - test set: 101
 Number of features: p=56
 Number of class: 2
 class 0 : 48.3%
 class 1 : 51.7%
 Optimal values are {'max_depth': 8, 'n_estimators': 128}
 cross validation score 81.93%

cluster 3 'Flâneurs' : 1384 elements
 Number exemple: 1367

```

- training set: 1093
- test set: 274
Number of features: p=56
Number of class: 2
class 0 : 31.6%
class 1 : 68.4%
Optimal values are {'max_depth': 32, 'n_estimators': 128}
cross validation score 86.28%

```

```

cluster 5 'Solitaires' : 1494 elements
Number exemple: 1472
- training set: 1177
- test set: 295
Number of features: p=56
Number of class: 2
class 0 : 40.9%
class 1 : 59.1%
Optimal values are {'max_depth': 32, 'n_estimators': 128}
cross validation score 83.77%

```

1.1.4 Performance gain obtained using clustering

```

In [17]: # F1 score
for score_method in score_clustering_methods:
    print(f"method {score_method['clustering_method']}:")
    average_score = 0
    total_size = 0
    for i, score_cluster in enumerate(score_method['cluster_scores']):
        print(f"cluster {score_cluster['cluster']} '{cluster_name[score_cluster['cluster_name']]}'")
        average_score += score_cluster['metrics']['f1_score']*score_cluster['size']
        total_size += score_cluster['size']

    average_score = average_score / total_size
    print(f"average f1 on clusters {100*average_score:0.1f}% gain {100*(average_score-r

method clust3:
cluster 2 'Equilibrés' : (3053), f1 macro 90.0%
cluster 4 'Domestiques modérés' : (2359), f1 macro 92.0%
cluster 6 'Domestiques stricts' : (2313), f1 macro 92.0%
cluster 1 'Civiques' : (528), f1 macro 89.6%
cluster 3 'Flâneurs' : (1384), f1 macro 92.1%
cluster 5 'Solitaires' : (1494), f1 macro 89.5%
average f1 on clusters 91.0% gain 9.9

```

```

In [18]: # accuracy

```

```

for score_method in score_clustering_methods:
    print(f"method {score_method['clustering_method']}:")
    average_score = 0
    total_size = 0
    for i, score_cluster in enumerate(score_method['cluster_scores']):
        print(f"cluster {score_cluster['cluster']} '{cluster_name[score_cluster['cluster_name']]}'")
        average_score = average_score + score_cluster['metrics']['accuracy']*score_cluster['size']
        total_size += score_cluster['size']
    average_score = average_score / total_size
    print(f"average accuracy on clusters {100*average_score:0.1f}% gain {100*(average_score - baseline):0.1f}%")

method clust3:
cluster 2 'Equilibrés' : (3053) accuracy 86.0%
cluster 4 'Domestiques modérés' : (2359) accuracy 88.2%
cluster 6 'Domestiques stricts' : (2313) accuracy 88.2%
cluster 1 'Civiques' : (528) accuracy 90.1%
cluster 3 'Flâneurs' : (1384) accuracy 88.3%
cluster 5 'Solitaires' : (1494) accuracy 87.5%
average accuracy on clusters 87.6% gain 14.1

```

1.1.5 Feature importance of the models & actionable variables

```

In [19]: # Feature importance by cluster
for score_method in score_clustering_methods:
    print(f"method {score_method['clustering_method']}:")
    for i, score_cluster in enumerate(score_method['cluster_scores']):
        print(f"cluster {score_cluster['cluster']} ({score_cluster['size']}), f1 macro {score_cluster['f1_macro']}")
        print(f"top 15 features:")
        indices = score_cluster['indices']
        features_name = score_cluster['features_name']
        importances = score_cluster['importances']
        features_short_name_sorted = [ name[:15] for name in features_name[indices]]

        # Plot the feature importances of the forest
        plt.figure()
        plt.gcf().subplots_adjust(bottom=0.4)
        plt.title(f"Feature importances\nHappiness model for '{cluster_name[score_cluster['cluster_name']]}'")
        plt.bar(range(n_features), importances[indices][:n_features],
                color="r", yerr=std[indices][:n_features], align="center")
        plt.xticks(range(n_features), features_short_name_sorted[:n_features], rotation=45)
        plt.xlim([-1, n_features])
        filename = path_fig / Path(f"feature_importance_cluster_{score_cluster['cluster_name']}")
        plt.savefig(filename, format='jpg')
        plt.show()

    for f in range(n_features_max):

```

```

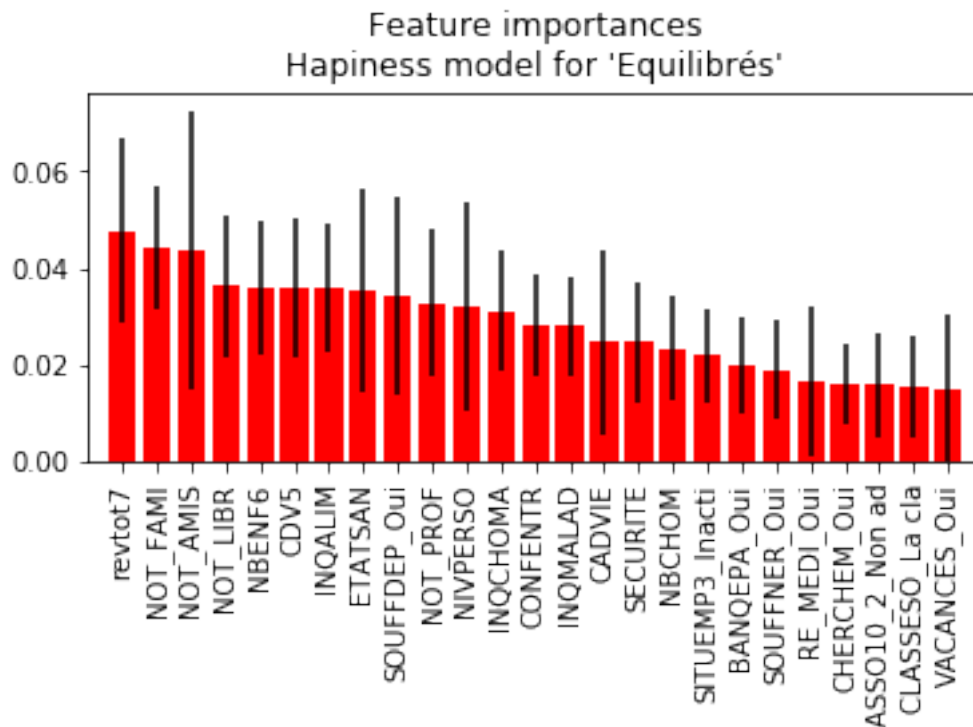
print("%d. feature %d -%s- (%f)" % (f + 1, indices[f], features_name[indices[f]], features_name[indices[f]]))
if features_name[indices[f]] in actionable_individual_1_features:
    print("\tActionable at individual level (1)")
if features_name[indices[f]] in actionable_individual_2_features:
    print("\tActionable at individual level (2)")
if features_name[indices[f]] in actionable_admin_1_features:
    print("\tActionable at administrative level (1)")
if features_name[indices[f]] in actionable_admin_2_features:
    print("\tActionable at administrative level (2)")

```

method clust3:

cluster 2 (3053), f1 macro 90.0%

top 15 features:



1. feature 27 -revtot7- (0.047657)
 Actionable at individual level (2)
 Actionable at administrative level (2)
2. feature 16 -NOT_FAMI- (0.044059)
 Actionable at individual level (1)
 Actionable at administrative level (2)
3. feature 49 -NOT_AMIS- (0.043665)
 Actionable at individual level (1)
 Actionable at administrative level (2)

4. feature 34 -NOT_LIBR- (0.036328)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
5. feature 54 -NBENF6- (0.035803)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
6. feature 8 -CDV5- (0.035799)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
7. feature 2 -INQALIM- (0.035799)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
8. feature 4 -ETATSAN- (0.035452)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
9. feature 48 -SOUFFDEP_Oui- (0.034260)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
10. feature 36 -NOT_PROF- (0.032705)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
11. feature 37 -NIVPERSO- (0.031991)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
12. feature 26 -INQCHOMA- (0.031153)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
13. feature 28 -CONFENTR- (0.028112)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
14. feature 44 -INQMALAD- (0.028024)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
15. feature 19 -CADVIE- (0.024752)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
16. feature 21 -SECURITE- (0.024632)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
17. feature 45 -NBCHOM- (0.023313)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
18. feature 5 -SITUEMP3_Inactif- (0.021885)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
19. feature 22 -BANQEPA_Oui- (0.020176)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)

20. feature 53 -SOUFFNER_Oui- (0.018885)
 Actionable at individual level (2)
 Actionable at administrative level (1)

21. feature 40 -RE_MEDI_Oui- (0.016657)
 Actionable at individual level (2)
 Actionable at administrative level (2)

22. feature 31 -CHERCHEM_Oui- (0.016081)
 Actionable at individual level (1)
 Actionable at administrative level (2)

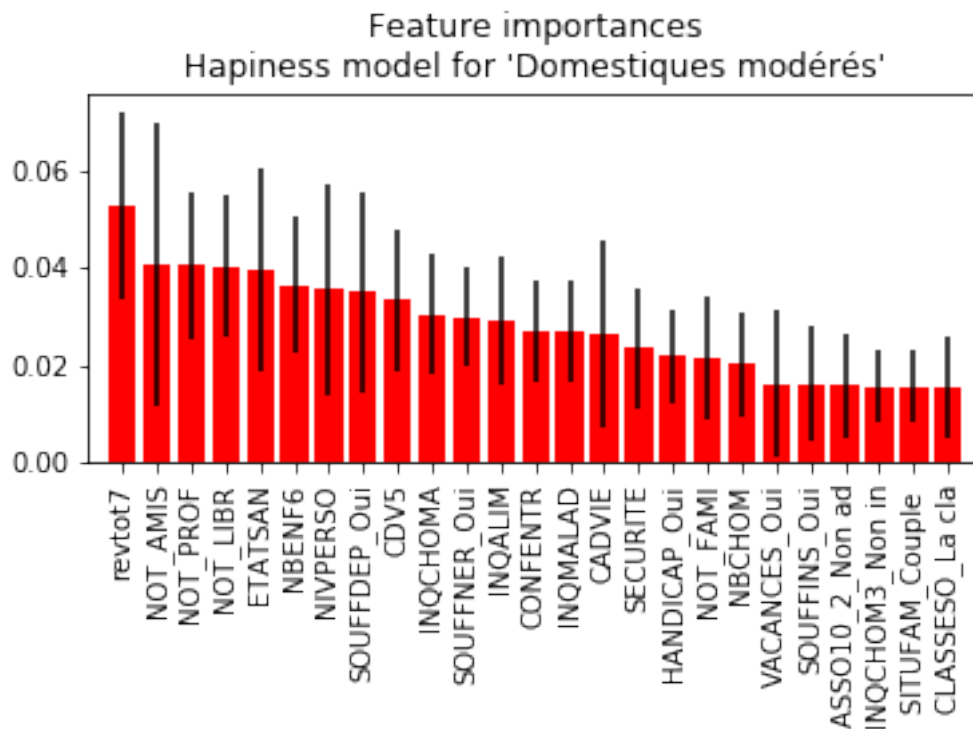
23. feature 46 -ASSO10_2_Non adhérent- (0.015903)
 Actionable at individual level (1)
 Actionable at administrative level (1)

24. feature 18 -CLASSES0_La classe moyenne supérieure- (0.015272)
 Actionable at individual level (1)
 Actionable at administrative level (1)

25. feature 35 -VACANCES_Oui- (0.015246)
 Actionable at individual level (1)
 Actionable at administrative level (1)

cluster 4 (2359), f1 macro 92.0%

top 15 features:



1. feature 27 -revtot7- (0.052969)
 Actionable at individual level (2)

- Actionable at administrative level (2)
- 2. feature 49 -NOT_AMIS- (0.040807)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
- 3. feature 36 -NOT_PROF- (0.040491)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
- 4. feature 34 -NOT_LIBR- (0.040371)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 5. feature 4 -ETATSAN- (0.039456)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 6. feature 54 -NBENF6- (0.036584)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
- 7. feature 37 -NIVPERSON- (0.035630)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
- 8. feature 48 -SOUFFDEP_Oui- (0.034990)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
- 9. feature 8 -CDV5- (0.033469)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
- 10. feature 26 -INQCHOMA- (0.030509)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 11. feature 53 -SOUFFNER_Oui- (0.029968)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
- 12. feature 2 -INQALIM- (0.029437)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 13. feature 28 -CONFENTR- (0.027193)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 14. feature 44 -INQMALAD- (0.027030)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 15. feature 19 -CADVIE- (0.026493)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 16. feature 21 -SECURITE- (0.023495)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
- 17. feature 41 -HANDICAP_Oui- (0.021881)
 - Actionable at administrative level (2)

18. feature 16 -NOT_FAMI- (0.021367)
 Actionable at individual level (1)
 Actionable at administrative level (2)

19. feature 45 -NBCHOM- (0.020399)
 Actionable at individual level (1)
 Actionable at administrative level (1)

20. feature 35 -VACANCES_Oui- (0.016365)
 Actionable at individual level (1)
 Actionable at administrative level (1)

21. feature 15 -SOUFFINS_Oui- (0.016315)
 Actionable at individual level (2)
 Actionable at administrative level (1)

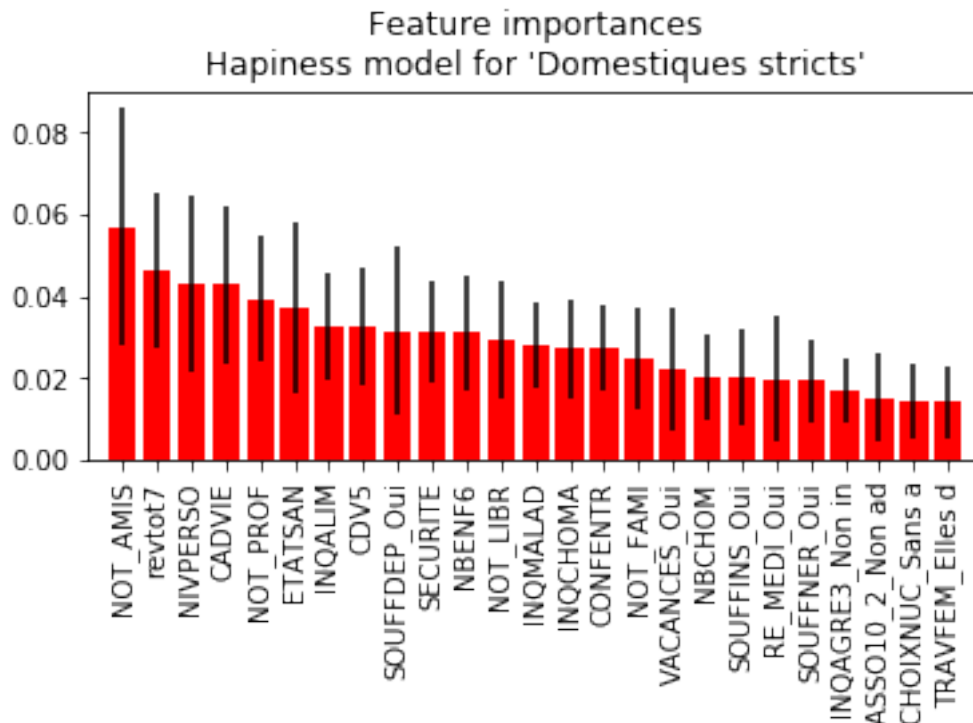
22. feature 46 -ASSO10_2_Non adhérent- (0.015830)
 Actionable at individual level (1)
 Actionable at administrative level (1)

23. feature 39 -INQCHOM3_Non inquiet- (0.015774)
 Actionable at individual level (1)
 Actionable at administrative level (1)

24. feature 42 -SITUFAM_Couple sans enfants- (0.015770)
 Actionable at individual level (2)

25. feature 18 -CLASSES0_La classe moyenne supérieure- (0.015742)
 Actionable at individual level (1)
 Actionable at administrative level (1)

cluster 6 (2313), f1 macro 92.0%
 top 15 features:



1. feature 49 -NOT_AMIS- (0.057038)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
2. feature 27 -revtot7- (0.046253)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
3. feature 37 -NIVPERSO- (0.043320)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
4. feature 19 -CADVIE- (0.042762)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
5. feature 36 -NOT_PROF- (0.039357)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
6. feature 4 -ETATSAN- (0.037085)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
7. feature 2 -INQALIM- (0.032531)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
8. feature 8 -CDV5- (0.032395)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
9. feature 48 -SOUFFDEP_Oui- (0.031513)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
10. feature 21 -SECURITE- (0.031189)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
11. feature 54 -NBENF6- (0.031022)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
12. feature 34 -NOT_LIBR- (0.029325)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
13. feature 44 -INQMALAD- (0.027975)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
14. feature 26 -INQCHOMA- (0.027109)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
15. feature 28 -CONFENTR- (0.027101)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)

16. feature 16 -NOT_FAMI- (0.024628)
 Actionable at individual level (1)
 Actionable at administrative level (2)

17. feature 35 -VACANCES_Oui- (0.022356)
 Actionable at individual level (1)
 Actionable at administrative level (1)

18. feature 45 -NBCHOM- (0.020099)
 Actionable at individual level (1)
 Actionable at administrative level (1)

19. feature 15 -SOUFFINS_Oui- (0.020066)
 Actionable at individual level (2)
 Actionable at administrative level (1)

20. feature 40 -RE_MEDI_Oui- (0.019796)
 Actionable at individual level (2)
 Actionable at administrative level (2)

21. feature 53 -SOUFFNER_Oui- (0.019412)
 Actionable at individual level (2)
 Actionable at administrative level (1)

22. feature 11 -INQAGRE3_Non inquiet- (0.016937)
 Actionable at individual level (1)
 Actionable at administrative level (1)

23. feature 46 -ASSO10_2_Non adhérent- (0.015155)
 Actionable at individual level (1)
 Actionable at administrative level (1)

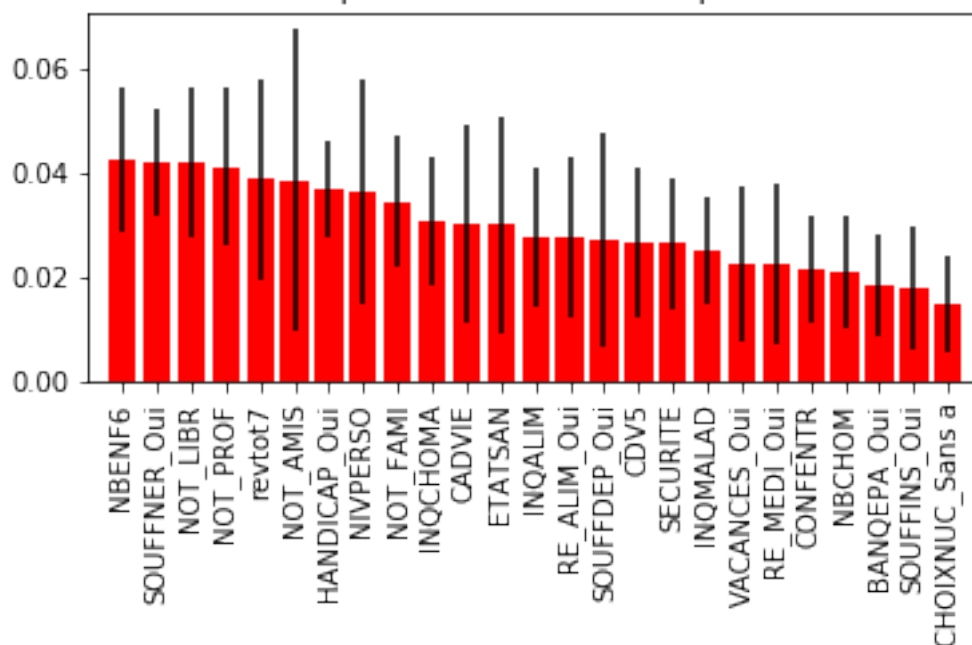
24. feature 17 -CHOIXNUC_Sans avis- (0.014534)
 Actionable at individual level (2)
 Actionable at administrative level (2)

25. feature 30 -TRAVFEM_Elles devraient travailler quand elles le désirent- (0.014191)
 Actionable at individual level (2)
 Actionable at administrative level (2)

cluster 1 (528), f1 macro 89.6%

top 15 features:

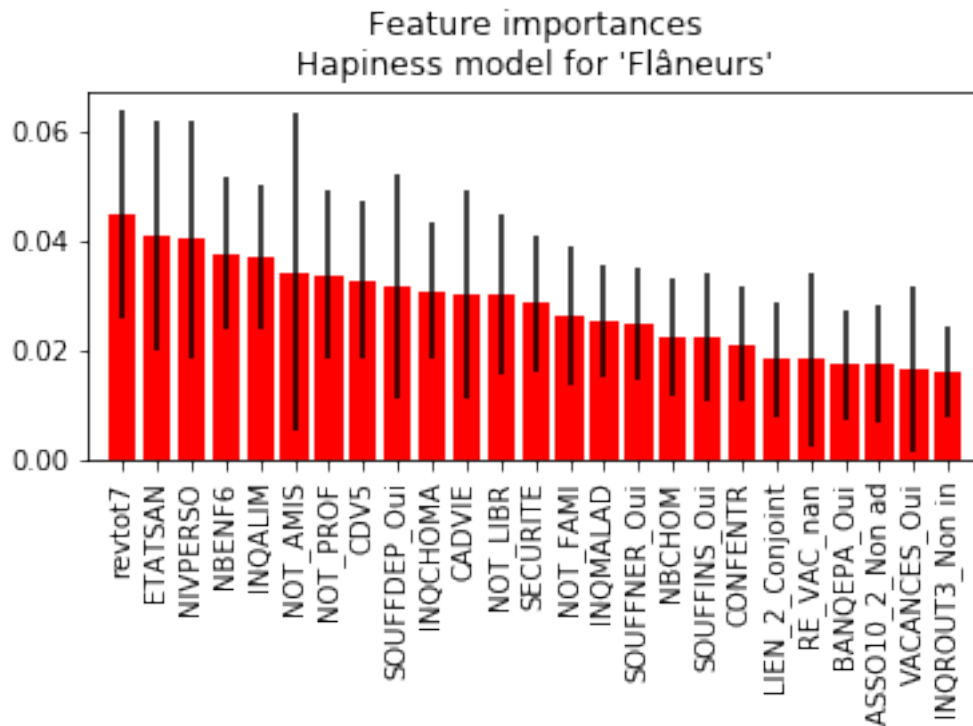
Feature importances
Hapiness model for 'Civiques'



1. feature 54 -NBENF6- (0.042514)
Actionable at individual level (2)
Actionable at administrative level (2)
2. feature 53 -SOUFFNER_Oui- (0.042101)
Actionable at individual level (2)
Actionable at administrative level (1)
3. feature 34 -NOT_LIBR- (0.041892)
Actionable at individual level (1)
Actionable at administrative level (1)
4. feature 36 -NOT_PROF- (0.041140)
Actionable at individual level (1)
Actionable at administrative level (2)
5. feature 27 -revtot7- (0.038693)
Actionable at individual level (2)
Actionable at administrative level (2)
6. feature 49 -NOT_AMIS- (0.038585)
Actionable at individual level (1)
Actionable at administrative level (2)
7. feature 41 -HANDICAP_Oui- (0.036950)
Actionable at administrative level (2)
8. feature 37 -NIVPERSO- (0.036182)
Actionable at individual level (2)
Actionable at administrative level (2)

9. feature 16 -NOT_FAMI- (0.034522)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
10. feature 26 -INQCHOMA- (0.030862)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
11. feature 19 -CADVIE- (0.030240)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
12. feature 4 -ETATSAN- (0.030033)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
13. feature 2 -INQALIM- (0.027554)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
14. feature 1 -RE_ALIM_Oui- (0.027496)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
15. feature 48 -SOUFFDEP_Oui- (0.026946)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
16. feature 8 -CDV5- (0.026487)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
17. feature 21 -SECURITE- (0.026457)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
18. feature 44 -INQMALAD- (0.024914)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
19. feature 35 -VACANCES_Oui- (0.022484)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
20. feature 40 -RE_MEDI_Oui- (0.022319)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
21. feature 28 -CONFENTR- (0.021463)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
22. feature 45 -NBCHOM- (0.020811)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
23. feature 22 -BANQEPA_Oui- (0.018319)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
24. feature 15 -SOUFFINS_Oui- (0.017853)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)

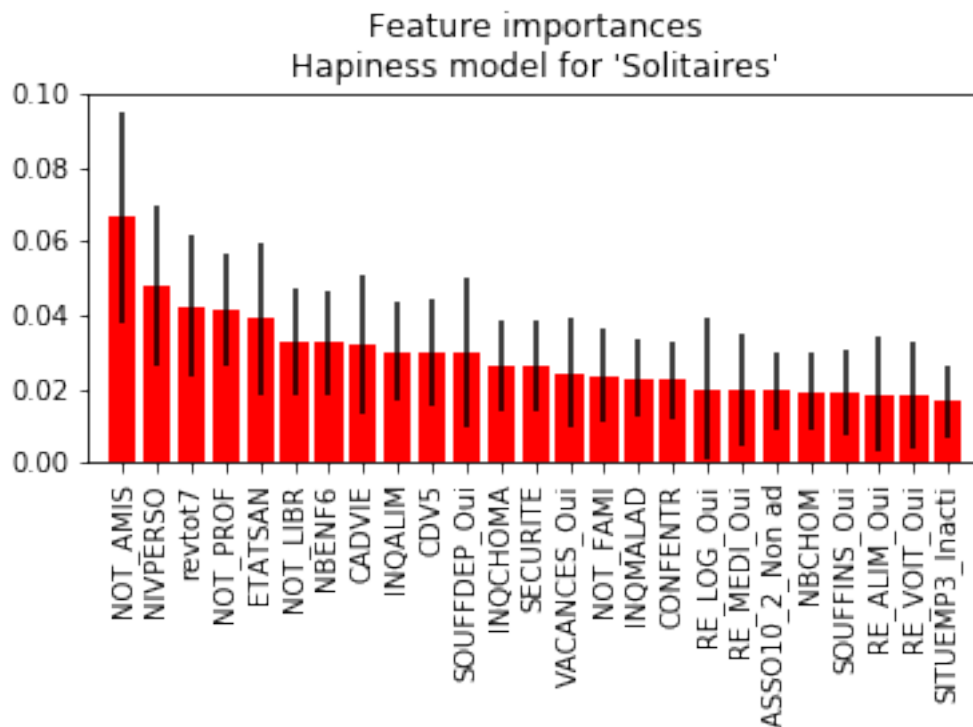
25. feature 17 -CHOIXNUC_Sans avis- (0.014681)
 Actionable at individual level (2)
 Actionable at administrative level (2)
 cluster 3 (1384), f1 macro 92.1%
 top 15 features:



1. feature 27 -revtot7- (0.044798)
 Actionable at individual level (2)
 Actionable at administrative level (2)
2. feature 4 -ETATSAN- (0.040740)
 Actionable at individual level (1)
 Actionable at administrative level (1)
3. feature 37 -NIVPERSO- (0.040127)
 Actionable at individual level (2)
 Actionable at administrative level (2)
4. feature 54 -NBENF6- (0.037606)
 Actionable at individual level (2)
 Actionable at administrative level (2)
5. feature 2 -INQALIM- (0.037001)
 Actionable at individual level (1)
 Actionable at administrative level (1)
6. feature 49 -NOT_AMIS- (0.034235)
 Actionable at individual level (1)

- Actionable at administrative level (2)
- 7. feature 36 -NOT_PROF- (0.033737)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
- 8. feature 8 -CDV5- (0.032764)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
- 9. feature 48 -SOUFFDEP_Oui- (0.031735)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
- 10. feature 26 -INQCHOMA- (0.030744)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 11. feature 19 -CADVIE- (0.030097)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 12. feature 34 -NOT_LIBR- (0.030080)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 13. feature 21 -SECURITE- (0.028604)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
- 14. feature 16 -NOT_FAMI- (0.026464)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
- 15. feature 44 -INQMALAD- (0.025425)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 16. feature 53 -SOUFFNER_Oui- (0.024982)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
- 17. feature 45 -NBCHOM- (0.022369)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 18. feature 15 -SOUFFINS_Oui- (0.022306)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
- 19. feature 28 -CONFENTR- (0.021033)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 20. feature 10 -LIEN_2_Conjoint ou compagnon- (0.018351)
- 21. feature 12 -RE_VAC_nan- (0.018308)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
- 22. feature 22 -BANQEPA_Oui- (0.017497)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 23. feature 46 -ASS010_2_Non adhérent- (0.017463)

Actionable at individual level (1)
 Actionable at administrative level (1)
 24. feature 35 -VACANCES_Oui- (0.016800)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 25. feature 0 -INQROUT3_Non inquiet- (0.016194)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 cluster 5 (1494), f1 macro 89.5%
 top 15 features:



1. feature 49 -NOT_AMIS- (0.066353)
 Actionable at individual level (1)
 Actionable at administrative level (2)
 2. feature 37 -NIVPERSO- (0.047735)
 Actionable at individual level (2)
 Actionable at administrative level (2)
 3. feature 27 -revtot7- (0.042365)
 Actionable at individual level (2)
 Actionable at administrative level (2)
 4. feature 36 -NOT_PROF- (0.041131)
 Actionable at individual level (1)
 Actionable at administrative level (2)

5. feature 4 -ETATSAN- (0.038959)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
6. feature 34 -NOT_LIBR- (0.032890)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
7. feature 54 -NBENF6- (0.032387)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
8. feature 19 -CADVIE- (0.031988)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
9. feature 2 -INQALIM- (0.030055)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
10. feature 8 -CDV5- (0.029810)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
11. feature 48 -SOUFFDEP_Oui- (0.029698)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
12. feature 26 -INQCHOMA- (0.026401)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
13. feature 21 -SECURITE- (0.026207)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
14. feature 35 -VACANCES_Oui- (0.024314)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
15. feature 16 -NOT_FAMI- (0.023395)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
16. feature 44 -INQMALAD- (0.022752)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
17. feature 28 -CONFENTR- (0.022350)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
18. feature 29 -RE_LOG_Oui- (0.019795)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
19. feature 40 -RE_MEDI_Oui- (0.019517)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
20. feature 46 -ASS010_2_Non adhérent- (0.019319)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)

```

21. feature 45 -NBCHOM- (0.019135)
    Actionable at individual level (1)
    Actionable at administrative level (1)
22. feature 15 -SOUFFINS_Oui- (0.019126)
    Actionable at individual level (2)
    Actionable at administrative level (1)
23. feature 1 -RE_ALIM_Oui- (0.018479)
    Actionable at individual level (2)
    Actionable at administrative level (2)
24. feature 6 -RE_VOIT_Oui- (0.018183)
    Actionable at individual level (2)
    Actionable at administrative level (2)
25. feature 5 -SITUEMP3_Inactif- (0.016503)
    Actionable at individual level (2)
    Actionable at administrative level (2)

```

1.2 Option 2

```

In [20]: # choosing set of features
scope = dict_features_sets.get('insee_environment_score_features', set())
scope = scope | dict_features_sets.get('insee_recreation_score_features', set())
#scope = scope | dict_features_sets.get('cdv_actionable_admin_1_features', set())
scope = scope | RFE_LogisticRegression_20_features
print(f"number of features : {len(scope)} ow actionable")
A = scope & dict_features_sets.get('cdv_actionable_admin_1_features', set())
B = scope & dict_features_sets.get('cdv_actionable_individual_1_features', set())
print(f"- at administrative level 1 : \t{len(A)}\n- at individual level 1 : \t{len(B)}")

```

```

number of features : 34 ow actionable
- at administrative level 1 :          11
- at individual level 1 :           9

```

```

In [21]: df = dataset.loc[:, :]
# reducing problem to a 2 class classification problem
df["HEUREUX_CLF"] = 0
df.loc[df["HEUREUX"]==4, "HEUREUX_CLF"] = 1
df.loc[df["HEUREUX"]==3, "HEUREUX_CLF"] = 1
df.loc[df["HEUREUX"]==5, "HEUREUX_CLF"] = None

scope = scope & set(dataset.columns)
n_max = 2000

df = df.loc[:, scope | {"HEUREUX_CLF"} ].dropna()
features = df.loc[:, scope ].columns

X = df.loc[:, scope]

```

```

y = df["HEUREUX_CLF"]

Xs, ys = resample(X, y, random_state=42)

Xs = Xs.iloc[0:n_max,:]
ys = ys.iloc[0:n_max]

X_train, X_test, y_train, y_test = train_test_split(Xs, ys,
                                                    test_size=0.2,
                                                    random_state=42
                                                    )

scaler = StandardScaler().fit(X_train)
X_train = scaler.transform(X_train)
X_test = scaler.transform(X_test)

print(f"Number exemple: {y.shape[0]}\n- training set: \
{y_train.shape[0]}\n- test set: {y_test.shape[0]}")
print(f"Number of features: p={X_train.shape[1]}")
print(f"Number of class: {len(np.unique(y))}")
for c in np.unique(y):
    print(f"class {c:0.0f} : {100*np.sum(y==c)/len(y):0.1f}%")

```

```

Number exemple: 10858
- training set: 1600
- test set: 400
Number of features: p=34
Number of class: 2
class 0 : 34.9%
class 1 : 65.1%

```

```

In [22]: startTime = time.time()
n_estimators_range = [32,64,128,256,512]
max_depth_range = [4,8,16,32,64]
param_grid = dict(n_estimators=n_estimators_range, max_depth = max_depth_range)

params = {'max_features' : 'sqrt', 'random_state' : 32,
          'min_samples_split' : 2, 'class_weight' : 'balanced'}
clf = RandomForestClassifier(**params)

grid = GridSearchCV(clf, scoring='accuracy', param_grid=param_grid)
grid.fit(X_train, y_train)
print(f"Determination of optimal hyperparameters in {time.time() - startTime:0.1f} s")
print(f"Optimal values are {grid.best_params_} \n\
Accuracy Score of cross validation {100*grid.best_score_:0.2f}%")

```

```

# Learning on full training set with optimal hyperparameters and score on test set
params = {'max_features' : 'sqrt', 'random_state' : 32,
          'min_samples_split' : 2, 'class_weight' : 'balanced',
          'n_estimators' : grid.best_params_['n_estimators'],
          'max_depth' : grid.best_params_['max_depth']}
clf = RandomForestClassifier(**params).fit(X_train, y_train)
clf.fit(X_train, y_train)
y_test_pred = clf.predict(X_test)

print(f"Random Forest, p={X_train.shape[1]}")
accuracy = clf.score(X_test, y_test)
f1 = f1_score(y_test, y_test_pred)
p = precision_score(y_test, y_test_pred)
r = recall_score(y_test, y_test_pred)
print(f"Model score\n- Accuracy : {accuracy*100:0.1f} %")
print(f"- Precision : {p*100:0.1f} % (Happy # positive class)")
print(f"- Recall : {r*100:0.1f} %")
print(f"- F1 score : {f1*100:0.1f} %")
res_full = {
    'f1_score' : f1,
    'accuracy' : accuracy,
    'precision' : p,
    'recall' : r
}

```

Determination of optimal hyperparameters in 63.2 s
 Optimal values are {'max_depth': 16, 'n_estimators': 256}
 Accuracy Score of cross validation 77.25%
 Random Forest, p=34
 Model score
 - Accuracy : 73.5 %
 - Precision : 71.9 % (Happy # positive class)
 - Recall : 94.0 %
 - F1 score : 81.5 %

```

In [23]: importances = clf.feature_importances_
std = np.std([tree.feature_importances_ for tree in clf.estimators_],
             axis=0)
indices = np.argsort(importances)[::-1]
features_name = np.array(features)
features_short_name_sorted = [ name[:15] for name in features_name[indices]]
print("Feature ranking:")

n_features_max = 25
n_features = min(X.shape[1], n_features_max)

# Plot the feature importances of the forest

```

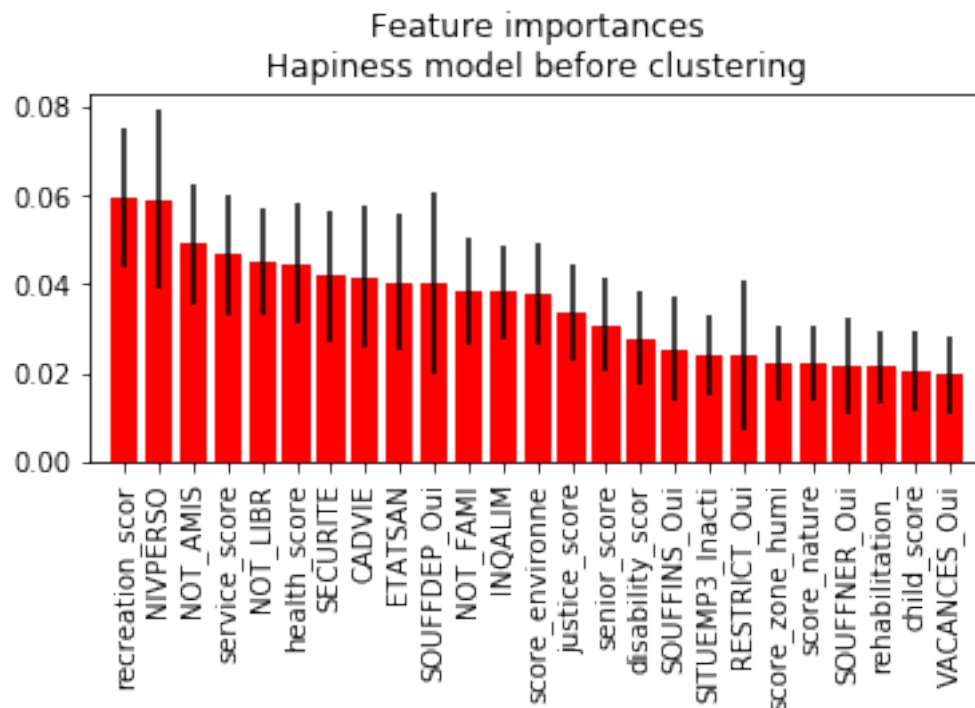
```

plt.figure()
plt.gcf().subplots_adjust(bottom=0.4)
plt.title("Feature importances\nHapiness model before clustering")
plt.bar(range(n_features), importances[indices][:n_features],
        color="r", yerr=std[indices[:n_features]], align="center")
plt.xticks(range(n_features), features_short_name_sorted[:n_features], rotation=90)
plt.xlim([-1, n_features])
filename = path_fig / Path("feature_importance_option2.jpg")
plt.savefig(filename, format='jpg')
plt.show()

for f in range(min(X.shape[1],n_features_max)):
    print("%d. feature %d -%s- (%f)" % (f + 1, indices[f],features_name[indices[f]], im
    if features_name[indices[f]] in actionable_individual_1_features:
        print("\tActionable at individual level (1)")
    if features_name[indices[f]] in actionable_individual_2_features:
        print("\tActionable at individual level (2)")
    if features_name[indices[f]] in actionable_admin_1_features:
        print("\tActionable at administrative level (1)")
    if features_name[indices[f]] in actionable_admin_2_features:
        print("\tActionable at administrative level (2)")

```

Feature ranking:



1. feature 13 -recreation_score- (0.059650)
2. feature 21 -NIVPERSON- (0.059020)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
3. feature 28 -NOT_AMIS- (0.049001)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
4. feature 11 -service_score- (0.046573)
5. feature 19 -NOT_LIBR- (0.045039)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
6. feature 26 -health_score- (0.044641)
7. feature 12 -SECURITE- (0.041818)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
8. feature 10 -CADVIE- (0.041653)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
9. feature 2 -ETATSAN- (0.040500)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
10. feature 27 -SOUFFDEP_Oui- (0.040179)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
11. feature 9 -NOT_FAMI- (0.038618)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
12. feature 1 -INQALIM- (0.038176)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
13. feature 23 -score_environnement- (0.037665)
14. feature 18 -justice_score- (0.033420)
15. feature 32 -senior_score- (0.030732)
16. feature 30 -disability_score- (0.027630)
17. feature 8 -SOUFFINS_Oui- (0.025481)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
18. feature 3 -SITUEMP3_Inactif- (0.024076)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
19. feature 17 -RESTRICT_Oui- (0.024073)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
20. feature 16 -score_zone_humide- (0.022225)
21. feature 22 -score_nature- (0.022072)
22. feature 31 -SOUFFNER_Oui- (0.021519)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)


```

23. feature 33 -rehabilitation_score- (0.021483)
24. feature 15 -child_score- (0.020440)
25. feature 20 -VACANCES_Oui- (0.019738)
    Actionable at individual level (1)
    Actionable at administrative level (1)

```

```

In [24]: print(f"number of features : {len(scope)} ow actionnable")
        A = scope & dict_features_sets.get('cdv_actionable_admin_1_features',set())
        B = scope & dict_features_sets.get('cdv_actionable_individual_1_features',set())
        print(f"- at administrative level 1 : \t{len(A)}\n- at individual level 1 : \t{len(B)}")
        important_features = set(features_name[indices][:10])
        C = A & important_features
        D = B & important_features
        print(f"- at administrative level 1 in top 10: \t{len(C)}\n- at individual level 1 in t

number of features : 34 ow actionnable
- at administrative level 1 :          11
- at individual level 1 :           9
- at administrative level 1 in top 10:          5
- at individual level 1 in top 10:           4

```

```

In [25]: n_estimators_range = [16,32,64,128]
        max_depth_range = [2,4,8,16,32,64]
        param_grid = dict(n_estimators=n_estimators_range, max_depth = max_depth_range)
        params = {'max_features' : 'sqrt',
                  'random_state' : 32,
                  'min_samples_split' : 2,
                  'class_weight' : 'balanced'
                  }
        #scope = ( SelectFromModel_LogisticRegression_features ) & set(dataset.columns)
        features = df.loc[:,scope].columns

```

```

In [26]: score_clustering_methods = []
        clustering_methods = clustTest1.columns[2:3]

        for method in clustering_methods:
            print("-----")
            print(f"\nAnalysis cluster method {method}")
            cluster_list = clustTest1[method].unique()
            print(f"liste of clusters : {cluster_list}")
            score_cluster = []
            for cluster in cluster_list:
                index_scope = clustTest1.loc[clustTest1[method]==cluster,:].index
                print(f"cluster {cluster} '{cluster_name[cluster]}' : {len(index_scope)} elemen

                Xc = X.loc[index_scope.intersection(X.index),:]
                yc = y[index_scope.intersection(X.index)]

```

```

Xs, ys = resample(Xc, yc, random_state=42)

Xs = Xs.iloc[0:n_max,:]
ys = ys.iloc[0:n_max]

X_train, X_test, y_train, y_test = train_test_split(Xs, ys,
                                                    test_size=0.2,
                                                    random_state=42)

scaler = StandardScaler().fit(X_train)
X_train = scaler.transform(X_train)
X_test = scaler.transform(X_test)

print(f"Number exemple: {ys.shape[0]}\n\
- training set: {y_train.shape[0]}\n\
- test set: {y_test.shape[0]}")
print(f"Number of features: p={X_train.shape[1]}")
print(f"Number of class: {len(np.unique(y))}")
for c in np.unique(y):
    print(f"class {c:0.0f} : {100*np.sum(yc==c)/len(yc):0.1f}%")

startTime = time.time()
clf = RandomForestClassifier(**params)
grid = GridSearchCV(clf,
                    scoring='accuracy',
                    param_grid=param_grid)

grid.fit(X_train, y_train)
print(f"Optimal values are {grid.best_params_} \n\
cross validation score {100*grid.best_score_:0.2f}%")
print()

# Learning on full training set with optimals hyperparameters and score on test
params_opt = {'max_features' : 'sqrt', 'random_state' : 32,
              'min_samples_split' : 2, 'class_weight' : 'balanced',
              'n_estimators' : grid.best_params_['n_estimators'],
              'max_depth' : grid.best_params_['max_depth']}
clf = RandomForestClassifier(**params_opt).fit(X_train, y_train)

y_test_pred = clf.predict(X_test)
accuracy = clf.score(X_test, y_test)
f1 = f1_score(y_test, y_test_pred)
p = precision_score(y_test, y_test_pred)
r = recall_score(y_test, y_test_pred)

```

```

res = {'f1_score' : f1,
       'accuracy' : accuracy,
       'precision' : p,
       'recall' : r}

importances = clf.feature_importances_
std = np.std([tree.feature_importances_ for tree in clf.estimators_],
             axis=0)
indices = np.argsort(importances)[::-1]
features_name = np.array(features)

cl = {'cluster' : cluster,
      'size' : len(index_scope),
      'model' : 'RandomForestClassifier',
      'params' : params_opt,
      'metrics' : res,
      'importances' : importances,
      'std' : std,
      'indices' : indices,
      'features_name' : features_name
      }

score_cluster.append(cl)

d = {'clustering_method' : method,
     'cluster_scores' : score_cluster
     }
score_clustering_methods.append(d)

```

```

-----

Analysis cluster method clust3
liste of clusters : [2 4 6 1 3 5]
cluster 2 'Equilibrés' : 3053 elements
Number exemple: 2000
    - training set: 1600
    - test set: 400
Number of features: p=34
Number of class: 2
class 0 : 34.7%
class 1 : 65.3%
Optimal values are {'max_depth': 16, 'n_estimators': 128}
cross validation score 81.69%

cluster 4 'Domestiques modérés' : 2359 elements
Number exemple: 2000

```

- training set: 1600
- test set: 400
Number of features: p=34
Number of class: 2
class 0 : 32.7%
class 1 : 67.3%
Optimal values are {'max_depth': 32, 'n_estimators': 128}
cross validation score 84.50%

cluster 6 'Domestiques stricts' : 2313 elements

Number exemple: 2000

- training set: 1600
- test set: 400
Number of features: p=34
Number of class: 2
class 0 : 32.7%
class 1 : 67.3%
Optimal values are {'max_depth': 16, 'n_estimators': 64}
cross validation score 83.56%

cluster 1 'Civiques' : 528 elements

Number exemple: 514

- training set: 411
- test set: 103
Number of features: p=34
Number of class: 2
class 0 : 48.2%
class 1 : 51.8%
Optimal values are {'max_depth': 16, 'n_estimators': 64}
cross validation score 82.00%

cluster 3 'Flâneurs' : 1384 elements

Number exemple: 1367

- training set: 1093
- test set: 274
Number of features: p=34
Number of class: 2
class 0 : 31.6%
class 1 : 68.4%
Optimal values are {'max_depth': 16, 'n_estimators': 64}
cross validation score 85.82%

cluster 5 'Solitaires' : 1494 elements

Number exemple: 1482

- training set: 1185
- test set: 297
Number of features: p=34
Number of class: 2

```
class 0 : 40.8%
class 1 : 59.2%
Optimal values are {'max_depth': 32, 'n_estimators': 128}
cross validation score 84.98%
```

```
In [27]: # F1 score
```

```
for score_method in score_clustering_methods:
    print(f"method {score_method['clustering_method']}:")
    average_score = 0
    total_size = 0
    for i, score_cluster in enumerate(score_method['cluster_scores']):
        print(f"cluster {score_cluster['cluster']} '{cluster_name[score_cluster['cluster']]}'")
        average_score += score_cluster['metrics']['f1_score']*score_cluster['size']
        total_size += score_cluster['size']

    average_score = average_score / total_size
    print(f"average f1 on clusters {100*average_score:0.1f}% gain {100*(average_score-r
```

```
method clust3:
```

```
cluster 2 'Equilibrés' : (3053), f1 macro 87.4%
cluster 4 'Domestiques modérés' : (2359), f1 macro 89.7%
cluster 6 'Domestiques stricts' : (2313), f1 macro 92.5%
cluster 1 'Civiques' : (528), f1 macro 86.0%
cluster 3 'Flâneurs' : (1384), f1 macro 93.0%
cluster 5 'Solitaires' : (1494), f1 macro 89.5%
average f1 on clusters 89.9% gain 8.4
```

```
In [28]: # accuracy
```

```
for score_method in score_clustering_methods:
    print(f"method {score_method['clustering_method']}:")
    average_score = 0
    total_size = 0
    for i, score_cluster in enumerate(score_method['cluster_scores']):
        print(f"cluster {score_cluster['cluster']} '{cluster_name[score_cluster['cluster']]}'")
        average_score = average_score + score_cluster['metrics']['accuracy']*score_cluster['size']
        total_size += score_cluster['size']
    average_score = average_score / total_size
    print(f"average accuracy on clusters {100*average_score:0.1f}% gain {100*(average_s
```

```
method clust3:
```

```
cluster 2 'Equilibrés' : (3053) accuracy 83.8%
cluster 4 'Domestiques modérés' : (2359) accuracy 86.0%
cluster 6 'Domestiques stricts' : (2313) accuracy 89.2%
cluster 1 'Civiques' : (528) accuracy 83.5%
cluster 3 'Flâneurs' : (1384) accuracy 89.8%
```

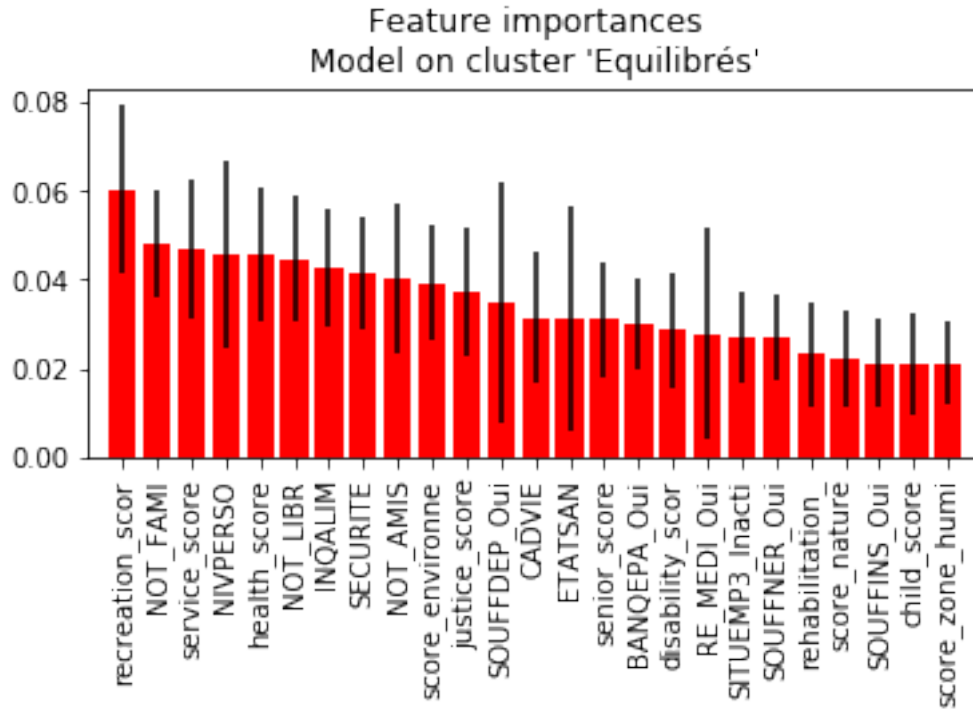
cluster 5 'Solitaires' : (1494) accuracy 88.6%
average accuracy on clusters 86.8% gain 13.3

```
In [29]: # Feature importance by cluster
for score_method in score_clustering_methods:
    print(f"method {score_method['clustering_method']}:")
    for i, score_cluster in enumerate(score_method['cluster_scores']):
        print(f"cluster {score_cluster['cluster']} ({score_cluster['size']}), f1 macro")
        print(f"top 15 features:")
        indices = score_cluster['indices']
        features_name = score_cluster['features_name']
        importances = score_cluster['importances']
        features_short_name_sorted = [ name[:15] for name in features_name[indices]]

        # Plot the feature importances of the forest
        plt.figure()
        plt.gcf().subplots_adjust(bottom=0.4)
        plt.title(f"Feature importances\nModel on cluster '{cluster_name[score_cluster['cluster_name']]}'")
        plt.bar(range(n_features), importances[indices][:n_features],
                color="r", yerr=std[indices][:n_features], align="center")
        plt.xticks(range(n_features), features_short_name_sorted[:n_features], rotation=45)
        plt.xlim([-1, n_features])
        filename = path_fig / Path(f"feature_importance_cluster_{score_cluster['cluster_name']}")
        plt.savefig(filename, format='jpg')
        plt.show()

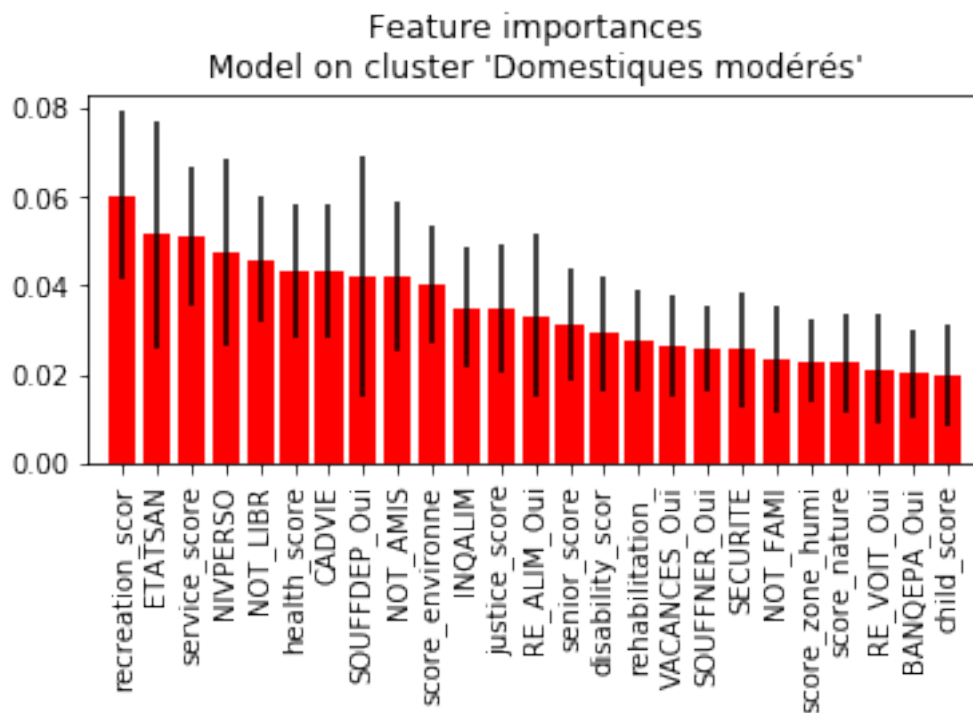
    for f in range(n_features_max):
        print("%d. feature %d -%s- (%f)" % (f + 1, indices[f], features_name[indices[f]], importances[indices[f]]))
        if features_name[indices[f]] in actionable_individual_1_features:
            print("\tActionable at individual level (1)")
        if features_name[indices[f]] in actionable_individual_2_features:
            print("\tActionable at individual level (2)")
        if features_name[indices[f]] in actionable_admin_1_features:
            print("\tActionable at administrative level (1)")
        if features_name[indices[f]] in actionable_admin_2_features:
            print("\tActionable at administrative level (2)")

method clust3:
cluster 2 (3053), f1 macro 87.4%
top 15 features:
```



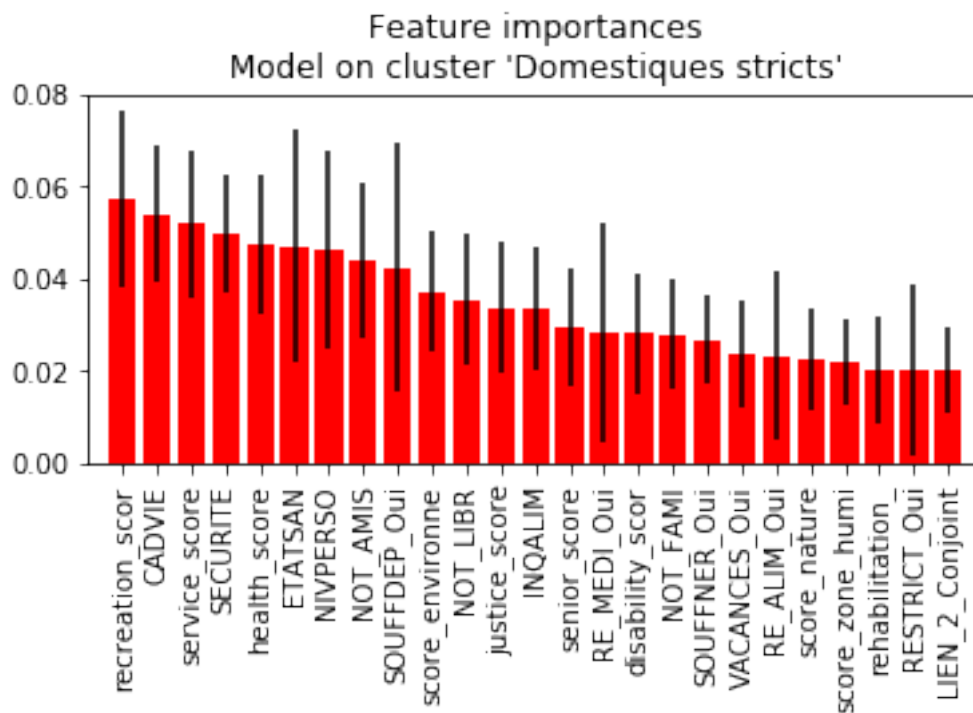
1. feature 13 -recreation_score- (0.060240)
2. feature 9 -NOT_FAMI- (0.048089)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
3. feature 11 -service_score- (0.046739)
4. feature 21 -NIVPERSO- (0.045620)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
5. feature 26 -health_score- (0.045372)
6. feature 19 -NOT_LIBR- (0.044479)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
7. feature 1 -INQALIM- (0.042796)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
8. feature 12 -SECURITE- (0.041377)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
9. feature 28 -NOT_AMIS- (0.040178)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
10. feature 23 -score_environnement- (0.039250)
11. feature 18 -justice_score- (0.037153)
12. feature 27 -SOUFFDEP_Oui- (0.034893)

Actionable at individual level (2)
 Actionable at administrative level (1)
 13. feature 10 -CADVIE- (0.031316)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 14. feature 2 -ETATSAN- (0.031159)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 15. feature 32 -senior_score- (0.030968)
 16. feature 14 -BANQEPA_Oui- (0.029996)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 17. feature 30 -disability_score- (0.028602)
 18. feature 24 -RE_MEDI_Oui- (0.027684)
 Actionable at individual level (2)
 Actionable at administrative level (2)
 19. feature 3 -SITUEMP3_Inactif- (0.027015)
 Actionable at individual level (2)
 Actionable at administrative level (2)
 20. feature 31 -SOUFFNER_Oui- (0.026765)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 21. feature 33 -rehabilitation_score- (0.023318)
 22. feature 22 -score_nature- (0.022155)
 23. feature 8 -SOUFFINS_Oui- (0.021280)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 24. feature 15 -child_score- (0.021161)
 25. feature 16 -score_zone_humide- (0.021134)
 cluster 4 (2359), f1 macro 89.7%
 top 15 features:



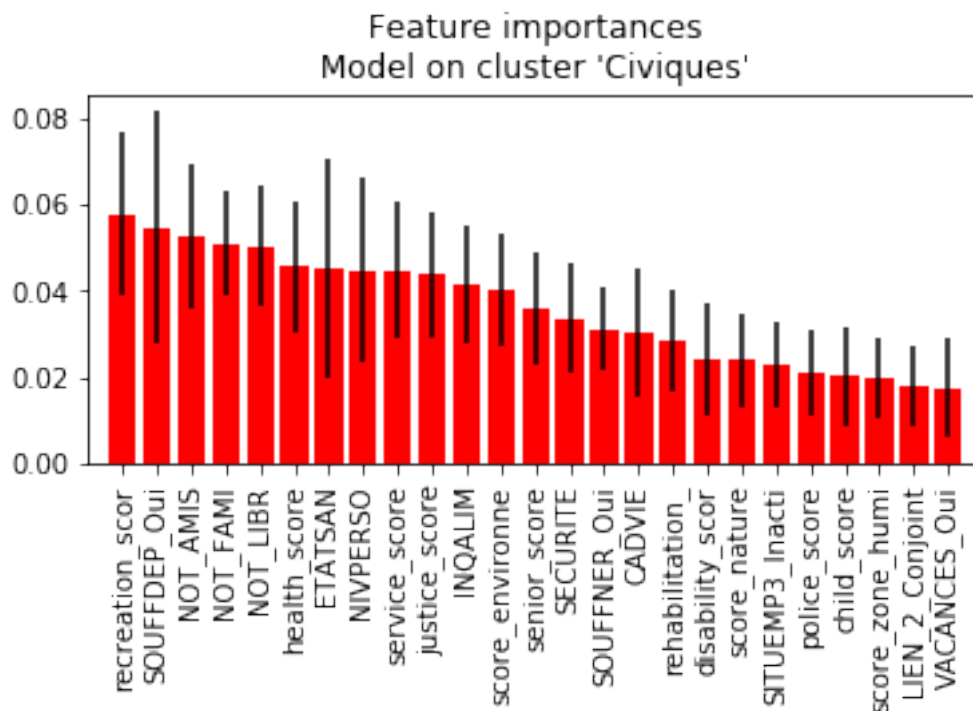
1. feature 13 -recreation_score- (0.060207)
2. feature 2 -ETATSAN- (0.051485)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
3. feature 11 -service_score- (0.050973)
4. feature 21 -NIVPERSO- (0.047533)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
5. feature 19 -NOT_LIBR- (0.045649)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
6. feature 26 -health_score- (0.043495)
7. feature 10 -CADVIE- (0.043246)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
8. feature 27 -SOUFFDEP_Oui- (0.042098)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
9. feature 28 -NOT_AMIS- (0.041891)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
10. feature 23 -score_environnement- (0.040035)
11. feature 1 -INQALIM- (0.035039)
 - Actionable at individual level (1)

Actionable at administrative level (1)
 12. feature 18 -justice_score- (0.034880)
 13. feature 0 -RE_ALIM_Oui- (0.033203)
 Actionable at individual level (2)
 Actionable at administrative level (2)
 14. feature 32 -senior_score- (0.031288)
 15. feature 30 -disability_score- (0.029228)
 16. feature 33 -rehabilitation_score- (0.027566)
 17. feature 20 -VACANCES_Oui- (0.026552)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 18. feature 31 -SOUFFNER_Oui- (0.025747)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 19. feature 12 -SECURITE- (0.025563)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 20. feature 9 -NOT_FAMI- (0.023414)
 Actionable at individual level (1)
 Actionable at administrative level (2)
 21. feature 16 -score_zone_humide- (0.023009)
 22. feature 22 -score_nature- (0.022681)
 23. feature 4 -RE_VOIT_Oui- (0.021207)
 Actionable at individual level (2)
 Actionable at administrative level (2)
 24. feature 14 -BANQEPA_Oui- (0.020110)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 25. feature 15 -child_score- (0.019883)
 cluster 6 (2313), f1 macro 92.5%
 top 15 features:



1. feature 13 -recreation_score- (0.057281)
2. feature 10 -CADVIE- (0.053844)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
3. feature 11 -service_score- (0.051812)
4. feature 12 -SECURITE- (0.049728)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
5. feature 26 -health_score- (0.047156)
6. feature 2 -ETATSAN- (0.047098)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
7. feature 21 -NIVPERSO- (0.046260)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
8. feature 28 -NOT_AMIS- (0.043982)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
9. feature 27 -SOUFFDEP_Oui- (0.042436)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
10. feature 23 -score_environnement- (0.037200)
11. feature 19 -NOT_LIBR- (0.035333)
 - Actionable at individual level (1)

Actionable at administrative level (1)
 12. feature 18 -justice_score- (0.033776)
 13. feature 1 -INQALIM- (0.033651)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 14. feature 32 -senior_score- (0.029387)
 15. feature 24 -RE_MEDI_Oui- (0.028344)
 Actionable at individual level (2)
 Actionable at administrative level (2)
 16. feature 30 -disability_score- (0.028124)
 17. feature 9 -NOT_FAMI- (0.027908)
 Actionable at individual level (1)
 Actionable at administrative level (2)
 18. feature 31 -SOUFFNER_Oui- (0.026827)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 19. feature 20 -VACANCES_Oui- (0.023851)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 20. feature 0 -RE_ALIM_Oui- (0.023308)
 Actionable at individual level (2)
 Actionable at administrative level (2)
 21. feature 22 -score_nature- (0.022559)
 22. feature 16 -score_zone_humide- (0.022062)
 23. feature 33 -rehabilitation_score- (0.020259)
 24. feature 17 -RESTRICT_Oui- (0.020255)
 Actionable at individual level (2)
 Actionable at administrative level (2)
 25. feature 5 -LIEN_2_Conjoint ou compagnon- (0.020177)
 cluster 1 (528), f1 macro 86.0%
 top 15 features:



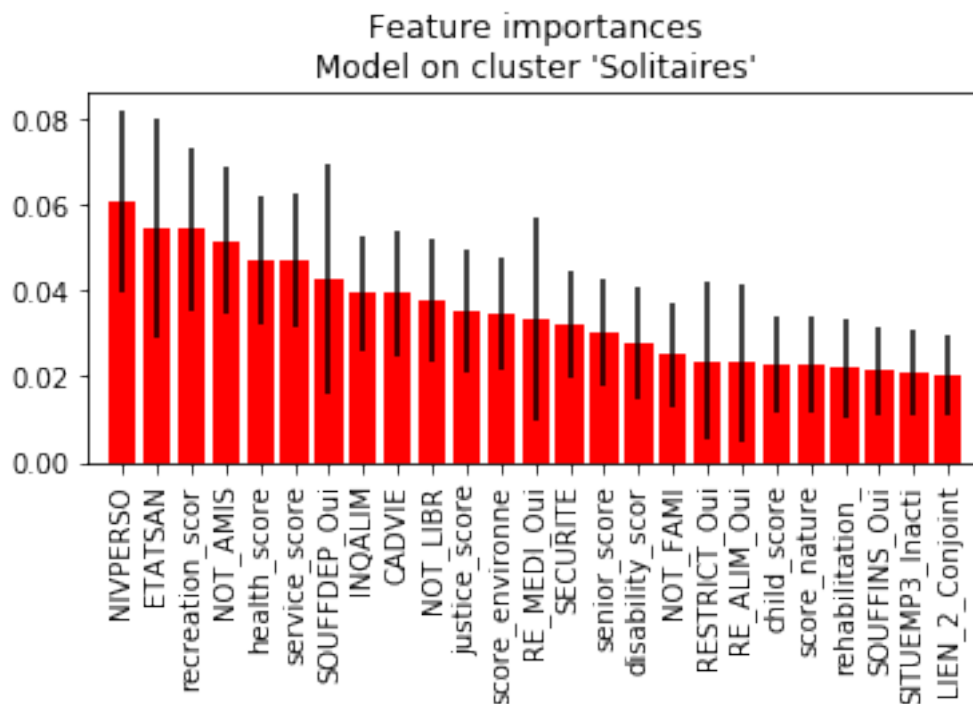
1. feature 13 -recreation_score- (0.057651)
2. feature 27 -SOUFFDEP_Oui- (0.054654)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
3. feature 28 -NOT_AMIS- (0.052522)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
4. feature 9 -NOT_FAMI- (0.050816)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
5. feature 19 -NOT_LIBR- (0.050318)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
6. feature 26 -health_score- (0.045595)
7. feature 2 -ETATSAN- (0.044938)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
8. feature 21 -NIVPERSO- (0.044766)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
9. feature 11 -service_score- (0.044677)
10. feature 18 -justice_score- (0.043685)
11. feature 1 -INQALIM- (0.041439)
 - Actionable at individual level (1)

Actionable at administrative level (1)
 12. feature 23 -score_environnement- (0.040245)
 13. feature 32 -senior_score- (0.035830)
 14. feature 12 -SECURITE- (0.033536)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 15. feature 31 -SOUFFNER_Oui- (0.030934)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 16. feature 10 -CADVIE- (0.030289)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 17. feature 33 -rehabilitation_score- (0.028356)
 18. feature 30 -disability_score- (0.023987)
 19. feature 22 -score_nature- (0.023846)
 20. feature 3 -SITUEMP3_Inactif- (0.022788)
 Actionable at individual level (2)
 Actionable at administrative level (2)
 21. feature 7 -police_score- (0.021120)
 22. feature 15 -child_score- (0.020151)
 23. feature 16 -score_zone_humide- (0.019625)
 24. feature 5 -LIEN_2_Conjoint ou compagnon- (0.017995)
 25. feature 20 -VACANCES_Oui- (0.017589)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 cluster 3 (1384), f1 macro 93.0%
 top 15 features:

Variable	Mean	Standard Deviation
recreation_scor	0.060	0.019
ETATSAN	0.059	0.025
INQALIM	0.046	0.013
NIVPERSO	0.044	0.021
service_scor	0.044	0.015
RESTRICT_Oui	0.040	0.018
CADVIE	0.040	0.015
health_scor	0.039	0.015
NOT_AMIS	0.037	0.017
score_environne	0.037	0.012
NOT_LIBR	0.036	0.016
SOUFFDEP_Oui	0.035	0.026
SECURITE	0.035	0.012
LIEN_2_Conjoint	0.032	0.009
justice_scor	0.031	0.014
SOUFFNER_Oui	0.030	0.010
senior_scor	0.028	0.012
disability_scor	0.027	0.013
NOT_FAMI	0.026	0.012
RE_ALIM_Oui	0.025	0.018
score_zone_humi	0.025	0.009
SOUFFINS_Oui	0.025	0.010
RE_VOIT_Oui	0.024	0.012
rehabilitation	0.022	0.011
VACANCES_Oui	0.022	0.011

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Actionable at administrative level (1)
 12. feature 27 -SOUFFDEP_Oui- (0.034982)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 13. feature 12 -SECURITE- (0.034640)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 14. feature 5 -LIEN_2_Conjoint ou compagnon- (0.031888)
 15. feature 18 -justice_score- (0.030354)
 16. feature 31 -SOUFFNER_Oui- (0.029586)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 17. feature 32 -senior_score- (0.027716)
 18. feature 30 -disability_score- (0.026399)
 19. feature 9 -NOT_FAMI- (0.025206)
 Actionable at individual level (1)
 Actionable at administrative level (2)
 20. feature 0 -RE_ALIM_Oui- (0.024695)
 Actionable at individual level (2)
 Actionable at administrative level (2)
 21. feature 16 -score_zone_humide- (0.024691)
 22. feature 8 -SOUFFINS_Oui- (0.024533)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 23. feature 4 -RE_VOIT_Oui- (0.024299)
 Actionable at individual level (2)
 Actionable at administrative level (2)
 24. feature 33 -rehabilitation_score- (0.022264)
 25. feature 20 -VACANCES_Oui- (0.021980)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 cluster 5 (1494), f1 macro 89.5%
 top 15 features:



1. feature 21 -NIVPERSO- (0.060923)
Actionable at individual level (2)
Actionable at administrative level (2)
2. feature 2 -ETATSAN- (0.054585)
Actionable at individual level (1)
Actionable at administrative level (1)
3. feature 13 -recreation_score- (0.054377)
4. feature 28 -NOT_AMIS- (0.051651)
Actionable at individual level (1)
Actionable at administrative level (2)
5. feature 26 -health_score- (0.047141)
6. feature 11 -service_score- (0.046994)
7. feature 27 -SOUFFDEP_Oui- (0.042860)
Actionable at individual level (2)
Actionable at administrative level (1)
8. feature 1 -INQALIM- (0.039438)
Actionable at individual level (1)
Actionable at administrative level (1)
9. feature 10 -CADVIE- (0.039396)
Actionable at individual level (1)
Actionable at administrative level (1)
10. feature 19 -NOT_LIBR- (0.037676)
Actionable at individual level (1)
Actionable at administrative level (1)

```

11. feature 18 -justice_score- (0.035045)
12. feature 23 -score_environnement- (0.034700)
13. feature 24 -RE_MEDI_Oui- (0.033579)
    Actionable at individual level (2)
    Actionable at administrative level (2)
14. feature 12 -SECURITE- (0.032083)
    Actionable at individual level (2)
    Actionable at administrative level (1)
15. feature 32 -senior_score- (0.030337)
16. feature 30 -disability_score- (0.027743)
17. feature 9 -NOT_FAMI- (0.024980)
    Actionable at individual level (1)
    Actionable at administrative level (2)
18. feature 17 -RESTRICT_Oui- (0.023518)
    Actionable at individual level (2)
    Actionable at administrative level (2)
19. feature 0 -RE_ALIM_Oui- (0.023200)
    Actionable at individual level (2)
    Actionable at administrative level (2)
20. feature 15 -child_score- (0.022877)
21. feature 22 -score_nature- (0.022672)
22. feature 33 -rehabilitation_score- (0.021916)
23. feature 8 -SOUFFINS_Oui- (0.021292)
    Actionable at individual level (2)
    Actionable at administrative level (1)
24. feature 3 -SITUEMP3_Inactif- (0.020953)
    Actionable at individual level (2)
    Actionable at administrative level (2)
25. feature 5 -LIEN_2_Conjoint ou compagnon- (0.020330)

```

1.3 Option 3

In [30]: *# choosing set of features*

```

scope_filter = RFE_RandomForestClassifier_100_features | RFE_LinearSVC_100_features
scope = dict_features_sets.get('insee_environment_score_features', set())
scope = scope | dict_features_sets.get('insee_recreation_score_features', set())
scope = scope | dict_features_sets.get('cdv_actionable_admin_1_features', set())
scope = scope | dict_features_sets.get('cdv_actionable_individual_1_features', set())
scope = scope | RFE_LogisticRegression_20_features
scope = scope & scope_filter
print(f"number of features : {len(scope)} ow actionable")
A = scope & dict_features_sets.get('cdv_actionable_admin_1_features', set())
B = scope & dict_features_sets.get('cdv_actionable_individual_1_features', set())
print(f"- at administrative level 1 : \t{len(A)}\n- at individual level 1 : \t{len(B)}")

```

```

number of features : 40 ow actionable
- at administrative level 1 :          29

```

- at individual level 1 : 26

```
In [31]: df = dataset.loc[:,:]
         # reducing problem to a 2 class classification problem
         df["HEUREUX_CLF"] = 0
         df.loc[df["HEUREUX"]==4, "HEUREUX_CLF"] = 1
         df.loc[df["HEUREUX"]==3, "HEUREUX_CLF"] = 1
         df.loc[df["HEUREUX"]==5, "HEUREUX_CLF"] = None

         scope = scope & set(dataset.columns)
         n_max = 2000

         df = df.loc[:,scope | {"HEUREUX_CLF"} ].dropna()
         features = df.loc[:,scope ].columns

         X = df.loc[:,scope]
         y = df["HEUREUX_CLF"]

         Xs, ys = resample(X, y, random_state=42)

         Xs = Xs.iloc[0:n_max,:]
         ys = ys.iloc[0:n_max]

         X_train, X_test, y_train, y_test = train_test_split(Xs, ys,
                                                             test_size=0.2,
                                                             random_state=42
                                                             )

         scaler = StandardScaler().fit(X_train)
         X_train = scaler.transform(X_train)
         X_test = scaler.transform(X_test)

         print(f"Number exemple: {y.shape[0]}\n- training set: \
               {y_train.shape[0]}\n- test set: {y_test.shape[0]}")
         print(f"Number of features: p={X_train.shape[1]}")
         print(f"Number of class: {len(np.unique(y))}")
         for c in np.unique(y):
             print(f"class {c:0.0f} : {100*np.sum(y==c)/len(y):0.1f}%")

Number exemple: 10605
- training set: 1600
- test set: 400
Number of features: p=40
Number of class: 2
class 0 : 34.9%
class 1 : 65.1%
```

```

In [32]: startTime = time.time()
n_estimators_range = [32,64,128,256,512]
max_depth_range = [4,8,16,32,64]
param_grid = dict(n_estimators=n_estimators_range, max_depth = max_depth_range)

params = {'max_features' : 'sqrt', 'random_state' : 32,
          'min_samples_split' : 2, 'class_weight' : 'balanced'}
clf = RandomForestClassifier(**params)

grid = GridSearchCV(clf, scoring='accuracy', param_grid=param_grid)
grid.fit(X_train, y_train)
print(f"Determination of optimal hyperparameters in {time.time() - startTime:0.1f} s")
print(f"Optimal values are {grid.best_params_} \n\
Accuracy Score of cross validation {100*grid.best_score_:0.2f}%")

# Learning on full training set with optimals hyperparameters and score on test set
params = {'max_features' : 'sqrt', 'random_state' : 32,
          'min_samples_split' : 2, 'class_weight' : 'balanced',
          'n_estimators' : grid.best_params_['n_estimators'],
          'max_depth' : grid.best_params_['max_depth']}
clf = RandomForestClassifier(**params).fit(X_train, y_train)
clf.fit(X_train, y_train)
y_test_pred = clf.predict(X_test)

print(f"Random Forest, p={X_train.shape[1]}")
accuracy = clf.score(X_test, y_test)
f1 = f1_score(y_test, y_test_pred)
p = precision_score(y_test, y_test_pred)
r = recall_score(y_test, y_test_pred)
print(f"Model score\n- Accuracy : {accuracy*100:0.1f} %")
print(f"- Precision : {p*100:0.1f} % (Happy # positive class)")
print(f"- Recall : {r*100:0.1f} %")
print(f"- F1 score : {f1*100:0.1f} %")
res_full = {
    'f1_score' : f1,
    'accuracy' : accuracy,
    'precision' : p,
    'recall' : r
}

```

Determination of optimal hyperparameters in 61.3 s
Optimal values are {'max_depth': 16, 'n_estimators': 512}
Accuracy Score of cross validation 75.88%
Random Forest, p=40
Model score
- Accuracy : 73.0 %
- Precision : 75.3 % (Happy # positive class)
- Recall : 85.4 %

- F1 score : 80.0 %

```
In [33]: importances = clf.feature_importances_
std = np.std([tree.feature_importances_ for tree in clf.estimators_],
              axis=0)
indices = np.argsort(importances)[::-1]
features_name = np.array(features)
features_short_name_sorted = [ name[:15] for name in features_name[indices]]
print("Feature ranking:")

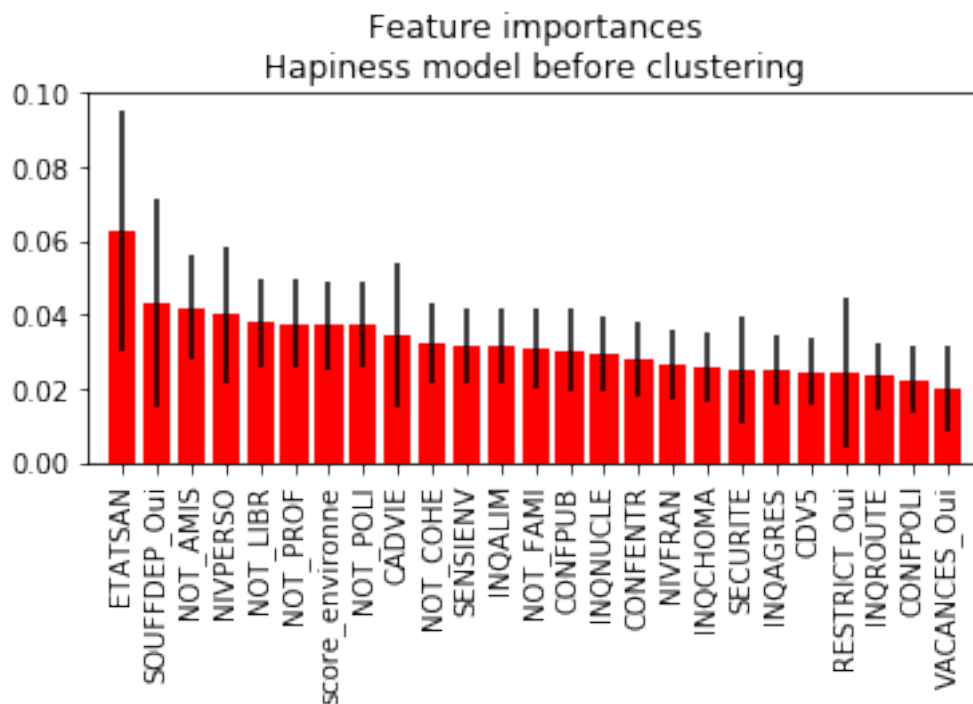
n_features_max = 25
n_features = min(X.shape[1],n_features_max)

# Plot the feature importances of the forest

plt.figure()
plt.gcf().subplots_adjust(bottom=0.4)
plt.title("Feature importances\nHapiness model before clustering")
plt.bar(range(n_features), importances[indices][:n_features],
        color="r", yerr=std[indices[:n_features]], align="center")
plt.xticks(range(n_features), features_short_name_sorted[:n_features], rotation=90)
plt.xlim([-1, n_features])
filename = path_fig / Path("feature_importance_option3.jpg")
plt.savefig(filename, format='jpg')
plt.show()

for f in range(min(X.shape[1],n_features_max)):
    print("%d. feature %d -%s- (%f)" % (f + 1, indices[f],features_name[indices[f]], im
    if features_name[indices[f]] in actionable_individual_1_features:
        print("\tActionable at individual level (1)")
    if features_name[indices[f]] in actionable_individual_2_features:
        print("\tActionable at individual level (2)")
    if features_name[indices[f]] in actionable_admin_1_features:
        print("\tActionable at administrative level (1)")
    if features_name[indices[f]] in actionable_admin_2_features:
        print("\tActionable at administrative level (2)"))
```

Feature ranking:



1. feature 3 -ETATSAN- (0.062689)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
2. feature 35 -SOUFFDEP_Oui- (0.042963)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
3. feature 36 -NOT_AMIS- (0.041908)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
4. feature 27 -NIVPERSO- (0.039819)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
5. feature 24 -NOT_LIBR- (0.037761)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
6. feature 26 -NOT_PROF- (0.037372)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
7. feature 28 -score_environnement- (0.037163)
8. feature 33 -NOT_POLI- (0.037133)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
9. feature 16 -CADVIE- (0.034626)
 - Actionable at individual level (1)

- Actionable at administrative level (1)
- 10. feature 31 -NOT_COHE- (0.032463)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 11. feature 32 -SENSIENV- (0.031718)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 12. feature 2 -INQALIM- (0.031470)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 13. feature 14 -NOT_FAMI- (0.030716)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
- 14. feature 6 -CONFPUB- (0.030303)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 15. feature 39 -INQNUCLE- (0.029431)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 16. feature 21 -CONFENTR- (0.027902)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 17. feature 8 -NIVFRAN- (0.026394)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
- 18. feature 20 -INQCHOMA- (0.025716)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 19. feature 17 -SECURITE- (0.025106)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
- 20. feature 9 -INQAGRES- (0.024943)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 21. feature 7 -CDV5- (0.024466)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
- 22. feature 23 -RESTRICT_Oui- (0.023934)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
- 23. feature 11 -INQROUTE- (0.023345)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 24. feature 30 -CONFPOLI- (0.022245)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 25. feature 25 -VACANCES_Oui- (0.020232)
 - Actionable at individual level (1)

Actionable at administrative level (1)

```
In [34]: print(f"number of features : {len(scope)} ow actionnable")
        A = scope & dict_features_sets.get('cdv_actionable_admin_1_features',set())
        B = scope & dict_features_sets.get('cdv_actionable_individual_1_features',set())
        print(f"- at administrative level 1 : \t{len(A)}\n- at individual level 1 : \t{len(B)}")
        important_features = set(features_name[indices][:10])
        C = A & important_features
        D = B & important_features
        print(f"- at administrative level 1 in top 10: \t{len(C)}\n- at individual level 1 in t
```

```
number of features : 40 ow actionnable
- at administrative level 1 :          29
- at individual level 1 :          26
- at administrative level 1 in top 10:          6
- at individual level 1 in top 10:          7
```

```
In [35]: n_estimators_range = [16,32,64,128]
        max_depth_range = [2,4,8,16,32,64]
        param_grid = dict(n_estimators=n_estimators_range, max_depth = max_depth_range)
        params = {'max_features' : 'sqrt',
                  'random_state' : 32,
                  'min_samples_split' : 2,
                  'class_weight' : 'balanced'
                  }
        #scope = ( SelectFromModel_LogisticRegression_features ) & set(dataset.columns)
        features = df.loc[:,scope].columns
```

```
In [36]: score_clustering_methods = []
        clustering_methods = clustTest1.columns[2:3]

        for method in clustering_methods:
            print("-----")
            print(f"\nAnalysis cluster method {method}")
            cluster_list = clustTest1[method].unique()
            print(f"liste of clusters : {cluster_list}")
            score_cluster = []
            for cluster in cluster_list:
                index_scope = clustTest1.loc[clustTest1[method]==cluster,:].index
                print(f"cluster {cluster} '{cluster_name[cluster]}' : {len(index_scope)} elements")

                Xc = X.loc[index_scope.intersection(X.index),:]
                yc = y[index_scope.intersection(X.index)]

                Xs, ys = resample(Xc, yc, random_state=42)

                Xs = Xs.iloc[0:n_max,:]
```



```

ys = ys.iloc[0:n_max]

X_train, X_test, y_train, y_test = train_test_split(Xs, ys,
                                                    test_size=0.2,
                                                    random_state=42)

scaler = StandardScaler().fit(X_train)
X_train = scaler.transform(X_train)
X_test = scaler.transform(X_test)

print(f"Number exemple: {ys.shape[0]}\n\
- training set: {y_train.shape[0]}\n\
- test set: {y_test.shape[0]}")
print(f"Number of features: p={X_train.shape[1]}")
print(f"Number of class: {len(np.unique(y))}")
for c in np.unique(y):
    print(f"class {c:0.0f} : {100*np.sum(yc==c)/len(yc):0.1f}%")

startTime = time.time()
clf = RandomForestClassifier(**params)
grid = GridSearchCV(clf,
                    scoring='accuracy',
                    param_grid=param_grid)

grid.fit(X_train, y_train)
print(f"Optimal values are {grid.best_params_} \n\
cross validation score {100*grid.best_score_:0.2f}%")
print()

# Learning on full training set with optimals hyperparameters and score on test
params_opt = {'max_features' : 'sqrt', 'random_state' : 32,
              'min_samples_split' : 2, 'class_weight' : 'balanced',
              'n_estimators' : grid.best_params_['n_estimators'],
              'max_depth' : grid.best_params_['max_depth']}
clf = RandomForestClassifier(**params_opt).fit(X_train, y_train)

y_test_pred = clf.predict(X_test)
accuracy = clf.score(X_test, y_test)
f1 = f1_score(y_test, y_test_pred)
p = precision_score(y_test, y_test_pred)
r = recall_score(y_test, y_test_pred)

res = {'f1_score' : f1,
       'accuracy' : accuracy,
       'precision' : p,
       'recall' : r}

```

```

importances = clf.feature_importances_
std = np.std([tree.feature_importances_ for tree in clf.estimators_],
              axis=0)
indices = np.argsort(importances)[::-1]
features_name = np.array(features)

cl = {'cluster' : cluster,
      'size' : len(index_scope),
      'model' : 'RandomForestClassifier',
      'params' : params_opt,
      'metrics' : res,
      'importances' : importances,
      'std' : std,
      'indices' : indices,
      'features_name' : features_name
      }

score_cluster.append(cl)

d = {'clustering_method' : method,
     'cluster_scores' : score_cluster
     }
score_clustering_methods.append(d)

```

```

-----

Analysis cluster method clust3
liste of clusters : [2 4 6 1 3 5]
cluster 2 'Equilibrés' : 3053 elements
Number exemple: 2000
    - training set: 1600
    - test set: 400
Number of features: p=40
Number of class: 2
class 0 : 34.6%
class 1 : 65.4%
Optimal values are {'max_depth': 16, 'n_estimators': 64}
cross validation score 81.94%

cluster 4 'Domestiques modérés' : 2359 elements
Number exemple: 2000
    - training set: 1600
    - test set: 400
Number of features: p=40
Number of class: 2

```

class 0 : 32.6%
class 1 : 67.4%
Optimal values are {'max_depth': 16, 'n_estimators': 128}
cross validation score 83.94%

cluster 6 'Domestiques stricts' : 2313 elements
Number exemple: 2000
- training set: 1600
- test set: 400
Number of features: p=40
Number of class: 2
class 0 : 32.8%
class 1 : 67.2%
Optimal values are {'max_depth': 32, 'n_estimators': 128}
cross validation score 83.12%

cluster 1 'Civiques' : 528 elements
Number exemple: 468
- training set: 374
- test set: 94
Number of features: p=40
Number of class: 2
class 0 : 50.2%
class 1 : 49.8%
Optimal values are {'max_depth': 8, 'n_estimators': 64}
cross validation score 73.80%

cluster 3 'Flâneurs' : 1384 elements
Number exemple: 1350
- training set: 1080
- test set: 270
Number of features: p=40
Number of class: 2
class 0 : 31.7%
class 1 : 68.3%
Optimal values are {'max_depth': 16, 'n_estimators': 64}
cross validation score 85.37%

cluster 5 'Solitaires' : 1494 elements
Number exemple: 1447
- training set: 1157
- test set: 290
Number of features: p=40
Number of class: 2
class 0 : 40.6%
class 1 : 59.4%
Optimal values are {'max_depth': 32, 'n_estimators': 64}
cross validation score 84.44%

```
In [37]: # F1 score
for score_method in score_clustering_methods:
    print(f"method {score_method['clustering_method']}:")
    average_score = 0
    total_size = 0
    for i, score_cluster in enumerate(score_method['cluster_scores']):
        print(f"cluster {score_cluster['cluster']} '{cluster_name[score_cluster['cluster_name']]}'")
        average_score += score_cluster['metrics']['f1_score']*score_cluster['size']
        total_size += score_cluster['size']

    average_score = average_score / total_size
    print(f"average f1 on clusters {100*average_score:0.1f}% gain {100*(average_score-r

method clust3:
cluster 2 'Equilibrés' : (3053), f1 macro 85.6%
cluster 4 'Domestiques modérés' : (2359), f1 macro 90.5%
cluster 6 'Domestiques stricts' : (2313), f1 macro 90.4%
cluster 1 'Civiques' : (528), f1 macro 92.1%
cluster 3 'Flâneurs' : (1384), f1 macro 95.4%
cluster 5 'Solitaires' : (1494), f1 macro 87.8%
average f1 on clusters 89.5% gain 9.5
```

```
In [38]: # accuracy
for score_method in score_clustering_methods:
    print(f"method {score_method['clustering_method']}:")
    average_score = 0
    total_size = 0
    for i, score_cluster in enumerate(score_method['cluster_scores']):
        print(f"cluster {score_cluster['cluster']} '{cluster_name[score_cluster['cluster_name']]}'")
        average_score = average_score + score_cluster['metrics']['accuracy']*score_cluster['size']
        total_size += score_cluster['size']
    average_score = average_score / total_size
    print(f"average accuracy on clusters {100*average_score:0.1f}% gain {100*(average_s

method clust3:
cluster 2 'Equilibrés' : (3053) accuracy 81.5%
cluster 4 'Domestiques modérés' : (2359) accuracy 86.8%
cluster 6 'Domestiques stricts' : (2313) accuracy 86.2%
cluster 1 'Civiques' : (528) accuracy 92.6%
cluster 3 'Flâneurs' : (1384) accuracy 93.3%
cluster 5 'Solitaires' : (1494) accuracy 84.5%
average accuracy on clusters 86.0% gain 13.0
```

```

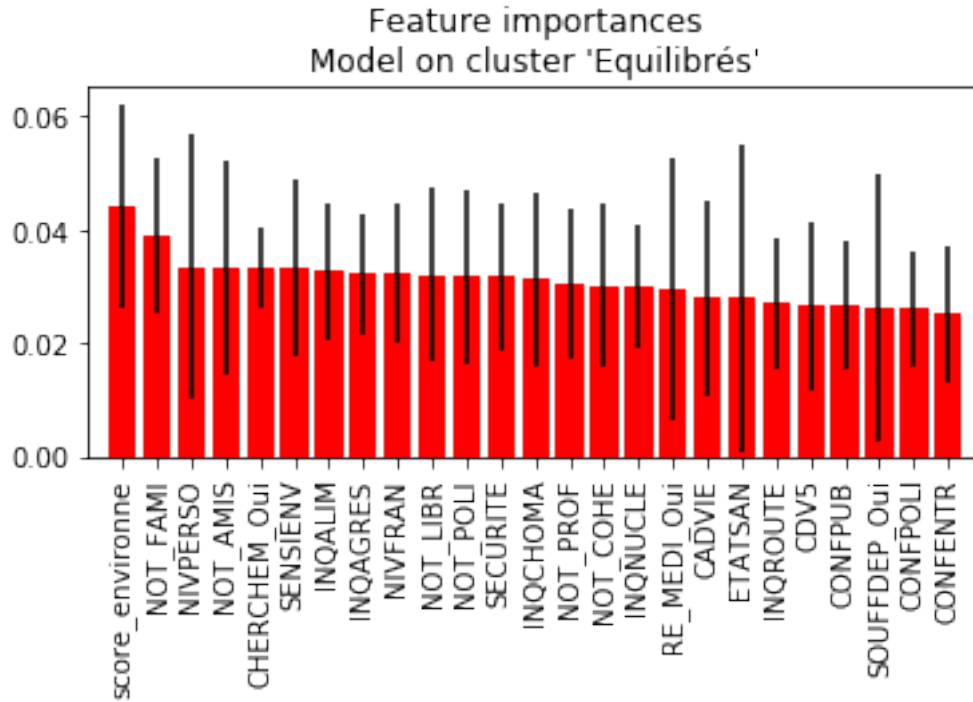
In [39]: # Feature importance by cluster
for score_method in score_clustering_methods:
    print(f"method {score_method['clustering_method']}:")
    for i, score_cluster in enumerate(score_method['cluster_scores']):
        print(f"cluster {score_cluster['cluster']} ({score_cluster['size']}), f1 macro")
        print(f"top 15 features:")
        indices = score_cluster['indices']
        features_name = score_cluster['features_name']
        importances = score_cluster['importances']
        features_short_name_sorted = [ name[:15] for name in features_name[indices]]

        # Plot the feature importances of the forest
        plt.figure()
        plt.gcf().subplots_adjust(bottom=0.4)
        plt.title(f"Feature importances\nModel on cluster '{cluster_name[score_cluster['cluster_name']]}'")
        plt.bar(range(n_features), importances[indices][:n_features],
                color="r", yerr=std[indices][:n_features], align="center")
        plt.xticks(range(n_features), features_short_name_sorted[:n_features], rotation=45)
        plt.xlim([-1, n_features])
        filename = path_fig / Path(f"feature_importance_cluster_{score_cluster['cluster_name']}")
        plt.savefig(filename, format='jpg')
        plt.show()

    for f in range(n_features_max):
        print("%d. feature %d -%s- (%f)" % (f + 1, indices[f], features_name[indices[f]], importances[indices[f]]))
        if features_name[indices[f]] in actionable_individual_1_features:
            print("\tActionable at individual level (1)")
        if features_name[indices[f]] in actionable_individual_2_features:
            print("\tActionable at individual level (2)")
        if features_name[indices[f]] in actionable_admin_1_features:
            print("\tActionable at administrative level (1)")
        if features_name[indices[f]] in actionable_admin_2_features:
            print("\tActionable at administrative level (2)")

method clust3:
cluster 2 (3053), f1 macro 85.6%
top 15 features:

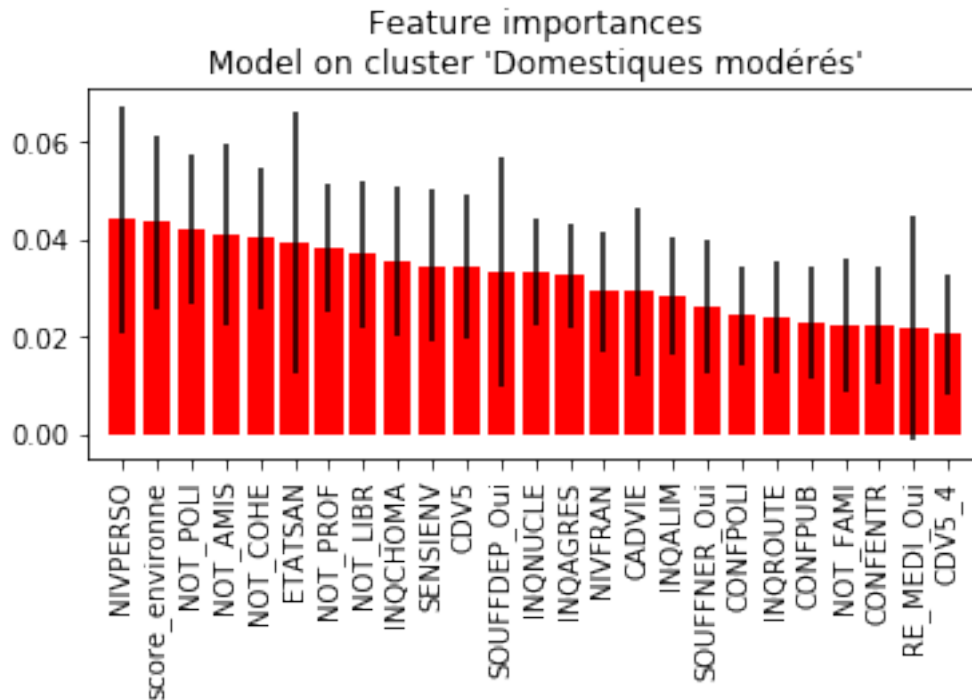
```



1. feature 28 -score_environne- (0.044090)
2. feature 14 -NOT_FAMI- (0.038867)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
3. feature 27 -NIV_PERSO- (0.033454)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
4. feature 36 -NOT_AMIS- (0.033307)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
5. feature 22 -CHERCHEM_Oui- (0.033212)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
6. feature 32 -SENSIENV- (0.033058)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
7. feature 2 -INQALIM- (0.032553)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
8. feature 9 -INQAGRES- (0.032290)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
9. feature 8 -NIVFRAN- (0.032261)
 - Actionable at individual level (2)

Actionable at administrative level (1)
 10. feature 24 -NOT_LIBR- (0.031972)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 11. feature 33 -NOT_POLI- (0.031706)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 12. feature 17 -SECURITE- (0.031628)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 13. feature 20 -INQCHOMA- (0.031245)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 14. feature 26 -NOT_PROF- (0.030478)
 Actionable at individual level (1)
 Actionable at administrative level (2)
 15. feature 31 -NOT_COHE- (0.030046)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 16. feature 39 -INQNUCLE- (0.029950)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 17. feature 29 -RE_MEDI_Oui- (0.029334)
 Actionable at individual level (2)
 Actionable at administrative level (2)
 18. feature 16 -CADVIE- (0.027998)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 19. feature 3 -ETATSAN- (0.027841)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 20. feature 11 -INQROUTE- (0.026949)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 21. feature 7 -CDV5- (0.026573)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 22. feature 6 -CONFPUB- (0.026497)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 23. feature 35 -SOUFFDEP_Oui- (0.026268)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 24. feature 30 -CONFPOLI- (0.025985)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 25. feature 21 -CONFENTR- (0.025161)
 Actionable at individual level (1)

Actionable at administrative level (1)
cluster 4 (2359), f1 macro 90.5%
top 15 features:



1. feature 27 -NIVPERSO- (0.043900)
Actionable at individual level (2)
Actionable at administrative level (2)
2. feature 28 -score_environnement- (0.043421)
3. feature 33 -NOT_POLI- (0.041819)
Actionable at individual level (1)
Actionable at administrative level (1)
4. feature 36 -NOT_AMIS- (0.040947)
Actionable at individual level (1)
Actionable at administrative level (2)
5. feature 31 -NOT_COHE- (0.040179)
Actionable at individual level (1)
Actionable at administrative level (1)
6. feature 3 -ETATSAN- (0.039159)
Actionable at individual level (1)
Actionable at administrative level (1)
7. feature 26 -NOT_PROF- (0.038041)
Actionable at individual level (1)
Actionable at administrative level (2)

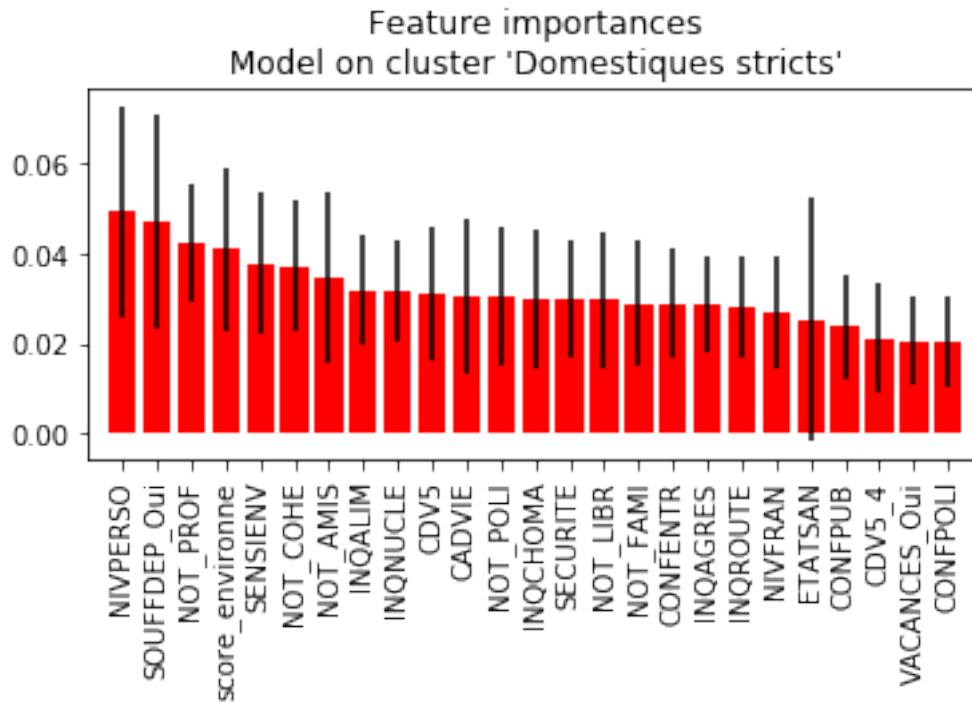
8. feature 24 -NOT_LIBR- (0.036783)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
9. feature 20 -INQCHOMA- (0.035285)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
10. feature 32 -SENSIENV- (0.034527)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
11. feature 7 -CDV5- (0.034380)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
12. feature 35 -SOUFFDEP_Oui- (0.033361)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
13. feature 39 -INQNUCLE- (0.033274)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
14. feature 9 -INQAGRES- (0.032460)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
15. feature 8 -NIVFRAN- (0.029314)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
16. feature 16 -CADVIE- (0.029248)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
17. feature 2 -INQALIM- (0.028259)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
18. feature 38 -SOUFFNER_Oui- (0.026139)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
19. feature 30 -CONFPOLI- (0.024280)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
20. feature 11 -INQROUTE- (0.023849)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
21. feature 6 -CONFPUB- (0.022800)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
22. feature 14 -NOT_FAMI- (0.022355)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
23. feature 21 -CONFENTR- (0.021992)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)

24. feature 29 -RE_MEDI_Oui- (0.021655)
 Actionable at individual level (2)
 Actionable at administrative level (2)

25. feature 4 -CDV5_4- (0.020346)
 Actionable at individual level (2)
 Actionable at administrative level (1)

cluster 6 (2313), f1 macro 90.4%

top 15 features:



1. feature 27 -NIVPERSO- (0.048936)
 Actionable at individual level (2)
 Actionable at administrative level (2)

2. feature 35 -SOUFFDEP_Oui- (0.046825)
 Actionable at individual level (2)
 Actionable at administrative level (1)

3. feature 26 -NOT_PROF- (0.042092)
 Actionable at individual level (1)
 Actionable at administrative level (2)

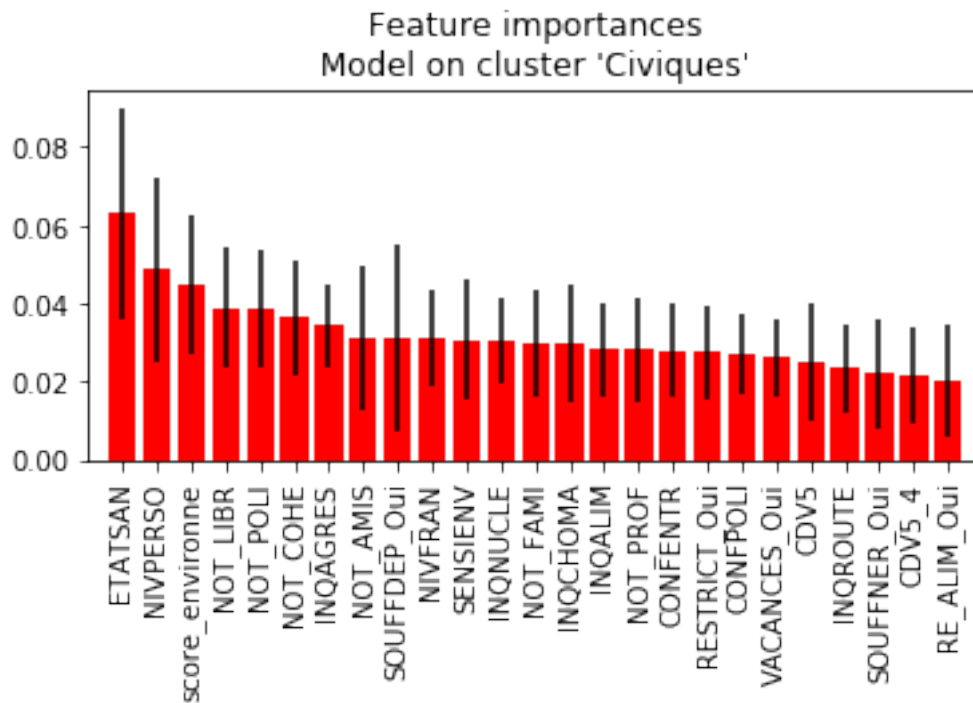
4. feature 28 -score_environnement- (0.040757)

5. feature 32 -SENSIENV- (0.037539)
 Actionable at individual level (1)
 Actionable at administrative level (1)

6. feature 31 -NOT_COHE- (0.036953)

- Actionable at individual level (1)
 - Actionable at administrative level (1)
- 7. feature 36 -NOT_AMIS- (0.034509)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
- 8. feature 2 -INQALIM- (0.031748)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 9. feature 39 -INQNUCLE- (0.031542)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 10. feature 7 -CDV5- (0.031029)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
- 11. feature 16 -CADVIE- (0.030458)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 12. feature 33 -NOT_POLI- (0.030329)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 13. feature 20 -INQCHOMA- (0.029968)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 14. feature 17 -SECURITE- (0.029904)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
- 15. feature 24 -NOT_LIBR- (0.029460)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 16. feature 14 -NOT_FAMI- (0.028673)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
- 17. feature 21 -CONFENTR- (0.028642)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 18. feature 9 -INQAGRES- (0.028567)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 19. feature 11 -INQROUTE- (0.027896)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 20. feature 8 -NIVFRAN- (0.026827)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
- 21. feature 3 -ETATSAN- (0.025247)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 22. feature 6 -CONFPUB- (0.023631)

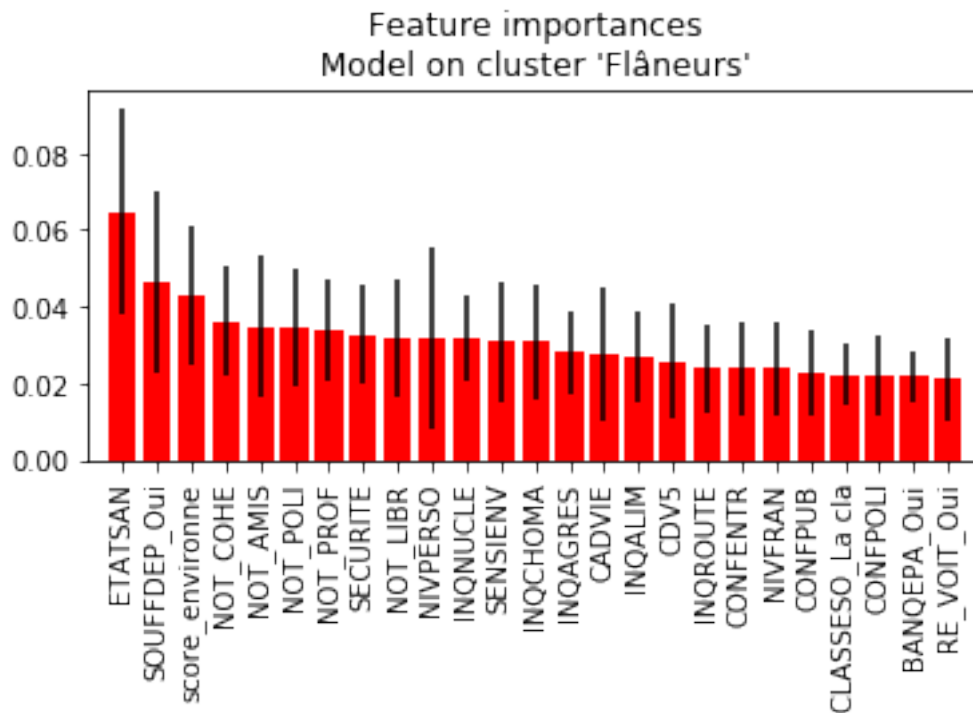
Actionable at individual level (1)
 Actionable at administrative level (1)
 23. feature 4 -CDV5_4- (0.021140)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 24. feature 25 -VACANCES_Oui- (0.020555)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 25. feature 30 -CONFPOLI- (0.020254)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 cluster 1 (528), f1 macro 92.1%
 top 15 features:



1. feature 3 -ETATSAN- (0.063048)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 2. feature 27 -NIVPERSO- (0.048830)
 Actionable at individual level (2)
 Actionable at administrative level (2)
 3. feature 28 -score_environnement- (0.044799)
 4. feature 24 -NOT_LIBR- (0.038897)
 Actionable at individual level (1)

- Actionable at administrative level (1)
- 5. feature 33 -NOT_POLI- (0.038633)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 6. feature 31 -NOT_COHE- (0.036256)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 7. feature 9 -INQAGRES- (0.034210)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 8. feature 36 -NOT_AMIS- (0.031163)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
- 9. feature 35 -SOUFFDEP_Oui- (0.031066)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
- 10. feature 8 -NIVFRAN- (0.030870)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
- 11. feature 32 -SENSIENV- (0.030765)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 12. feature 39 -INQNUCLE- (0.030399)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 13. feature 14 -NOT_FAMI- (0.029980)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
- 14. feature 20 -INQCHOMA- (0.029647)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 15. feature 2 -INQALIM- (0.028161)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 16. feature 26 -NOT_PROF- (0.028152)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
- 17. feature 21 -CONFENTR- (0.027804)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 18. feature 23 -RESTRICT_Oui- (0.027385)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
- 19. feature 30 -CONFPOLI- (0.027211)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
- 20. feature 25 -VACANCES_Oui- (0.026125)
 - Actionable at individual level (1)

Actionable at administrative level (1)
 21. feature 7 -CDV5- (0.025027)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 22. feature 11 -INQROUTE- (0.023451)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 23. feature 38 -SOUFFNER_Oui- (0.021919)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 24. feature 4 -CDV5_4- (0.021575)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 25. feature 1 -RE_ALIM_Oui- (0.020150)
 Actionable at individual level (2)
 Actionable at administrative level (2)
 cluster 3 (1384), f1 macro 95.4%
 top 15 features:



1. feature 3 -ETATSAN- (0.064748)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 2. feature 35 -SOUFFDEP_Oui- (0.046520)

Actionable at individual level (2)
 Actionable at administrative level (1)
 3. feature 28 -score_environnement- (0.042737)
 4. feature 31 -NOT_COHE- (0.036064)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 5. feature 36 -NOT_AMIS- (0.034809)
 Actionable at individual level (1)
 Actionable at administrative level (2)
 6. feature 33 -NOT_POLI- (0.034676)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 7. feature 26 -NOT_PROF- (0.033644)
 Actionable at individual level (1)
 Actionable at administrative level (2)
 8. feature 17 -SECURITE- (0.032646)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 9. feature 24 -NOT_LIBR- (0.031848)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 10. feature 27 -NIVPERSO- (0.031829)
 Actionable at individual level (2)
 Actionable at administrative level (2)
 11. feature 39 -INQNUCLE- (0.031628)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 12. feature 32 -SENSIENV- (0.030718)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 13. feature 20 -INQCHOMA- (0.030694)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 14. feature 9 -INQAGRES- (0.027941)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 15. feature 16 -CADVIE- (0.027490)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 16. feature 2 -INQALIM- (0.027016)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 17. feature 7 -CDV5- (0.025719)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 18. feature 11 -INQROUTE- (0.023945)
 Actionable at individual level (1)
 Actionable at administrative level (1)

19. feature 21 -CONFENTR- (0.023823)
 Actionable at individual level (1)
 Actionable at administrative level (1)

20. feature 8 -NIVFRAN- (0.023807)
 Actionable at individual level (2)
 Actionable at administrative level (1)

21. feature 6 -CONFPUB- (0.022498)
 Actionable at individual level (1)
 Actionable at administrative level (1)

22. feature 15 -CLASSES0_La classe moyenne supérieure- (0.022294)
 Actionable at individual level (1)
 Actionable at administrative level (1)

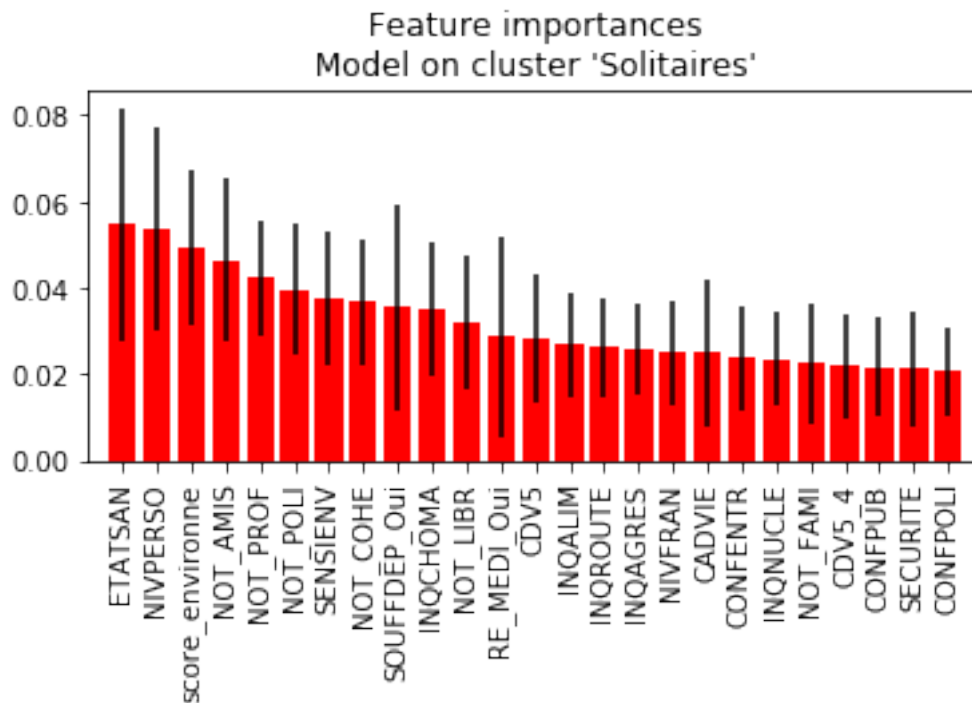
23. feature 30 -CONFPOLI- (0.021941)
 Actionable at individual level (1)
 Actionable at administrative level (1)

24. feature 18 -BANQEPA_Oui- (0.021660)
 Actionable at individual level (1)
 Actionable at administrative level (1)

25. feature 5 -RE_VOIT_Oui- (0.020998)
 Actionable at individual level (2)
 Actionable at administrative level (2)

cluster 5 (1494), f1 macro 87.8%

top 15 features:



1. feature 3 -ETATSAN- (0.054595)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
2. feature 27 -NIVPERSO- (0.053860)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
3. feature 28 -score_environnement- (0.048987)
4. feature 36 -NOT_AMIS- (0.046352)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
5. feature 26 -NOT_PROF- (0.042180)
 - Actionable at individual level (1)
 - Actionable at administrative level (2)
6. feature 33 -NOT_POLI- (0.039579)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
7. feature 32 -SENSIENV- (0.037516)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
8. feature 31 -NOT_COHE- (0.036678)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
9. feature 35 -SOUFFDEP_Oui- (0.035340)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
10. feature 20 -INQCHOMA- (0.034935)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
11. feature 24 -NOT_LIBR- (0.031904)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
12. feature 29 -RE_MEDI_Oui- (0.028559)
 - Actionable at individual level (2)
 - Actionable at administrative level (2)
13. feature 7 -CDV5- (0.028030)
 - Actionable at individual level (2)
 - Actionable at administrative level (1)
14. feature 2 -INQALIM- (0.026677)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
15. feature 11 -INQROUTE- (0.026259)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
16. feature 9 -INQAGRES- (0.025731)
 - Actionable at individual level (1)
 - Actionable at administrative level (1)
17. feature 8 -NIVFRAN- (0.024799)
 - Actionable at individual level (2)

Actionable at administrative level (1)
 18. feature 16 -CADVIE- (0.024798)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 19. feature 21 -CONFENTR- (0.023701)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 20. feature 39 -INQNUCLE- (0.023500)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 21. feature 14 -NOT_FAMI- (0.022355)
 Actionable at individual level (1)
 Actionable at administrative level (2)
 22. feature 4 -CDV5_4- (0.021720)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 23. feature 6 -CONFPUB- (0.021568)
 Actionable at individual level (1)
 Actionable at administrative level (1)
 24. feature 17 -SECURITE- (0.021161)
 Actionable at individual level (2)
 Actionable at administrative level (1)
 25. feature 30 -CONFPOLI- (0.020678)
 Actionable at individual level (1)
 Actionable at administrative level (1)