

ESSeRE Lab
(Evolution of Software SystEms and Reverse Engineering Laboratory)

# WEKA nose Tutorial



allows weka to smell your code

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#### 1 What is it for?

WekaNose is environment which allows to perform an experiment, that have as goal to studies the code smell detection throught machine learning tecniques. This experiment is divided in two main part: (i) the first one concern the creation of the dataset and (ii) the second part where the machine learning algorithms are trained and tested using the dataset created in the first part.

#### 1.1 Dataset Creation

The first part can be separated in six steps, three of which can be performed using WekaNose.

#### • Choose the code smell and generate a definition for that

The first thing to be done is select which code smell you want to take in consideration and create a definition for it, hopefully using some literature. This initial step is very important because the definition will affect the entire experiment, so make sure that the definition underline the code smells characteristics on which you want to focus the most;

# Collection of a large repository of heterogeneous software systems written in Java

The next step is to collect some software systems that will be analyzed and from which the instances of the dataset can be extracted. The systems have to be the most heterogeneous you can find (or create), so that the instances wont refers mostly to the some type of class (o method) that you can find in specific type of software system. An exemple of it is QualitasCorpus (http://qualitascorpus.com/);

#### • Metrics Extraction

Then you have to extract a large set of object-oriented metrics from systems at class, method, package and project levels. The metrics are considered as independent variables in the machine learning approach. The selected metrics are at class, method, package levels. This step is performed by WekaNose using DFMC4J (Design Features and Metrics for Java) that is part of a bigger tool called JCodeOdor developed by the ESSeRe Lab, for more information about it http://essere.disco.unimib.it/wiki/jcodeodor.

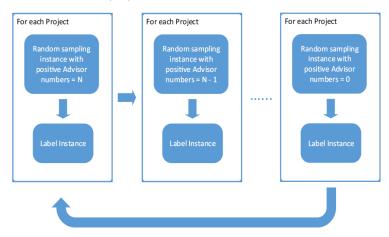
#### • Choice of the Advisors

An Advisor is a deterministic rule that gives a classification of a code element, telling if it is a code smell or not. The idea is that Advisors should approximate the label better than the random choice, and by aggregating their suggestions we should be able to sample more code elements affected by code smells. The rules can be extracted by some code smells detector or by some article, but, once again, the rules will affected the experiment so be sure that for each code smell, you select at least three Adivisor: (i)

one that most likely characterizes the not smell instances, (ii) one that most likely characterizes the affected instances, (iii) and one that include all the instances that cant be easily classified. WekaNose will apply the rules automatically. For more information about the metrics you can use to specify the advisors: http://essere.disco.unimib.it/reverse/MLCSD.html;

#### • Sampling

Following the output of the Advisors, the instances will be divided in groups based on how many advisors consider the instances affected by the code smell. After that sampling process describe by figure below will be followed automatically by WekaNose;



#### • Labeling

At this point the dataset is already created but the most important step has to be performed: for each instances contained in the dataset you have to manually evaluate it as affected or not affected by the code smell using the definition created at the first step. If the definition produced cant create a straight line that divide the instances affected by the code smell for the not affected ones, which most likely is going to happen, you can help yourself defining a serious of factor that characterize the code smell, they should be extracted by the definition itself if possible in order to minimize the aspect that can influence the result. Then order the factors from the one that, for you experimentation, is the most important to the less important one. Once you have you ordered list you have to keep it close to you for the hole evaluation process, and stick to that list every time you find an instance that for some reason is hard to evaluate

## 1.2 Machine Learning Approach

The second part can be separated in three steps, two of which can be performed using WekaNose. The hole process of training and testing of the ma-

chine learning algorithms is performed using OUTLINE (https://github.com/UmbertoAzadi/OUTLINE), that is basically a tool that allows to use the algorithms implemented in WEKA (https://www.cs.waikato.ac.nz/ml/weka/) through a more handy API.

#### • Select the machine learning algorithms

This step require some knowledge about the machine learning algorithms, plus you may want to take in consideration some of the studies performed about the usage of machine learning techniques for code smells detection:

- Arcelli Fontana, Francesca, Mntyl, Mika V., Zanoni, Marco, & Marino, Alessandro. 2016. Com- paring and Experimenting Machine Learning Techniques for Code Smell Detection. Empirical Softw. Engg., 21(3), 1143 - 1191.
- Fontana, F. A., Zanoni, M., Marino, A., & Mntyl, M. V. 2013 (Sept).
   Code Smell Detection: Towards a Machine Learning-Based Approach.
   Pages 396 399 of: 2013 IEEE International Conference on Software Maintenance.
- Maiga, A., Ali, N., Bhattacharya, N., Sabane, A., Gueheneuc, Y., & Aimeur, E. 2012a. SMURF: A SVM-based Incremental Antipattern Detection Approach. Pages 466 475 of: Proceedings of the 19th Working Conference on Reverse Engineering (WCRE 2012). Kingston, Ontario, Canada: IEEE.
- you can find a lot more just with a quick research on google

#### • Apply the grid search

As you may know each machine learning algorithms has some parameters that it require in input. So the second step of the second part of this experiment is to find, for each algorithms selected, the best parameter. The parameters are considered the best if the performance of the machine learning algorithms are maximized by them, id est if there arent any other values for the parameters that allows to reach a better performance. You can perform this step using WekaNose, in fact WEKA provide a serious of algorithms of its own to perform the greed search that can be used in WekaNose (https://weka.wikispaces.com/Optimizing+parameters)

#### • Compare the machine learning algorithms with each others

At this point you should be quite sure about the machine learning algorithms to use and the how to initilize them in order to maximize the performance, so what is left to do is run the algorithms and choose the best one. You can perform this step using WekaNose, in fact one you configured all the algorithms and the dataset through WekaNoses GUI you can Start the experiment, which means that a n-fold cross-validation tests with m repetitions will be performed for each classifiers, the result of this execution will be saved and compared using the corrected paired t-tests. (This procedure is already implemented in WEKA: https://weka.wikispaces.com/Using+the+Experiment+API)

## 2 How to install?

For install WekaNose you need:

- to download the repository from https://github.com/UmbertoAzadi/ WekaNose
- to have installed Java 6 or higher (for download https://www.java.com/it/download/)
- Right now this is not enough because some of the algorithms, for example LibSVM, create automatically a parameter -model (WEKA path), where WEKA path is C:\User\ProgramFiles\Weka in Windows, /home/user/Weka otherwise. Basically if you want to be sure that all the machine learning algorithms will run without problem you need to install WEKA:
  - Ubuntu: sudo apt-get install weka
  - Other: https://www.cs.waikato.ac.nz/ml/weka/downloading.html

### 3 How to run?

For run WekaNose you just need to

- open the command line:
  - Windows: WindowsKey+R and then type "cmd"
  - Ubuntu: Ctrl+Alt+t
- then type: "cd (path of WekaNose)" (es: cd /home/user/Download/WekaNose)
- and finally type: java -jar WekaNose.jar

As you may come to understand on your own, this software were tested only with Windows (10 Pro) and Ubuntu (16.04 LTS)

# 4 Exemple of usage

In this section we will see how to perform all the steps presented above through WekaNose GUI. Take in consideration that at this point you should already select the code smell and the repository of software systems (first two point in the Section 1.1)

#### 4.1 Dataset Creation

#### 4.1.1 HomePage

Considering that this is a new study you have to press the button Dataset Creation

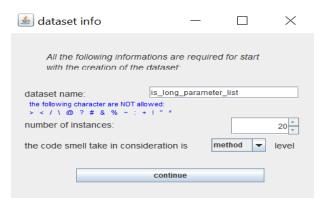


#### 4.1.2 Dataset information

Here you must specify:

- the name of the dataset/experiment
- how many instaces the dataset had to contain once it will be created
- if the code smelllevel is: class or method

Then press continue.



As result of this first step you should see that a fold named as  $\langle dataset\ name \rangle$  has been created in ./WekaNose/result

```
C:\Users\uazad\Desktop\WekaNose\result>tree
Elenco del percorso delle cartelle per il volume C
Numero di serie del volume: 000000DA-2CA5:9D10
C:.

-is_long_parameter_list

--TestClass
--classification_result
--freemind
--jasml
--jFin
--TestMothod
```

#### 4.1.3 Load source

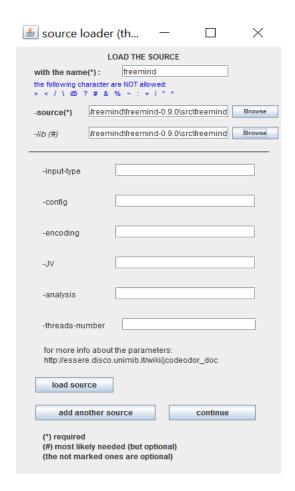
Now you have to specify one by one the source, id est the project you want to analyze. The parameter that you must specify are:

- the name of the source
- the path of the source

The optionals one are:

- the library needed by the source
- the remaining five can be ignored, unless you have some specific requirement, in that case you can find a more information here: http://essere.disco.unimib.it/wiki/jcodeodor\_doc

One you specified all the information you have to press the button load source, you should take in consideration that if the source is quite big it could take a few minutes. Then if you want to specify another source you have to press the button add another source, otherwise you can press the button continue.



As result of each loading you should see that, inside the fold created in the initial step, a fold named as  $\langle name \rangle$  has been created and it contain a file called  $\langle name \rangle$ . SQLite, which contain the result of the analyzes

```
C:\Users\uazad\Desktop\WekaNose\result \is_long_parameter_list tree /f
Elenco del percorso delle cartelle per il volume C
Numero di serie del volume: 00000008F-2CA5:9D10
C:.
freemind
freemind.SQLite

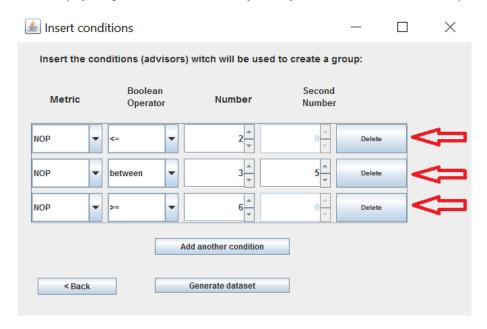
-jasml
jasml.SQLite
```

#### 4.1.4 Specify the Advisors

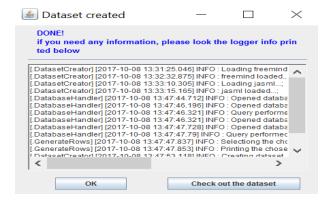
The next step is to specify the advisors, as you can see you can:

- Add another condition
- Delete an Advisor already created
- go Back, in case you forgot to specify some sources

One you specified all the Advisors you have to press the button Generate Dataset (if you specified a lot of source you may need to wait a few minutes).



Once the dataset is generated this windows will pop-up and you can open the directory where the dataset has been saved using the "Check out the result button



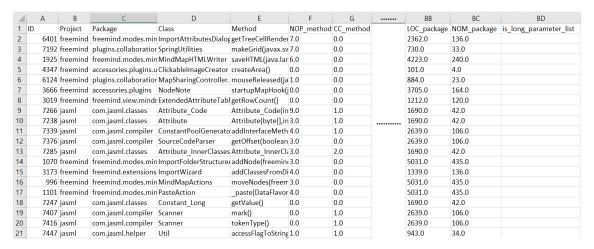
As result of this step you should see that, inside the fold created in the initial step, a new file called dataset.csv has been created and inside of each sources fold some new files has been created:

- Method.csv contain all the metrics at method level
- Class.csv contain all the metrics at class level
- Package.csv contain all the metrics at package level
- One file for each advisors selected, witch will contain the instances (methods in this case) that match with the characteristic specified
- chosen\_instaces.csv contain all the sources instances that have been chosen to be part of the dataset

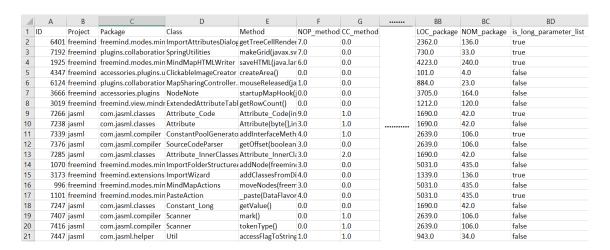
```
:\Users\uazad\Desktop\WekaNose\result\is_long_parameter_list>tree
Elenco del percorso delle cartelle per il volume C
Numero di serie del volume: 000000BC-2CA5:9D10
   dataset.csv
   freemind
       chosen_instances.csv
       Class.csv
       freemind.SQLite
       Method.csv
       NOP_method_between_3_5.csv
       NOP_method_greater_eq_6.csv
       NOP_method_less_eq_2.csv
       Package.csv
  —jasml
       chosen_instances.csv
       Class.csv
       jasml.SQLite
       Method.csv
       NOP_method_between_3_5.csv
       NOP method greater eq 6.csv
       NOP_method_less_eq_2.csv
       Package.csv
```

#### 4.1.5 Instances Labeling

If you open the file called dataset.csv it should be looking something like this:



Now you have to label the instances, which means that you have to write in the last column "true" if you think that the instance take in consideration is affected by the code smell, "false" otherwise. Of course you dont have to use "true or "false", any kind of label will do just as well.



#### 4.2 Machine Learning Tecnique

As you can see the windows is divided in three part:

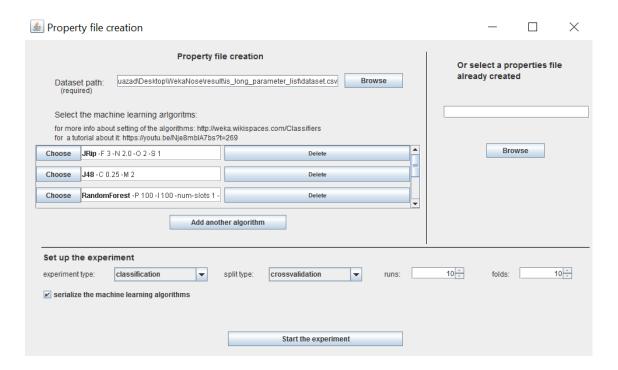
• the first part (top-left) allows to create a properties file by specifying the dataset path and at least one classifier. The properties file will be use by OUTLINE to perform the experiment.

- If you had already created the properties file, in a former experiment for example, you can select it using the the second part of the window (top-right), if you do that make sure that the path field of the properties file is consistent.
- finally the bottom part of the windows allows you to specify some experiments characteristic:
  - the experiment type: classification (default), regression or custom
  - the experiment split type: cross validation(default), random split or custom

For more information about the custom experiment: https://github.com/UmbertoAzadi/OUTLINE

- how many times the experiment has to be repeated (default = 10)
- in how many subset the dataset has to be split during the cross-validation (default = 10)

Once you insert all the information require you have to press the Start the experiment button, now based on which and how many many classifiers you selected the time required could be very high. For example during my stage i used a properties file with 32 classifier, 16 of which used the boosting technique AdaBoostM1, and it took 20 hour more or less.



As result of this step you should see that, inside the fold specified in the properties files field called path, a fold called classification\_result is created, it will contain the following file:

- Experiment\_result.txt that contain the ranked algorithms
- ExperimentStatistics.arff that contain the result of each classification/regression
- One file for each algorithm selected with extension .txt that contain the human-readable result of the classification. Take in consideration that not all the algoritms result can be described in a human readable way.
- (Only if the flag serialize the ... was selected) One file for each algorithm selected with extension .model that are the serialized algorithms

```
::\Users\uazad\Desktop\WekaNose\result\is_long_parameter_list>tree /f
Elenco del percorso delle cartelle per il volume C
Numero di serie del volume: 0000000B-2CA5:9D10
    dataset.csv
   dataset.properties
   -classification_result
        1_is_long_parameter_list_JRip.model
1_is_long_parameter_list_JRip.txt
2_is_long_parameter_list_RandomForest_RandomTree.model
        2_is_long_parameter_list_RandomForest_RandomTree.txt
        3_is_long_parameter_list_J48.model
        3_is_long_parameter_
                              _list_
                                    J48.txt
        ExperimentStatistics.arff
        Experiment_Result.txt
    freemind
        chosen_instances.csv
        Class.csv
        freemind.SQLite
        Method.csv
        NOP_method_between_3_5.csv
        NOP method greater_eq_6.csv
        NOP_method_less_eq_2.csv
        Package.csv
   -jasml
        chosen_instances.csv
        Class.csv
jasml.SQLite
        Method.csv
        NOP_method_between_3_5.csv
        NOP_method_greater_eq_6.csv
        NOP method less eq 2.csv
        Package.csv
```