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i. PROJECT OVERVIEW

Credit scoring models are statistical analysis used by banks that evaluate the worthiness to receive credit. The agencies select statistical characteristics found in a person's credit payment patterns, analyze them and come up with a credit score.

Scoring calculations are based on *payment record, frequency of payments, amount of debts, credit charge-offs, and number of credit cards* held. A weight is assigned to each factor considered in the model's formula and a credit score is assigned based on the evaluation.

Lenders use credit scores to determine the risk involved in making a loan, the terms of the loan, and the interest rate. The higher the score, the better the terms of a loan will be offered.

Credit scoring is perhaps *one of the most classic applications for predictive modeling* to predict whether or not the credit extended to an applicant will likely result in profit or loss for the lending institution. Credit scoring is the set of decision models and their underlying techniques that aid lenders in the granting of consumer credit. These techniques determine who will get credit, how much credit they should get, and what operational strategies will enhance the profitability of the borrowers to the lenders. Further, they help to assess the risk in lending. Credit scoring is a dependable assessment of a person's credit worthiness since it is based on actual data.

• The target variable to predict credit score for current credit card customer is "Bad label".

If Bad_label = 0, the customer has good credit history.

Bad label = 1, the customer has bad credit history.

• So, it is a *binary classification* that is used to predict the probability of a categorical dependent variable. The dependent variable is a binary variable that contains data coded as 1 (no) or 0 (yes).

ii. REQUIREMENTS

- a. Dataset of credit history of customer
- b. Python 3
- c. Python libraries
 - o Numpy
 - o Pandas
 - o Sklearn
 - o Matplotlib
 - o Seaborn
- d. SonarPython
- e. Pylint
- f. Pybuilder
- g. Travis CI

iii. UML (Unified Modeling Language)

UML is a modern approach to modeling and documenting software. It is based on *diagrammatic representations* of software components.

Sequence Diagram of the "credit score modelling" system is shown in Fig 1.

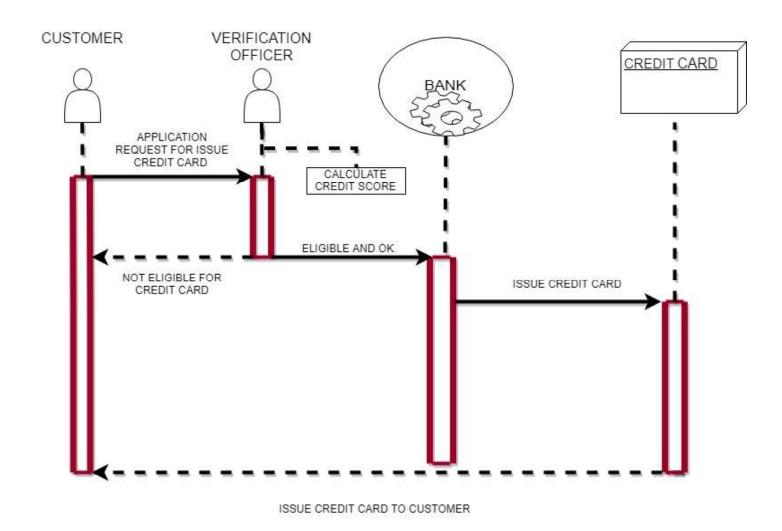


Fig 1: Sequence Diagram of credit score modelling

Class diagram of the system is shown in Fig 2.

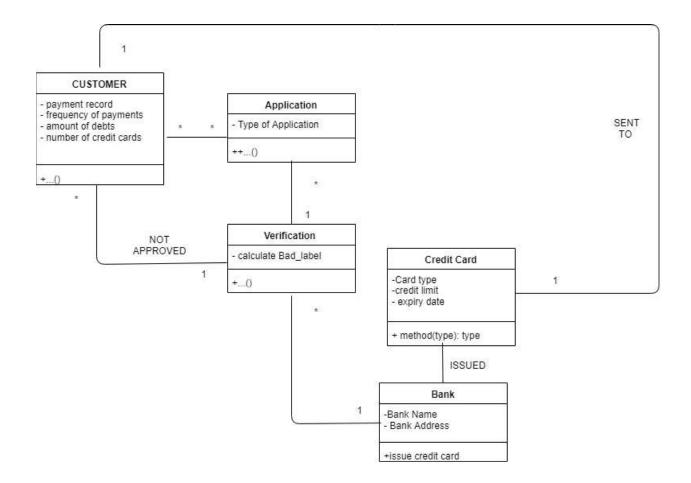


Fig 2: Class Diagram of credit score modelling

Use case diagram of the system is shown in Fig 3.

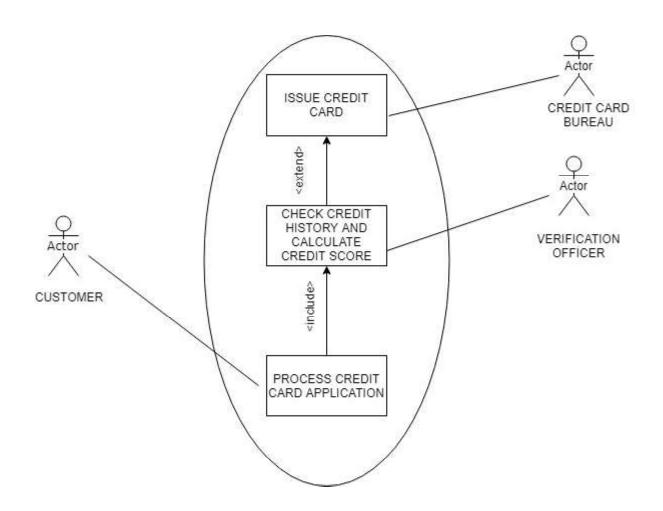


Fig 3: Use Case Diagram of credit score modelling

iv. Metrics

A software metric is a measure that says something about the software. There are many types of metrics related to software development, for example:

- sourcecode
- binary code
- external Libs
- buildinformation
- versioning information
- forms of meta- and additional Information (e.g. Annotations)

It is helpful to have an overview of very simple code metrics. These include the following metrics:

- lines of code
- comments
- methods/functions
- classes
- packages
- files
- duplication

To measure these simple code metrices, we can use SonarQube. SonarQube is static code analyzer to detect bugs, code smells, and security vulnerabilities on different programming languages. SonarQube offers reports on duplicated code, coding standards, unit tests, code coverage, code complexity, comments, bugs, and security vulnerabilities.

Analysis of the python code written for credit score modelling is done by **SonarPython** scanner which is available at -

https://docs.sonarqube.org/display/PLUG/SonarPython. The report is shown in Fig 4 &5.

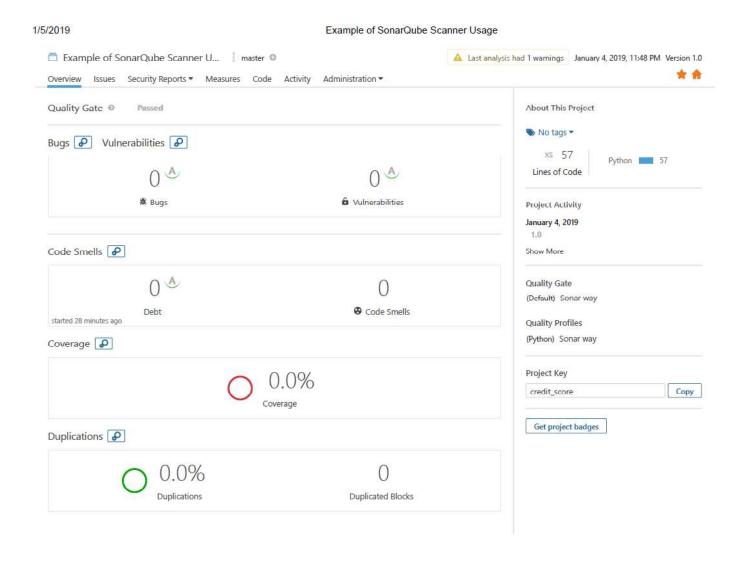


Fig 4: Analysis of basic metrics by SonarPython

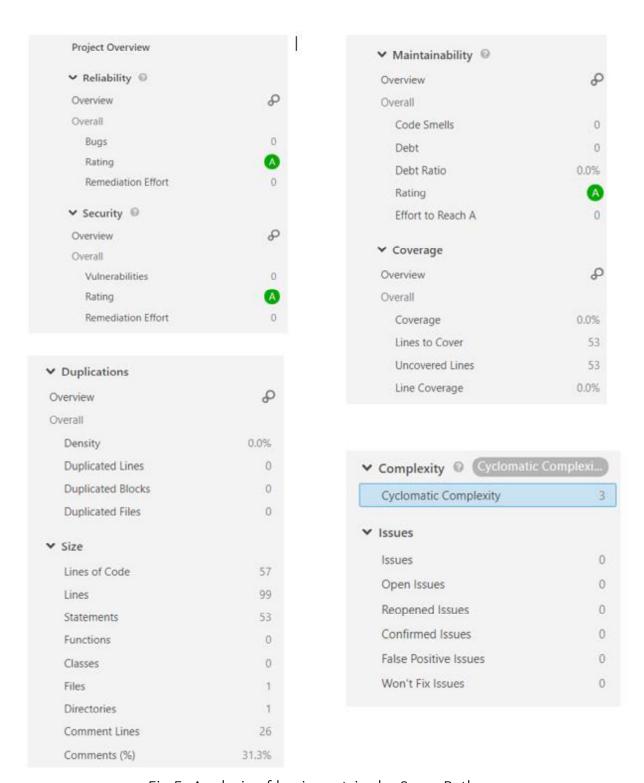


Fig 5: Analysis of basic metrics by SonarPython

V. Clean Code Development

Clean code is code that is *easy to understand* (i.e. easy to understand the execution flow of the entire application, easy to understand how the different objects collaborate with each other, easy to understand the role and responsibility of each class, easy to understand what each method does, easy to understand what is the purpose of each expression and variable) and *easy to change* (i.e. code is easy to extend and refactor, and it's easy to fix bugs in the codebase.

Pylint has been used here for clean code development. Pylint is a source-code, bug and quality checker for the Python programming language. It is similar to Pychecker and Pyflakes, but includes the following features:

- Checking the length of each line
- Checking that variable names are well-formed according to the project's coding standard
 - Checking that declared interfaces are truly implemented.

The pylint report of the above-mentioned codebase is attached below (Fig 6).

Fig 6: Pylint Report

i. Meaningful Names

```
# reading dataframe
DF = pd.read_csv('C:/Users/PALLAVI/Desktop/BEUTH_1ST_SEM/pet_project/Banking.csv')
print(DF.shape)
print(DF.columns))
print(DF.isnull().sum().sum())

# dependent (X) and independent (y) variables
X = data_final.drop('Bad_label', axis=1)
y = data_final.Bad_label
feat_labels = data_final.columns[1:]

# train-test split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3)
```

ii. Comments

iii. Formatting

Setting a limit of characters (120) per line of code.

Use spaces between operators, parameters, and commas.

```
f dependent and independent variables
X_final = data_final.drop('Bad_label', axis=1)
y final = data_final.Bad_label
X_final_train, X_final_test, y_final_train, y_final_test = \
train_test_split(X_final, y_final, test_sits=0.3)
```

iv. Less arguments in function

Better Function (More than three arguments are evil)

```
# train-test split
x_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.3)

# Logistic Regression Model
classifier = LogisticRegression(random_state=0)
classifier.fit(X_final_train, y_final_train)
y_pred = classifier.predict(X_final_test)
```

v. Use correct Indentation

```
for var in cat_vars:
    cat_list = pd.get_dummies(DF[var], prefix=var)
    DF_updated = DF.join(cat_list)
    DF = DF_updated
```

VI. Build Management

For build management, *PyBuilder* has been used in this project. PyBuilder is a multipurpose software build tool. Most commonly it targets the building and management of software with a strong focus on Python. The advantages of PyBuilder are –

- Automatic execution of unit and integration tests on every build
- Automatic analysis of the code coverage
- Automatic execution and result interpretation of analysis tools, such as flake8
- Automatic generation of distutils script setup.py

A complete walkthrough for the pybuilder project is given as follows:

i. Installing PyBuilder:

```
C:\>mkdir my_petproject
C:\>cd my_petproject
C:\my_petproject>
```

ii. Create Virtualenvironment and installing PyBuild:

```
C:\my_petproject>virtualenv venv
New python executable in C:\my_petproject\venv\Scripts\python.exe
Installing setuptools, pip, wheel...
done.
C:\my_petproject>venv\Scripts\activate
(venv) C:\my_petproject>pip install pybuilder

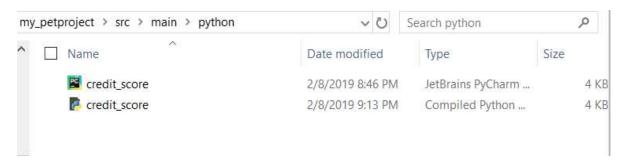
Collecting pybuilder
Requirement already satisfied: pippippippippip
Collecting taller (from pybuilder)
Collecting setuptools—39.8.0 (from pybuilder)
Using cached https://files.pythonhosted.org/packages/20/d7/04a0b690d30351d3e2ff288fdb0eadbf6ed8058Scc121c00bfd9Sala8c2e/setuptools-39.0.1-py2.py3-none-any.whl
Requirement already satisfied: wheel==0.31 in c:\my_petproject\venv\lib\site-packages (from pybuilder)
Using cached https://files.pythonhosted.org/packages/4a/82/lb9fba6e93629a8557f9784cd8f1ae063c8762c26446367a6764edd328ce/tblib-1.3.2-py2.py3-none-any.whl
Installing collected packages: tailer, setuptools, tblib, pybuilder
Found existing installation: setuptools 40.8.0
Uninstalling setuptools-40.8.0
Successfully uninstalled setuptools-40.8.0
Successfully installed setuptools-40.8.0
Successfully installed setuptools-40.8.0
Successfully installed setuptools-40.8.0
Successfully installed setuptools-40.8.0
You are using pip version 10.0.1, however version 19.0.1 is available.
You should consider upgrading via the 'python -m pip install --upgrade pip' command.
```

iii. Scaffolding:

```
(venv) C:\my_petproject>pyb_ --start-project
Project name (default: 'my_petproject') : credit_score
Source directory (default: 'src/main/python') :
Docs directory (default: 'docs') :
Unittest directory (default: 'src/unittest/python') :
Scripts directory (default: 'src/main/scripts') :
Use plugin python.flake8 (Y/n)? (default: 'y') : y
Use plugin python.coverage (Y/n)? (default: 'y') : y
Use plugin python.distutils (Y/n)? (default: 'y') : y
Created 'setup.py'.
```

iv. Adding Python Source Files:

The next step is to add a Python module that contains our sources. PyBuilder separates source files and expects them in different directories based on their meaning. The default location for main python sources is src/main/python.



v. New build.py:

build.py is the centralized project description for the pet project.

```
from pybuilder.core import use_plugin, init, Author

use_plugin("python.core")
use_plugin("python.unittest")
use_plugin("python.install_dependencies")
use_plugin("python.flake8")
use_plugin("python.coverage")
use_plugin("python.distutils")

name = "credit_score"
default_task = "publish"
authors = [Author("EALLANI MITEA", "pallavi.mitra1992@gmai.com")]

einit

einit
def set_properties(project):
pass
```

Let's run PyBuilder and see what happens:

```
(vero) (:]us_petrojectopyb_
[i]shPybuilder version 0.1.1700m

Build started at 2019-02-08 20:30:25

Z[sm[INFO] E[dm Executing publid in C:Nuy_petroject

Z[sm[INFO] E[dm Executing publid in Genedatory Vorone

Z[sm[INFO] E[dm Executing publid in dependency vorone

Z[sm[INFO] E[dm Installing plugid dependency vorone

Z[sm[INFO] E[dm Installing plugid dependency publid

Z[sm[INFO] E[dm Executing publid in Execution of vorone vorone

Z[sm[INFO] E[dm Executing unit tests from Python modules in c:Nuy_petroject\src\unittest/mython

Z[sm[INFO] E[dm Exiting in Stribution in Stribution in C:Nuy_petroject\src\unittest/mython

Z[sm[INFO] E[dm Exiting in Stribution in Stribution in C:Nuy_petroject\src\unittest/mython

Z[sm[INFO] E[dm Exiting in Stribution in C:Nuy_petroject\src\unittest/mython

Z[s
```

We don't have any tests right now. To avoid the break in the build, we need to modify build.py.

vi. Unittest Plugin:

Let's start with a test at src/unittest/python/credit_score_tests.py:

```
from mockito import mock, verify
import unittest
from credit_score import testing

pelacs Test(unittest.TestCase):
    def test_should_issue_message(self):
        out = mock()
        testing(out)
        verify(out).write("Worked fine\n")

if name == '___main__':
        unittest.main()
```

vii. Dependency:

Since we're using various libraries of python, we'll have to install it by telling our initializer in build.py about it:

We can install our dependency by running PyBuilder with the corresponding task

```
(venv) C:\my_petproject>pyb_ install_dependencies
[1mPyBuilder version 0.11.170[0m
Build started at 2019-02-08 20:54:32
@[1m[INFO] @[0m Building credit_score version 1.0.dev0
2[1m[INFO] 2[0m Executing build in C:\my_petproject
[1m[INFO] D[Om Going to execute task install_dependencies
[1m[INFO] D[Om Installing all dependencies
[1m[INFO] D[Om Processing batch dependency 'matplotlib'
1 [1m[INFO] D[Om Processing batch dependency 'mockito'
1 [1m[INFO] D[Om Processing batch dependency 'pandas'
[1m[INFO] [0m Processing batch dependency 'seaborn'
[[1m[INFO] [][0m Processing batch dependency 'sklearn'

[1;32mBUILD SUCCESSFUL□[0m]

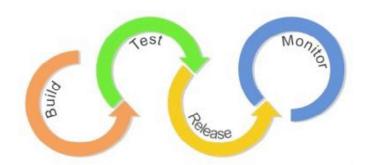
Build Summary
              Project: credit_score
              Version: 1.0.dev0
      Base directory: C:\my_petproject
                 Tasks: install_dependencies [174507 ms]
Build finished at 2019-02-08 20:57:27
```

viii. Running the test:

```
| James | Color | James | Jame
```

VII. CONTINUOUS INTEGRATION

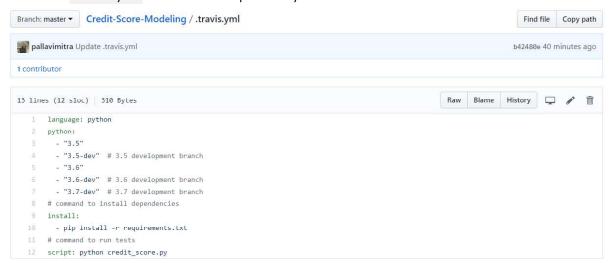
Continuous Integration is a development practice of integrating code into a single build multiple times per day. Builds are usually automated and depend on thorough automatic test to ensure they complete properly and without issue.



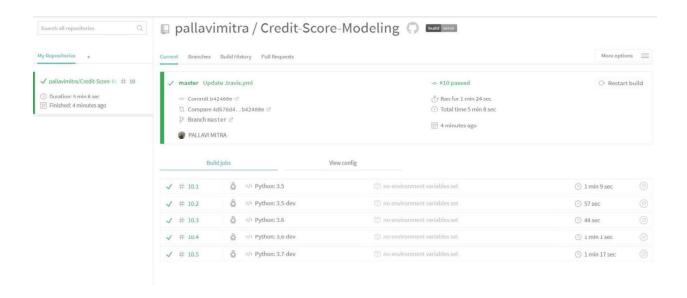
Here, "*Travis Cl*" has been used for continuous integration. It's built to communicate with github and can be set up to build for every push and pull in request that comes in for the repository.

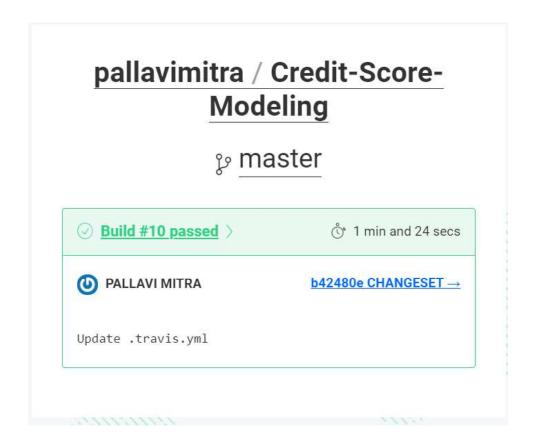
To get started with Travis Ci followings steps must be taken.

- 1. Go to Travis-ci.com and Sign up with GitHub.
- 2. Accept the Authorization of Travis CI and redirect to GitHub.
- 3. Click the green *Activate* button and select the repository that want to use with Travis Cl.
- 4. Add a .travis.yml file to the repository to tell Travis CI what to do.



- 5. Add the .travis.yml file to git, commit and push, to trigger a Travis CI build.
- 6. Check the build status page to see if your build passes or fails.





VIII. DOMAIN SPECIFIC LANGUAGES

A *domain-specific language* (*DSL*) is a computer language specialized to a particular application domain. There are a wide variety of DSLs, ranging from widely used languages for common domains, such as HTML for web pages, down to languages used by only one or a few pieces of software.

In the previous section i.e. in continuous integration, we have to use a yml code to trigger Travis CI build. So, this .travis.yml can be considered as DSL for this pet project.

```
language: python:
- "3.5"
- "3.5-dev" # 3.5 development branch
- "3.6"
- "3.6-dev" # 3.6 development branch
- "3.7-dev" # 3.7 development branch
# command to install dependencies
| install:
- pip install -r requirements.txt
# command to run tests
| script: python credit_score.py
```

IX. FUNCTIONAL PROGRAMMING

Functional programming languages are specially designed to handle symbolic computation and list processing applications. Functional programming is based on mathematical functions. Some of the popular functional programming languages include: Lisp, Python, Erlang, Haskell, Clojure, etc.

Following functional programming have been utilized in the pet project.

h. The Lambda Expression:

Instead of the def syntax for function declaration, we can use a lambda expression to write Python functions. The lambda expression takes in a comma separated sequences of inputs (like def). Then, immediately following the colon, it returns the expression without using an explicit return statement. Finally, when assigning the lambda expression to a variable, it acts exactly like a Python function, and can be called using the function call syntax.

```
# Sort on the second tuple value (the float).
print(sorted(feature, key=lambda x: x[1]))
```

Here, the sorted() function takes in an optional key argument (a function) that describes how the items in a list should be sorted.

ii. Zip Function:

zip is a common functional programming method like map or fold. They are designed to perform common batch operations on lists. The zip() function take iterables (can be zero or more), makes iterator that aggregates elements based on the iterables passed, and returns an iterator of tuples.

```
# Train A Random Forest Classifier

clf = RandomForestClassifier(n_estimators=10000, random_state=0, n_jobs=-1)

clf.fit(X_train, Y_train)

for feature in zip(FEATURE_LABELS, clf.feature_importances_):
    print(feature)
```