

PyAudit: Python Data Audit Library API

Wenqiang Feng and Ming Chen

CONTENTS

1	Prefa	ace .	3
	1.1	About	3
		1.1.1 About this API	3
		1.1.2 About the author	3
	1.2	Feedback and suggestions	4
2	How	to Install	5
	2.1	Clone the Repository	5
	2.2	Install	5
	2.3	Uninstall	5
	2.4	Test	5
3	Pytho	on Data Audit Functions	9
	3.1	dtypes_class	9
	3.2	missing_rate	9
	3.3		10
	3.4		10
	3.5		11
	3.6		12
4	Demo	os	13
5	Main	Reference	15
D.			
Вi	bliogra	aphy	17
Рy	thon N	Module Index	19
In	dex		21



Welcome to our **PyAudit: Python Data Audit Library API!** The PDF version can be downloaded from HERE.

CONTENTS 1

2 CONTENTS

CHAPTER

ONE

PREFACE

Chinese proverb

Good tools are prerequisite to the successful execution of a job. – old Chinese proverb

1.1 About

1.1.1 About this API

This document is the API book for our PyAudit: Python Data Audit Library [PyAudit] API. The PDF version can be downloaded from HERE. You may download and distribute it. Please be aware, however, that the note contains typos as well as inaccurate or incorrect description.

The API assumes that the reader has a preliminary knowledge of python programing and Linux. And this document is generated automatically by using sphinx.

1.1.2 About the author

Wengiang Feng

- Sr. Data Scientist and PhD in Mathematics
- University of Tennessee at Knoxville
- Webpage: http://web.utk.edu/~wfeng1/
- Email: von198@gmail.com

• Ming Chen

- Data Scientist and PhD in Genome Science and Technology
- University of Tennessee at Knoxville

- Email: ming.chen0919@gmail.com

Biography

Wenqiang Feng is Data Scientist within DST's Applied Analytics Group. Dr. Feng's responsibilities include providing DST clients with access to cutting-edge skills and technologies, including Big Data analytic solutions, advanced analytic and data enhancement techniques and modeling.

Dr. Feng has deep analytic expertise in data mining, analytic systems, machine learning algorithms, business intelligence, and applying Big Data tools to strategically solve industry problems in a cross-functional business. Before joining DST, Dr. Feng was an IMA Data Science Fellow at The Institute for Mathematics and its Applications (IMA) at the University of Minnesota. While there, he helped startup companies make marketing decisions based on deep predictive analytics.

Dr. Feng graduated from University of Tennessee, Knoxville, with Ph.D. in Computational Mathematics and Master's degree in Statistics. He also holds Master's degree in Computational Mathematics from Missouri University of Science and Technology (MST) and Master's degree in Applied Mathematics from the University of Science and Technology of China (USTC).

Declaration

The work of Wenqiang Feng was supported by the IMA, while working at IMA. However, any opinion, finding, and conclusions or recommendations expressed in this material are those of the author and do not necessarily reflect the views of the IMA, UTK and DST.

1.2 Feedback and suggestions

Your comments and suggestions are highly appreciated. I am more than happy to receive corrections, suggestions or feedbacks through email (Wenqiang Feng: von198@gmail.com and Ming Chen: ming.chen0919@gmail.com) for improvements.

CHAPTER

TWO

HOW TO INSTALL

2.1 Clone the Repository

git clone https://github.com/runawayhorse001/PyAudit.git

2.2 Install

```
cd PyAudit
pip install -r requirements.txt
python setup.py install
```

2.3 Uninstall

```
pip uninstall statspy
```

2.4 Test

```
cd PyAudit/test
python test1.py
```

test1.py

```
from PyAudit.basics import missing_rate, zero_rate, dtypes_class
from PyAudit.basics import feature_variance, freq_items_df, feature_len
from PyAudit.basics import numeric_summary
import pandas as pd
```

(continues on next page)

(continued from previous page)

```
d = \{'A': [1, 0, None, 3],
     'B': [1, 0, 0, 0],
     'C': ['a', None, 'c', 'd']}
# create DataFrame
df = pd.DataFrame(d)
print (missing_rate(df))
print (zero_rate(df))
print (feature_variance(df))
print(df)
print (feature_len (df))
(num_fields, cat_fields, bool_fields, data_types, type_class) = dtypes_
⇔class(df)
print (numeric_summary (df[num_fields]))
d = {
    'num': list('1223334444'),
    'cat': list('wxxyyyzzzz')
df = pd.DataFrame(d)
df = df.astype({"num": int, "cat": object})
print(freq_items_df(df, top_n=4))
# read df
df = pd.read_csv('Heart.csv', dtype={'Sex': bool})
print (df.head(5))
(num_fields, cat_fields, bool_fields, data_types, type_class) = dtypes_

class(df)
print (num_fields)
print (cat_fields)
print (bool_fields)
print (data_types)
print (type_class)
print (missing_rate(df))
#print(zero_rate(df))
print(freq_items_df(df, top_n=4))
print (feature_len(df))
print (numeric_summary(df[num_fields]))
```

Results:

```
feature missing_rate
       Α
                  0.25
1
                  0.00
       В
                  0.25
       С
 feature zero_rate
          0.333333
       Α
           0.750000
       В
           0.000000
       С
 feature feature_variance
0
       Α
                       1.0
                       0.5
       В
1
2
                       1.0
         Sex
                 ChestPain RestBP
                                   Chol ... Oldpeak Slope
     Thal AHD
0
   63
        True
                   typical
                               145
                                     233
                                         . . .
                                                   2.3
                                                               0.0
    fixed
            No
        True asymptomatic
                               160
                                     286
                                                               3.0
   67
                                                   1.5
  normal Yes
   67
        True asymptomatic
                               120
                                     229
                                                   2.6
                                                               2.0
⇔reversable Yes
        True
                nonanginal
                               130
                                     250
                                                   3.5
                                                               0.0
   37
                                          . . .
→ normal
   41 False
                nontypical
                               130
                                     204 ...
                                                   1.4
                                                            1 0.0
→ normal No
[5 rows x 14 columns]
['Age', 'RestBP', 'Chol', 'Fbs', 'RestECG', 'MaxHR', 'ExAng', 'Oldpeak
→', 'Slope', 'Ca']
['ChestPain', 'Thal', 'AHD']
['Sex']
     feature
              dtypes
               int64
0
         Age
1
         Sex
                bool
2
   ChestPain
             object
3
               int64
      RestBP
4
        Chol
               int64
5
               int64
         Fbs
6
     RestECG
               int64
7
       MaxHR
               int64
8
       ExAng
               int64
9
     Oldpeak float64
10
       Slope
               int64
11
          Ca float64
12
              object
        Thal
13
         AHD
             object
     feature dtypes
                          class
```

(continues on next page)

2.4. Test 7

(continued from previous page)

0	Age	int64	numeric
1	Sex	bool	bool
2	ChestPain	object	category
3	RestBP	int64	numeric
4	Chol	int64	numeric
5	Fbs	int64	numeric
6	RestECG	int64	numeric
7	MaxHR	int64	numeric
8	ExAng	int64	numeric
9	Oldpeak	float64	numeric
10	Slope	int64	numeric
11	Ca	float64	numeric
12	Thal	object	category
13	AHD	object	category
	feature	missing_	rate
0	Age	0.00	0000
1	Sex	0.00	0000
2	ChestPain	0.00	0000
3	RestBP	0.00	0000
4	Chol	0.00	0000
5	Fbs	0.00	0000
6	RestECG	0.00	0000
7	MaxHR	0.00	0000
8	ExAng	0.00	0000
9	Oldpeak	0.00	0000
10	Slope	0.00	0000
11	Ca	0.01	3201
12	Thal	0.00	6601
13	AHD	0.00	0000
Pro	cess finish	ed with e	xit code (

CHAPTER

THREE

PYTHON DATA AUDIT FUNCTIONS

3.1 dtypes_class

```
PyAudit.basics.dtypes_class (df_in) numerical, categorical and bool name list in the DataFrame
```

Parameters df_in – input pandas DataFrame

Returns numerical, categorical and bool name list

Author Wenqiang Feng and Ming Chen

Email von198@gmail.com

3.2 missing_rate

```
PyAudit.basics.missing_rate (df_in) calculate missing rate for each feature in the DataFrame
```

Parameters df_in - input pandas DataFrame

Returns missing rate

Author Wenqiang Feng and Ming Chen

Email von198@gmail.com

```
>>> import pandas as pd
>>> d = {'A': [1, 0, None, 3],
        'B': [1, 0, 0, 0],
        'C': ['a', None, 'c', 'd']}
>>> # create DataFrame
>>> df = pd.DataFrame(d)
>>> from PyAudit.basics import missing_rate
>>> missing_rate(df)
        feature missing_rate
      0 A 0.25
             В
                       0.00
      1
      2
             С
                       0.25
```

3.3 zero_rate

```
PyAudit.basics.zero_rate(df_in)
```

calculate the percentage of 0 value for each feature in the DataFrame

Parameters df_in – input pandas DataFrame

Returns zero rate

Author Wenqiang Feng and Ming Chen

Email von198@gmail.com

3.4 feature_variance

```
PyAudit.basics.feature_variance (df_in) calculate the variance for each feature
```

Parameters df_in - input pandas DataFrame

Returns feature variance

Author Wenqiang Feng and Ming Chen

Email von198@gmail.com

```
>>> import pandas as pd
>>> d = {'A': [1, 0, None, 3],
         'B': [1, 0, 0, 0],
         'C': ['a', None, 'c', 'd']}
>>> # create DataFrame
>>> df = pd.DataFrame(d)
>>> from PyAudit.basics import zero_rate
>>> zero_rate(df)
         feature feature_variance
       0
               Α
                                1.0
                                0.5
               В
       1
       2
               С
                                1.0
```

3.5 freq_items_df

PyAudit.basics.freq_items_df(df_in, top_n=3)

find out the top n values and the corresponding frequency for each feature

Parameters

- **df_in** input pandas DataFrame
- top_n the number of the top values

Returns top n values and the corresponding frequency for each feature

3.6 feature_len

```
PyAudit.basics.feature_len(df_in)
```

find out the min and max length of values for each feature

Parameters df_in – input pandas DataFrame

Returns min and max length DataFrame

```
>>> d = {'A': [1, 0, None, 3],
>>> 'B': [1, 0, 0, 0],
>>> 'C': ['a', None, 'c', 'd']}
```

```
>>> # create DataFrame
>>> df = pd.DataFrame(d)
>>> print(df)
        А В
      1.0 1
   1 0.0 0 None
   2 NaN 0
     3.0 0
                 d
>>> print(feature_len(df))
   feature min_length max_length
 0
                     3
         Α
 1
         В
                     1
                                 1
         С
                                 4
                     1
```

CHAPTER

FOUR

DEMOS

This is a usage of PyAudit.basics.dtypes_class():

For example:

```
>>> from PyAudit.basics import missing_rate, zero_rate, dtypes_class
>>> df = pd.read_csv('Heart.csv', dtype={'Sex': bool})
>>> (num_fields, cat_fields, bool_fields, data_types, type_class) = ...
→dtypes_class(df)
['Age', 'RestBP', 'Chol', 'Fbs', 'RestECG', 'MaxHR', 'ExAng', 'Oldpeak
→', 'Slope', 'Ca']
['ChestPain', 'Thal', 'AHD']
['Sex']
     feature
              dtypes
                int64
         Age
1
         Sex
                 bool
   ChestPain
2
              object
3
      RestBP
               int64
4
        Chol
                int64
5
               int64
         Fbs
6
     RestECG
               int64
7
                int64
       MaxHR
8
               int64
       ExAng
9
     Oldpeak float64
10
        Slope
               int64
11
          Ca float64
12
        Thal
              object
13
         AHD object
     feature
              dtypes
                          class
0
         Age
               int64
                       numeric
1
         Sex
                 bool
                           bool
2
   ChestPain
              object category
3
      RestBP
               int64
                       numeric
               int64 numeric
4
        Chol
5
         Fbs
               int64 numeric
```

(continues on next page)

(continued from previous page)

```
RestECG
              int64 numeric
7
               int64 numeric
       MaxHR
8
       ExAng
              int64 numeric
9
     Oldpeak float64 numeric
10
       Slope
             int64 numeric
11
          Ca float64 numeric
12
             object category
        Thal
13
              object
         AHD
                      category
```

This is a usage of PyAudit.basics.feature_variance():

For example:

```
.,,.
   ,;;*;;;;
  .-'``;-');;.
    \d
                        .;;;,
                     ,;*;;;*;,
          \_.-') ___) --.;;;;**;;;,
          `""`;;\
            ; * ; ; ;
            ;;;;
  *;*;\|
 ;;;;/|
;;;*;/ \
;;;;;'.;
,;*;;;\/
          1/
                           ';;;
 ;;;;;/
 '*wf*/
                          ; *;
     \simeq in in in in \simeq
                 ^{8} и и и и ^{8}
                          ; 1
```

CHAPTER FIVE

MAIN REFERENCE

BIBLIOGRAPHY

[PyAudit] Wenqiang Feng and Ming Chen. Python Data Audit Library API, 2019.

18 Bibliography

PYTHON MODULE INDEX

p

PyAudit.basics, 12

INDEX

```
D
dtypes_class()
                         module
                    (in
                                  PyAu-
      dit.basics), 9
F
feature_len() (in module PyAudit.basics),
feature_variance() (in module PyAu-
      dit.basics), 10
freq_items_df()
                     (in module PyAu-
      dit.basics), 11
M
missing_rate()
                    (in
                         module
                                  PyAu-
      dit.basics), 9
P
PyAudit.basics (module), 9-12
Z
zero_rate() (in module PyAudit.basics), 10
```