

rk2

May 28, 2022

0.1 №2

0.1.1 :

0.1.2 .. 5-65 -19

:

```
[ ]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt

from sklearn import preprocessing
from sklearn.impute import SimpleImputer
from sklearn.preprocessing import LabelEncoder
from sklearn.model_selection import train_test_split
from sklearn import svm
from sklearn.metrics import accuracy_score, balanced_accuracy_score
from sklearn.metrics import precision_score, recall_score, f1_score, \
    classification_report

from sklearn.ensemble import AdaBoostClassifier
```

```
[ ]: data = pd.read_csv('./investments_VC.csv', encoding='latin1', sep=",")
target_col = 'status'
```

```
[ ]: data.head()
```

```
[ ]:
      permalink      name \
0      /organization/waywire      #waywire
1  /organization/tv-communications  &TV Communications
2    /organization/rock-your-paper  'Rock' Your Paper
3    /organization/in-touch-network  (In)Touch Network
4    /organization/r-ranch-and-mine  -R- Ranch and Mine

      homepage_url \
0      http://www.waywire.com
```

```

1      http://enjoyandtv.com
2      http://www.rockyourpaper.org
3      http://www.InTouchNetwork.com
4      NaN

```

```

                                category_list      market  \
0      |Entertainment|Politics|Social Media|News|      News
1                                |Games|      Games
2                                |Publishing|Education|      Publishing
3      |Electronics|Guides|Coffee|Restaurants|Music|i...      Electronics
4                                |Tourism|Entertainment|Games|      Tourism

```

```

      funding_total_usd      status country_code state_code      region ... \
0      17,50,000      acquired      USA      NY      New York City ...
1      40,00,000      operating      USA      CA      Los Angeles ...
2      40,000      operating      EST      NaN      Tallinn ...
3      15,00,000      operating      GBR      NaN      London ...
4      60,000      operating      USA      TX      Dallas ...

```

```

      secondary_market      product_crowdfunding      round_A      round_B      round_C      round_D  \
0      0.0      0.0      0.0      0.0      0.0      0.0
1      0.0      0.0      0.0      0.0      0.0      0.0
2      0.0      0.0      0.0      0.0      0.0      0.0
3      0.0      0.0      0.0      0.0      0.0      0.0
4      0.0      0.0      0.0      0.0      0.0      0.0

```

```

      round_E      round_F      round_G      round_H
0      0.0      0.0      0.0      0.0
1      0.0      0.0      0.0      0.0
2      0.0      0.0      0.0      0.0
3      0.0      0.0      0.0      0.0
4      0.0      0.0      0.0      0.0

```

[5 rows x 39 columns]

```
[ ]: data.shape
```

```
[ ]: (54294, 39)
```

```
[ ]: data.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 54294 entries, 0 to 54293
Data columns (total 39 columns):
#   Column                Non-Null Count  Dtype
---  -
0   permalink              49438 non-null  object

```

1	name	49437	non-null	object
2	homepage_url	45989	non-null	object
3	category_list	45477	non-null	object
4	market	45470	non-null	object
5	funding_total_usd	49438	non-null	object
6	status	48124	non-null	object
7	country_code	44165	non-null	object
8	state_code	30161	non-null	object
9	region	44165	non-null	object
10	city	43322	non-null	object
11	funding_rounds	49438	non-null	float64
12	founded_at	38554	non-null	object
13	founded_month	38482	non-null	object
14	founded_quarter	38482	non-null	object
15	founded_year	38482	non-null	float64
16	first_funding_at	49438	non-null	object
17	last_funding_at	49438	non-null	object
18	seed	49438	non-null	float64
19	venture	49438	non-null	float64
20	equity_crowdfunding	49438	non-null	float64
21	undisclosed	49438	non-null	float64
22	convertible_note	49438	non-null	float64
23	debt_financing	49438	non-null	float64
24	angel	49438	non-null	float64
25	grant	49438	non-null	float64
26	private_equity	49438	non-null	float64
27	post_ipo_equity	49438	non-null	float64
28	post_ipo_debt	49438	non-null	float64
29	secondary_market	49438	non-null	float64
30	product_crowdfunding	49438	non-null	float64
31	round_A	49438	non-null	float64
32	round_B	49438	non-null	float64
33	round_C	49438	non-null	float64
34	round_D	49438	non-null	float64
35	round_E	49438	non-null	float64
36	round_F	49438	non-null	float64
37	round_G	49438	non-null	float64
38	round_H	49438	non-null	float64

dtypes: float64(23), object(16)
memory usage: 16.2+ MB

0.1.3

```
[ ]: data.isnull().mean()
```

```
[ ]: permalink      0.089439
      name          0.089457
      homepage_url  0.152963
      category_list  0.162394
      market        0.162523
      funding_total_usd 0.089439
      status         0.113641
      country_code   0.186558
      state_code     0.444487
      region         0.186558
      city           0.202085
      funding_rounds 0.089439
      founded_at     0.289903
      founded_month  0.291229
      founded_quarter 0.291229
      founded_year   0.291229
      first_funding_at 0.089439
      last_funding_at 0.089439
      seed           0.089439
      venture        0.089439
      equity_crowdfunding 0.089439
      undisclosed    0.089439
      convertible_note 0.089439
      debt_financing 0.089439
      angel          0.089439
      grant          0.089439
      private_equity 0.089439
      post_ipo_equity 0.089439
      post_ipo_debt  0.089439
      secondary_market 0.089439
      product_crowdfunding 0.089439
      round_A        0.089439
      round_B        0.089439
      round_C        0.089439
      round_D        0.089439
      round_E        0.089439
      round_F        0.089439
      round_G        0.089439
      round_H        0.089439
      dtype: float64
```

```
[ ]: data=data.drop(['permalink', 'category_list', 'founded_at', 'founded_month',
                    'founded_quarter',
                    'first_funding_at', 'last_funding_at'],axis=1)
```

4856,

```
[ ]: data.isnull().sum()
```

```
[ ]: name                4857
      homepage_url       8305
      market             8824
      funding_total_usd   4856
      status             6170
      country_code       10129
      state_code         24133
      region             10129
      city               10972
      funding_rounds      4856
      founded_year       15812
      seed               4856
      venture            4856
      equity_crowdfunding 4856
      undisclosed         4856
      convertible_note    4856
      debt_financing      4856
      angel              4856
      grant              4856
      private_equity      4856
      post_ipo_equity     4856
      post_ipo_debt       4856
      secondary_market    4856
      product_crowdfunding 4856
      round_A            4856
      round_B            4856
      round_C            4856
      round_D            4856
      round_E            4856
      round_F            4856
      round_G            4856
      round_H            4856
      dtype: int64
```

```
[ ]: data=data.dropna(how="all")
```

0.1.4 SimpleImputer

SimpleImputer

```
[ ]: imp = SimpleImputer(missing_values=np.nan, strategy='most_frequent')

      imputed = {}

      for col in data:
```

```

contains_nan = data[col].isnull().sum() != 0
if contains_nan:
    data_imp = data[[col]]
    data_imp = imp.fit_transform(data_imp)
    imputed[col] = data_imp

for col_name in imputed:
    df = pd.DataFrame({col_name:imputed[col_name].T[0]})
    data[col_name] = df.copy()

data.info()

```

```
<class 'pandas.core.frame.DataFrame'>
```

```
Int64Index: 49438 entries, 0 to 49437
```

```
Data columns (total 32 columns):
```

#	Column	Non-Null Count	Dtype
0	name	49438 non-null	object
1	homepage_url	49438 non-null	object
2	market	49438 non-null	object
3	funding_total_usd	49438 non-null	object
4	status	49438 non-null	object
5	country_code	49438 non-null	object
6	state_code	49438 non-null	object
7	region	49438 non-null	object
8	city	49438 non-null	object
9	funding_rounds	49438 non-null	float64
10	founded_year	49438 non-null	float64
11	seed	49438 non-null	float64
12	venture	49438 non-null	float64
13	equity_crowdfunding	49438 non-null	float64
14	undisclosed	49438 non-null	float64
15	convertible_note	49438 non-null	float64
16	debt_financing	49438 non-null	float64
17	angel	49438 non-null	float64
18	grant	49438 non-null	float64
19	private_equity	49438 non-null	float64
20	post_ipo_equity	49438 non-null	float64
21	post_ipo_debt	49438 non-null	float64
22	secondary_market	49438 non-null	float64
23	product_crowdfunding	49438 non-null	float64
24	round_A	49438 non-null	float64
25	round_B	49438 non-null	float64
26	round_C	49438 non-null	float64
27	round_D	49438 non-null	float64
28	round_E	49438 non-null	float64
29	round_F	49438 non-null	float64

```

30 round_G          49438 non-null float64
31 round_H          49438 non-null float64
dtypes: float64(23), object(9)
memory usage: 12.4+ MB

```

```

, funding_total_usd 11,11,231

```

```

[ ]: data.columns=data.columns.str.strip()

data['funding_total_usd']=data['funding_total_usd'].str.replace(",","")
data["funding_total_usd"]=pd.
↳to_numeric(data["funding_total_usd"],errors="coerce").convert_dtypes()
funding_mode=data['funding_total_usd'].mode()[0]
data['funding_total_usd']=data["funding_total_usd"].fillna(funding_mode)

```

```

[ ]: data.head()

```

```

[ ]:
      name                homepage_url      market \
0      #waywire      http://www.waywire.com      News
1  &TV Communications      http://enjoyandtv.com      Games
2  'Rock' Your Paper      http://www.rockyourpaper.org      Publishing
3  (In)Touch Network      http://www.InTouchNetwork.com      Electronics
4  -R- Ranch and Mine      http://app.thotz.co/      Tourism

      funding_total_usd      status      country_code      state_code      region \
0      1750000      acquired      USA      NY      New York City
1      4000000      operating      USA      CA      Los Angeles
2      40000      operating      EST      CA      Tallinn
3      1500000      operating      GBR      CA      London
4      60000      operating      USA      TX      Dallas

      city      funding_rounds      ...      secondary_market      product_crowdfunding \
0      New York      1.0      ...      0.0      0.0
1      Los Angeles      2.0      ...      0.0      0.0
2      Tallinn      1.0      ...      0.0      0.0
3      London      1.0      ...      0.0      0.0
4      Fort Worth      2.0      ...      0.0      0.0

      round_A      round_B      round_C      round_D      round_E      round_F      round_G      round_H
0      0.0      0.0      0.0      0.0      0.0      0.0      0.0      0.0
1      0.0      0.0      0.0      0.0      0.0      0.0      0.0      0.0
2      0.0      0.0      0.0      0.0      0.0      0.0      0.0      0.0
3      0.0      0.0      0.0      0.0      0.0      0.0      0.0      0.0
4      0.0      0.0      0.0      0.0      0.0      0.0      0.0      0.0

```

```

[5 rows x 32 columns]

```

```

[ ]: data.shape

```

```
[ ]: (49438, 32)
```

```
[ ]: data.dtypes
```

```
[ ]: name                object
    homepage_url         object
    market               object
    funding_total_usd     Int64
    status                object
    country_code           object
    state_code            object
    region                object
    city                  object
    funding_rounds         float64
    founded_year           float64
    seed                   float64
    venture                 float64
    equity_crowdfunding    float64
    undisclosed            float64
    convertible_note       float64
    debt_financing         float64
    angel                  float64
    grant                  float64
    private_equity         float64
    post_ipo_equity        float64
    post_ipo_debt          float64
    secondary_market       float64
    product_crowdfunding   float64
    round_A                float64
    round_B                float64
    round_C                float64
    round_D                float64
    round_E                float64
    round_F                float64
    round_G                float64
    round_H                float64
    dtype: object
```

```
[ ]: data.select_dtypes('O').describe()
```

0.1.5 LabelEncoding

LabelEncoder

```
[ ]: encoder_name=LabelEncoder()
    encoder_market=LabelEncoder()
    encoder_country_code=LabelEncoder()
```



```

encoder_homepage_url=LabelEncoder()
encoder_status=LabelEncoder()
encoder_state_code=LabelEncoder()
encoder_city=LabelEncoder()
encoder_region=LabelEncoder()

```

```

[ ]: data['name']=encoder_name.fit_transform(data['name'])
data['market']=encoder_market.fit_transform(data['market'])
data['country_code']=encoder_country_code.fit_transform(data['country_code'])
data['status']=encoder_homepage_url.fit_transform(data['status'])
data['homepage_url']=encoder_homepage_url.fit_transform(data['homepage_url'])
data['state_code']=encoder_state_code.fit_transform(data['state_code'])
data['city']=encoder_city.fit_transform(data['city'])
data['region']=encoder_region.fit_transform(data['region'])

```

```

[ ]: data.head()

```

```

[ ]:
  name  homepage_url  market  funding_total_usd  status  country_code  \
0     0           43610     465           1750000         0           110
1     1           4422     277           4000000         2           110
2     2          37197     543            40000         2           35
3     3          15435     211          1500000         2           38
4     4           1124     683            60000         2           110

  state_code  region  city  funding_rounds  ...  secondary_market  \
0          40     699  2547             1.0  ...              0.0
1           6     570  2098             2.0  ...              0.0
2           6     956  3645             1.0  ...              0.0
3           6     568  2085             1.0  ...              0.0
4          53     251  1234             2.0  ...              0.0

  product_crowdfunding  round_A  round_B  round_C  round_D  round_E  round_F  \
0                   0.0      0.0      0.0      0.0      0.0      0.0
1                   0.0      0.0      0.0      0.0      0.0      0.0
2                   0.0      0.0      0.0      0.0      0.0      0.0
3                   0.0      0.0      0.0      0.0      0.0      0.0
4                   0.0      0.0      0.0      0.0      0.0      0.0

  round_G  round_H
0       0.0      0.0
1       0.0      0.0
2       0.0      0.0
3       0.0      0.0
4       0.0      0.0

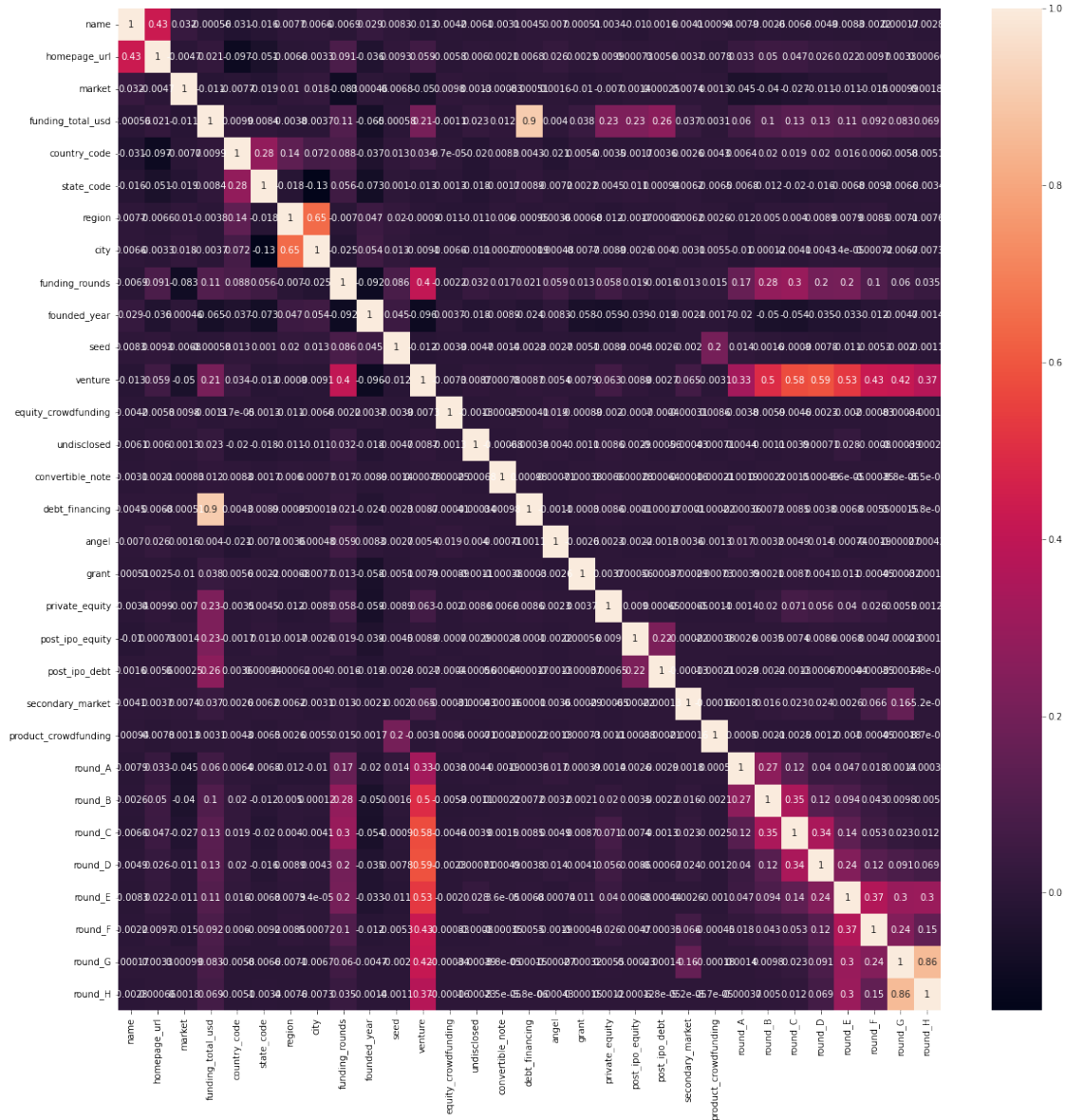
```

```

[5 rows x 32 columns]

```

```
[ ]: data_corr=data.drop("status",axis=1).corr()
plt.figure(figsize=(20,20))
sns.heatmap(data_corr,annot=True)
plt.show()
```



```
[ ]: threshold=0.5
def correlation_fun(ds,threshold):
    corr_col=set()
    corr_mat=ds.corr()
```

```

for i in range(len(corr_mat.columns)):
    for j in range(i):
        if abs(corr_mat.iloc[i,j])>threshold:
            colname=corr_mat.columns[i]
            corr_col.add(colname)
return corr_col

```

```
[ ]: correlation_fun(data.drop("status",axis=1),threshold)
```

```
[ ]: {'city', 'debt_financing', 'round_C', 'round_D', 'round_E', 'round_H'}
```

```
[ ]: data=data.drop(['city', 'round_C', 'round_D', 'round_E', 'round_H'],axis=1)
```

0.1.6

```
[ ]: target = data[target_col]
data_X_train, data_X_test, data_y_train, data_y_test = train_test_split(data,
    ↪target, test_size=0.2, random_state=1)
```

```
[ ]: data_X_train.shape
```

```
[ ]: (39550, 27)
```

```
[ ]: data_X_test.shape
```

```
[ ]: (9888, 27)
```

```
[ ]: np.unique(target)
```

```
[ ]: array([0, 1, 2])
```

0.1.7

```
[ ]: svr_1 = svm.LinearSVC()
svr_1.fit(data_X_train, data_y_train)
```

```

/home/artiom/.local/lib/python3.8/site-packages/sklearn/svm/_base.py:1206:
ConvergenceWarning: Liblinear failed to converge, increase the number of
iterations.
  warnings.warn(

```

```
[ ]: LinearSVC()
```

```
[ ]: data_y_pred_1 = svr_1.predict(data_X_test)
accuracy_score(data_y_test, data_y_pred_1)
```

```
[ ]: 0.8080501618122977
```

```
[ ]: f1_score(data_y_test, data_y_pred_1, average='micro')
```

```
[ ]: 0.8080501618122977
```

```
[ ]: f1_score(data_y_test, data_y_pred_1, average='macro')
```

```
[ ]: 0.3295056626437861
```

```
[ ]: f1_score(data_y_test, data_y_pred_1, average='weighted')
```

```
[ ]: 0.7868994820528529
```

```
[ ]: svr_2 = svm.LinearSVC(C=1.0, max_iter=10000)
svr_2.fit(data_X_train, data_y_train)
```

```
/home/artiom/.local/lib/python3.8/site-packages/sklearn/svm/_base.py:1206:
ConvergenceWarning: Liblinear failed to converge, increase the number of
iterations.
  warnings.warn(
```

```
[ ]: LinearSVC(max_iter=10000)
```

```
[ ]: data_y_pred_2 = svr_2.predict(data_X_test)
accuracy_score(data_y_test, data_y_pred_2)
```

```
[ ]: 0.6844660194174758
```

```
[ ]: f1_score(data_y_test, data_y_pred_2, average='micro')
```

```
[ ]: 0.6844660194174758
```

```
[ ]: f1_score(data_y_test, data_y_pred_2, average='macro')
```

```
[ ]: 0.35340656910368473
```

```
[ ]: f1_score(data_y_test, data_y_pred_2, average='weighted')
```

```
[ ]: 0.7253240978043891
```

```
[ ]: svr_3 = svm.LinearSVC(C=1.0, penalty='l1', dual=False, max_iter=10000)
svr_3.fit(data_X_train, data_y_train)
```

```
/home/artiom/.local/lib/python3.8/site-packages/sklearn/svm/_base.py:1206:
ConvergenceWarning: Liblinear failed to converge, increase the number of
iterations.
  warnings.warn(
```

```
[ ]: LinearSVC(dual=False, max_iter=10000, penalty='l1')
```

```
[ ]: data_y_pred_3_0 = svr_3.predict(data_X_train)
      accuracy_score(data_y_train, data_y_pred_3_0)
```

```
[ ]: 0.9997218710493047
```

```
[ ]: data_y_pred_3 = svr_3.predict(data_X_test)
      accuracy_score(data_y_test, data_y_pred_3)
```

```
[ ]: 0.9998988673139159
```

```
[ ]: f1_score(data_y_test, data_y_pred_3, average='micro')
```

```
[ ]: 0.9998988673139159
```

```
[ ]: f1_score(data_y_test, data_y_pred_3, average='weighted')
```

```
[ ]: 0.9998988213894379
```

0.1.8

```
[ ]: ab1 = AdaBoostClassifier()
      ab1.fit(data_X_train, data_y_train)
      data_y_pred_1 = ab1.predict(data_X_test)
      data_y_pred_1_0 = ab1.predict(data_X_train)
      accuracy_score(data_y_train, data_y_pred_1_0)
```

```
[ ]: 1.0
```

```
[ ]: accuracy_score(data_y_test, data_y_pred_1)
```

```
[ ]: 1.0
```

```
[ ]: f1_score(data_y_test, data_y_pred_1, average='micro')
```

```
[ ]: 1.0
```

```
[ ]: f1_score(data_y_test, data_y_pred_1, average='macro')
```

```
[ ]: 1.0
```

```
[ ]: f1_score(data_y_test, data_y_pred_1, average='weighted')
```

```
[ ]: 1.0
```

0.1.9

```

100% .
: svr_3 = svm.LinearSVC(C=1.0, penalty='l1', dual=False, max_iter=10000)
100% . ,
, ,
, .
, , .
```