

task2

April 14, 2024

0.1 CREDIT CARD FRAUD DETECTION

```
[120]: import pandas as pd
import numpy as np
data = pd.read_csv("/content/fraudTest.csv")
data.head()
```

```
[120]: Unnamed: 0 trans_date_trans_time cc_num \
0 0 2020-06-21 12:14:25 2291163933867244
1 1 2020-06-21 12:14:33 3573030041201292
2 2 2020-06-21 12:14:53 3598215285024754
3 3 2020-06-21 12:15:15 3591919803438423
4 4 2020-06-21 12:15:17 3526826139003047

merchant category amt first \
0 fraud_Kirlin and Sons personal_care 2.86 Jeff
1 fraud_Sporer-Keebler personal_care 29.84 Joanne
2 fraud_Swaniawski, Nietzsche and Welch health_fitness 41.28 Ashley
3 fraud_Haley Group misc_pos 60.05 Brian
4 fraud_Johnston-Casper travel 3.19 Nathan

last gender street ... lat long \
0 Elliott M 351 Darlene Green ... 33.9659 -80.9355
1 Williams F 3638 Marsh Union ... 40.3207 -110.4360
2 Lopez F 9333 Valentine Point ... 40.6729 -73.5365
3 Williams M 32941 Krystal Mill Apt. 552 ... 28.5697 -80.8191
4 Massey M 5783 Evan Roads Apt. 465 ... 44.2529 -85.0170

city_pop job dob \
0 333497 Mechanical engineer 1968-03-19
1 302 Sales professional, IT 1990-01-17
2 34496 Librarian, public 1970-10-21
3 54767 Set designer 1987-07-25
4 1126 Furniture designer 1955-07-06

trans_num unix_time merch_lat merch_long \
0 2da90c7d74bd46a0caf3777415b3ebd3 1371816865 33.986391 -81.200714
1 324cc204407e99f51b0d6ca0055005e7 1371816873 39.450498 -109.960431
```

```

2  c81755dbbba9d5c77f094348a7579be  1371816893  40.495810  -74.196111
3  2159175b9efe66dc301f149d3d5abf8c  1371816915  28.812398  -80.883061
4  57ff021bd3f328f8738bb535c302a31b  1371816917  44.959148  -85.884734

```

```

      is_fraud
0           0
1           0
2           0
3           0
4           0

```

[5 rows x 23 columns]

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

```

[121]: Unnamed: 0          0
      trans_date_trans_time  0
      cc_num                0
      merchant              0
      category              0
      amt                   0
      first                 0
      last                  0
      gender                0
      street                0
      city                  0
      state                 0
      zip                   0
      lat                   0
      long                  0
      city_pop              0
      job                   0
      dob                   0
      trans_num             0
      unix_time             0
      merch_lat             0
      merch_long            0
      is_fraud              0
      dtype: int64

```

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

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 555719 entries, 0 to 555718
Data columns (total 23 columns):
 #   Column                Non-Null Count  Dtype
---  -

```

```

0  Unnamed: 0          555719 non-null int64
1  trans_date_trans_time 555719 non-null object
2  cc_num              555719 non-null int64
3  merchant            555719 non-null object
4  category            555719 non-null object
5  amt                 555719 non-null float64
6  first               555719 non-null object
7  last                555719 non-null object
8  gender              555719 non-null object
9  street              555719 non-null object
10 city                555719 non-null object
11 state               555719 non-null object
12 zip                 555719 non-null int64
13 lat                 555719 non-null float64
14 long                555719 non-null float64
15 city_pop            555719 non-null int64
16 job                 555719 non-null object
17 dob                 555719 non-null object
18 trans_num           555719 non-null object
19 unix_time           555719 non-null int64
20 merch_lat           555719 non-null float64
21 merch_long          555719 non-null float64
22 is_fraud            555719 non-null int64

```

dtypes: float64(5), int64(6), object(12)

memory usage: 97.5+ MB

```
[123]: data.describe()
```

```

[123]:      Unnamed: 0      cc_num      amt      zip  \
count  555719.000000  5.557190e+05  555719.000000  555719.000000
mean    277859.000000  4.178387e+17    69.392810   48842.628015
std     160422.401459  1.309837e+18   156.745941   26855.283328
min         0.000000  6.041621e+10    1.000000    1257.000000
25%     138929.500000  1.800429e+14    9.630000   26292.000000
50%     277859.000000  3.521417e+15   47.290000   48174.000000
75%     416788.500000  4.635331e+15   83.010000   72011.000000
max     555718.000000  4.992346e+18  22768.110000  99921.000000

      lat      long      city_pop      unix_time  \
count  555719.000000  555719.000000  5.557190e+05  5.557190e+05
mean     38.543253   -90.231325  8.822189e+04  1.380679e+09
std       5.061336   13.721780  3.003909e+05  5.201104e+06
min      20.027100  -165.672300  2.300000e+01  1.371817e+09
25%      34.668900   -96.798000  7.410000e+02  1.376029e+09
50%      39.371600   -87.476900  2.408000e+03  1.380762e+09
75%      41.894800   -80.175200  1.968500e+04  1.385867e+09
max      65.689900   -67.950300  2.906700e+06  1.388534e+09

```

	merch_lat	merch_long	is_fraud
count	555719.000000	555719.000000	555719.000000
mean	38.542798	-90.231380	0.003860
std	5.095829	13.733071	0.062008
min	19.027422	-166.671575	0.000000
25%	34.755302	-96.905129	0.000000
50%	39.376593	-87.445204	0.000000
75%	41.954163	-80.264637	0.000000
max	66.679297	-66.952026	1.000000

```
[124]: data['city_pop'].fillna(data['city_pop'].median(), inplace=True)
data['unix_time'].fillna(data['unix_time'].median(), inplace=True)
data['merch_lat'].fillna(data['merch_lat'].median(), inplace=True)
data['merch_long'].fillna(data['merch_long'].median(), inplace=True)
data['is_fraud'].fillna(0, inplace=True)
```

```
[125]: data.dropna(subset=['unix_time', 'merch_lat', 'merch_long', 'is_fraud'],
    ↪inplace=True)
```

```
[126]: # Check for missing values in the entire dataset
missing_values = data.isnull().sum()
print(missing_values)
```

```
Unnamed: 0      0
trans_date_trans_time  0
cc_num          0
merchant        0
category        0
amt            0
first          0
last           0
gender         0
street         0
city           0
state          0
zip            0
lat            0
long           0
city_pop       0
job            0
dob            0
trans_num      0
unix_time      0
merch_lat      0
merch_long     0
is_fraud       0
```

dtype: int64

```
[127]: X = data.drop('is_fraud', axis=1)
       y = data['is_fraud']
```

```
[128]: data['Unnamed: 0'],unnamed_name=pd.factorize(data['Unnamed: 0'])
       print(unnamed_name)
```

```
Index([    0,     1,     2,     3,     4,     5,     6,     7,     8,
         9,
        ...
       555709, 555710, 555711, 555712, 555713, 555714, 555715, 555716, 555717,
       555718],
      dtype='int64', length=555719)
```

```
[129]: data['cc_num'],cc_name=pd.factorize(data['cc_num'])
       print(cc_name)
```

```
Index([2291163933867244, 3573030041201292, 3598215285024754, 3591919803438423,
       3526826139003047,  30407675418785,  213180742685905, 3589289942931264,
       3596357274378601, 3546897637165774,
        ...
       3550412175018089,    586100864972,  372965408103277,  180020605265701,
       347399333635231,    4883407061576,  4295296907373, 4087542780207162,
       3588001568691267, 2242176657877538],
      dtype='int64', length=924)
```

```
[130]: data['category'],category_name=pd.factorize(data['category'])
       print(category_name)
```

```
Index(['personal_care', 'health_fitness', 'misc_pos', 'travel', 'kids_pets',
       'shopping_pos', 'food_dining', 'home', 'entertainment', 'shopping_net',
       'misc_net', 'grocery_pos', 'gas_transport', 'grocery_net'],
      dtype='object')
```

```
[131]: data['trans_date_trans_time'],time_name=pd.
       ↪factorize(data['trans_date_trans_time'])
       print(time_name)
```

```
Index(['2020-06-21 12:14:25', '2020-06-21 12:14:33', '2020-06-21 12:14:53',
       '2020-06-21 12:15:15', '2020-06-21 12:15:17', '2020-06-21 12:15:37',
       '2020-06-21 12:15:44', '2020-06-21 12:15:50', '2020-06-21 12:16:10',
       '2020-06-21 12:16:11',
        ...
       '2020-12-31 23:57:18', '2020-12-31 23:57:50', '2020-12-31 23:57:56',
       '2020-12-31 23:58:04', '2020-12-31 23:58:34', '2020-12-31 23:59:07',
       '2020-12-31 23:59:09', '2020-12-31 23:59:15', '2020-12-31 23:59:24',
       '2020-12-31 23:59:34'],
```

```
dtype='object', length=544760)
```

```
[132]: data['amt'],amt_name=pd.factorize(data['amt'])  
print(amt_name)
```

```
Index([ 2.86, 29.84, 41.28, 60.05, 3.19, 19.55, 133.93, 10.37,  
       4.37, 66.54,  
       ...  
       2149.66, 537.02, 1309.21, 256.67, 500.31, 850.87, 516.74, 255.42,  
       302.79, 1164.37],  
      dtype='float64', length=37256)
```

```
[133]: data['merchant'],merchant_name=pd.factorize(data['merchant'])  
print(merchant_name)
```

```
Index(['fraud_Kirlin and Sons', 'fraud_Sporer-Keebler',  
      'fraud_Swaniawski, Nitzsche and Welch', 'fraud_Haley Group',  
      'fraud_Johnston-Casper', 'fraud_Daugherty LLC', 'fraud_Romaguera Ltd',  
      'fraud_Reichel LLC', 'fraud_Goyette, Howell and Collier',  
      'fraud_Kilback Group',  
      ...  
      'fraud_Rippin, Kub and Mann', 'fraud_Rempel PLC',  
      'fraud_Leannon-Nikolaus', 'fraud_Monahan, Hermann and Johns',  
      'fraud_Block-Hauck', 'fraud_Hagenes, Hermann and Stroman',  
      'fraud_Hermann-Gaylord', 'fraud_Mante Group', 'fraud_Corwin-Gorczy',  
      'fraud_McCullough Group'],  
      dtype='object', length=693)
```

```
[134]: data['zip'],zip_name=pd.factorize(data['zip'])  
print(zip_name)
```

```
Index([29209, 84002, 11710, 32780, 49632, 14816, 95528, 57374, 16858, 76678,  
      ...  
      40502, 13795, 87417, 66958, 65745, 98118, 52658, 73044, 99921, 38668],  
      dtype='int64', length=912)
```

```
[135]: data['lat'],lat_name=pd.factorize(data['lat'])  
print(lat_name)
```

```
Index([33.9659, 40.3207, 40.6729, 28.5697, 44.2529, 42.1939, 40.507, 43.7557,  
      41.0001, 31.6591,  
      ...  
      38.0174, 42.0695, 36.741, 39.8616, 36.5276, 47.5412, 40.7067, 35.833,  
      55.4732, 34.6323],  
      dtype='float64', length=910)
```

```
[136]: data['long'],long_name=pd.factorize(data['long'])  
print(long_name)
```

```
Index([
      -80.9355,      -110.436,      -73.5365,
      -80.8191, -85.01700000000001,      -76.7361,
      -123.9743,      -97.5936,      -78.2357,
      -96.8094,
      ...
      -84.4854,      -75.7967,      -108.351,
      -97.1825,      -93.9359,      -122.275,
      -91.2268,      -97.436,      -133.1171,
      -89.8855],
      dtype='float64', length=910)
```

```
[137]: data['city_pop'],city_name=pd.factorize(data['city_pop'])
       print(city_name)
```

```
Index([333497,    302,  34496,  54767,   1126,    520,   1139,    343,   3688,
        263,
        ...
        296965,  3800,   6910,    314,   2693, 837792,   1071,  20226,   1920,
        14462],
      dtype='int64', length=835)
```

```
[138]: data['is_fraud'],fraud_name=pd.factorize(data['is_fraud'])
       print(fraud_name)
```

```
Index([0, 1], dtype='int64')
```

```
[139]: data['first'],first_name=pd.factorize(data['first'])
       print(first_name)
```

```
Index(['Jeff', 'Joanne', 'Ashley', 'Brian', 'Nathan', 'Danielle', 'Kayla',
      'Paula', 'David', 'Samuel',
      ...
      'Katelyn', 'Wesley', 'Sonya', 'Collin', 'Tommy', 'Guy', 'Dennis',
      'Bruce', 'Evan', 'Nicole'],
      dtype='object', length=341)
```

```
[140]: data['last'],last_name=pd.factorize(data['last'])
       print(last_name)
```

```
Index(['Elliott', 'Williams', 'Lopez', 'Massey', 'Evans', 'Sutton', 'Estrada',
      'Everett', 'Obrien', 'Jenkins',
      ...
      'Bridges', 'Raymond', 'Davidson', 'Osborne', 'Webster', 'Freeman',
      'Bartlett', 'Santiago', 'Bates', 'Robbins'],
      dtype='object', length=471)
```

```
[141]: data['street'],street_name=pd.factorize(data['street'])
       print(street_name)
```

```
Index(['351 Darlene Green', '3638 Marsh Union', '9333 Valentine Point',
      '32941 Krystal Mill Apt. 552', '5783 Evan Roads Apt. 465',
      '76752 David Lodge Apt. 064', '010 Weaver Land', '350 Stacy Glens',
      '4138 David Fall', '7921 Robert Port Suite 343',
      ...
      '742 Sellers Ferry', '4481 Maldonado Hollow',
      '53199 Laurie Mills Apt. 864', '7908 Derrick Mount',
      '13128 Hall Station Suite 588', '6386 Bailey Hill Apt. 421',
      '007 Tonya Isle Suite 299', '537 Brian Island', '5942 Thomas Park',
      '1327 Rose Causeway Apt. 610'],
      dtype='object', length=924)
```

```
[142]: data['job'], job_name=pd.factorize(data['job'])
       print(job_name)
```

```
Index(['Mechanical engineer', 'Sales professional, IT', 'Librarian, public',
      'Set designer', 'Furniture designer', 'Psychotherapist',
      'Therapist, occupational', 'Development worker, international aid',
      'Advice worker', 'Barrister',
      ...
      'Medical technical officer', 'Charity officer', 'Administrator, arts',
      'Occupational therapist', 'Solicitor, Scotland', 'Sports administrator',
      'Artist', 'Engineer, water', 'Operational investment banker',
      'Software engineer'],
      dtype='object', length=478)
```

```
[143]: data['dob'], dob_name=pd.factorize(data['dob'])
       print(dob_name)
```

```
Index(['1968-03-19', '1990-01-17', '1970-10-21', '1987-07-25', '1955-07-06',
      '1991-10-13', '1951-01-15', '1972-03-05', '1973-05-27', '1956-05-30',
      ...
      '1962-12-30', '1968-07-06', '1956-02-02', '2002-03-17', '1968-02-05',
      '1936-12-23', '1998-08-02', '1969-11-08', '1997-06-17', '1959-03-03'],
      dtype='object', length=910)
```

```
[144]: data['trans_num'], trans_name=pd.factorize(data['trans_num'])
       print(trans_name)
```

```
Index(['2da90c7d74bd46a0caf3777415b3ebd3', '324cc204407e99f51b0d6ca0055005e7',
      'c81755dbbbee9d5c77f094348a7579be', '2159175b9efe66dc301f149d3d5abf8c',
      '57ff021bd3f328f8738bb535c302a31b', '798db04aaceb4febd084f1a7c404da93',
      '17003d7ce534440eadb10c4750e020e5', '8be473af4f05fc6146ea55ace73e7ca2',
      '71a1da150d1ce510193d7622e08e784e', 'a7915132c7c4240996ba03a47f81e3bd',
      ...
      'a7105564935ea3977dc61ff9ced3bf5e', '9fc9f6f9be3182d519a61a119cf97199',
      'a8310343c189e4a5b6316050d2d6b014', 'bd7071fd5c9510a5594ee196368ac80e',
      '6d04313bfe4b661b8ca2b6a499a320fe', '9b1f753c79894c9f4b71f04581835ada',
```



```
'2090647dac2c89a1d86c514c427f5b91', '6c5b7c8add471975aa0fec023b2e8408',
'14392d723bb7737606b2700ac791b7aa', '1765bb45b3aa3224b4cdcb6e7a96cee3'],
dtype='object', length=555719)
```

```
[145]: data['gender'],gender_name=pd.factorize(data['gender'])
print(gender_name)
```

```
Index(['M', 'F'], dtype='object')
```

```
[146]: data['city'],city_name=pd.factorize(data['city'])
print(city_name)
```

```
Index(['Columbia', 'Altonah', 'Bellmore', 'Titusville', 'Falmouth',
'Breesport', 'Carlotta', 'Spencer', 'Morrisdale', 'Prairie Hill',
...
'Lexington', 'Kirkwood', 'Kirtland', 'Morrowville', 'Seligman',
'Seattle', 'Wever', 'Guthrie', 'Craig', 'Senatobia'],
dtype='object', length=849)
```

```
[147]: data['state'],state_name=pd.factorize(data['state'])
print(state_name)
```

```
Index(['SC', 'UT', 'NY', 'FL', 'MI', 'CA', 'SD', 'PA', 'TX', 'KY', 'WY', 'AL',
'LA', 'GA', 'CO', 'OH', 'WI', 'VT', 'AR', 'NJ', 'IA', 'MD', 'MS', 'KS',
'IL', 'MO', 'ME', 'TN', 'DC', 'AZ', 'MT', 'MN', 'OK', 'WA', 'WV', 'NM',
'MA', 'NE', 'VA', 'ID', 'OR', 'IN', 'NC', 'NH', 'ND', 'CT', 'NV', 'HI',
'RI', 'AK'],
dtype='object')
```

```
[148]: x=data.iloc[:,0:-1]
y=data.iloc[:,-1]
print(x)
print(y)
```

```

      Unnamed: 0  trans_date_trans_time  cc_num  merchant  category  amt  \
0              0              0          0          0          0      0
1              1              1          1          1          0      1
2              2              2          2          2          1      2
3              3              3          3          3          2      3
4              4              4          4          4          3      4
...           ...                   ...      ...      ...      ...
555714        555714        544755        757        296          1  5224
555715        555715        544756        136         35          4  5504
555716        555716        544757        607         39          4  5273
555717        555717        544758        350        163          3  2243
555718        555718        544759        250        173          8 13727
```

```
first last  gender  street  ...  zip  lat  long  city_pop  job  dob  \
```

0	0	0	0	0	...	0	0	0	0	0	0
1	1	1	1	1	...	1	1	1	1	1	1
2	2	2	1	2	...	2	2	2	2	2	2
3	3	1	0	3	...	3	3	3	3	3	3
4	4	3	0	4	...	4	4	4	4	4	4
...
555714	89	325	0	757	...	747	747	746	694	16	746
555715	102	109	0	136	...	136	136	136	135	127	136
555716	285	353	1	607	...	600	600	599	565	38	597
555717	204	232	0	350	...	348	348	347	338	262	347
555718	9	184	0	250	...	249	249	249	242	205	249

	trans_num	unix_time	merch_lat	merch_long
0	0	1371816865	33.986391	-81.200714
1	1	1371816873	39.450498	-109.960431
2	2	1371816893	40.495810	-74.196111
3	3	1371816915	28.812398	-80.883061
4	4	1371816917	44.959148	-85.884734
...
555714	555714	1388534347	39.946837	-91.333331
555715	555715	1388534349	29.661049	-96.186633
555716	555716	1388534355	46.658340	-119.715054
555717	555717	1388534364	44.470525	-117.080888
555718	555718	1388534374	36.210097	-97.036372

[555719 rows x 22 columns]

0	0
1	0
2	0
3	0
4	0
...	...
555714	0
555715	0
555716	0
555717	0
555718	0

Name: is_fraud, Length: 555719, dtype: int64

```
[149]: from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.2,random_state=0)
```

```
[150]: feature_names = x.columns
```

```
[155]: from sklearn.linear_model import LogisticRegression
clf = LogisticRegression()
clf.fit(x_train, y_train)
```

```
[155]: LogisticRegression()
```

```
[156]: y_pred = dtree.predict(x_test)
```

```
[157]: from sklearn.metrics import accuracy_score  
accuracy = accuracy_score(y_test, y_pred)  
print("LogisticRegression:")  
print("Accuracy:", accuracy)
```

```
LogisticRegression:  
Accuracy: 0.9961131505074498
```

```
[ ]:
```