

Лабораторная работа №3
по дисциплине
«Методы машинного обучения»
на тему
«Обработка пропусков, кодирование категориальных
признаков, масштабирование данных»

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1. Обработка пропусков в данных, кодирование категориальных признаков, масштабирование данных.

Мы научимся обрабатывать пропуски в данных для количественных (числовых) и категориальных признаков и масштабировать данные. Также мы научимся преобразовывать категориальные признаки в числовые.

```
[98]: import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
sns.set(style="ticks")
```

1.1. Загрузка и первичный анализ данных

Используем данные из соревнования House Prices: Advanced Regression Techniques

```
[99]: # Будем использовать только обучающую выборку
data = pd.read_csv('data/gun_violence.csv', sep=",")
```

```
[100]: # размер набора данных
data.shape
```

```
[100]: (162867, 29)
```

```
[101]: # типы колонок
data.dtypes
```

```
[101]: incident_id          int64
date                    object
state                   object
city_or_county          object
address                 object
n_killed                int64
n_injured               int64
incident_url            object
source_url              object
incident_url_fields_missing    bool
congressional_district  float64
gun_stolen              object
gun_type                object
incident_characteristics  object
latitude                float64
location_description     object
longitude               float64
n_guns_involved         float64
notes                   object
participant_age          object
participant_age_group    object
participant_gender       object
```

```

participant_name      object
participant_relationship  object
participant_status     object
participant_type       object
sources               object
state_house_district  float64
state_senate_district float64
dtype: object

```

```
[102]: # проверим есть ли пропущенные значения
data.isnull().sum()
```

```

[102]: incident_id      0
date                  0
state                0
city_or_county       0
address              12303
n_killed             0
n_injured            0
incident_url         0
source_url           276
incident_url_fields_missing 0
congressional_district 4865
gun_stolen           99311
gun_type             99299
incident_characteristics 242
latitude             4715
location_description 140476
longitude            4715
n_guns_involved      99299
notes                56008
participant_age       63464
participant_age_group 27678
participant_gender    23832
participant_name      84207
participant_relationship 152618
participant_status    18510
participant_type      16327
sources               516
state_house_district  24163
state_senate_district 20659
dtype: int64

```

```
[103]: # Первые 5 строк датасета
data.head()
```

```

[103]:   incident_id      date      state city_or_county \
0      461105  2013-01-01  Pennsylvania  Mckeesport
1      460726  2013-01-01   California  Hawthorne
2      478855  2013-01-01         Ohio      Lorain

```

3	478925	2013-01-05	Colorado	Aurora
4	478959	2013-01-07	North Carolina	Greensboro

	address	n_killed	n_injured
↪ \			
0	1506 Versailles Avenue and Coursin Street	0	4
1	13500 block of Cerise Avenue	1	3
2	1776 East 28th Street	1	3
3	16000 block of East Ithaca Place	4	0
4	307 Mourning Dove Terrace	2	2

	incident_url \
0	http://www.gunviolencearchive.org/incident/461105
1	http://www.gunviolencearchive.org/incident/460726
2	http://www.gunviolencearchive.org/incident/478855
3	http://www.gunviolencearchive.org/incident/478925
4	http://www.gunviolencearchive.org/incident/478959

	source_url \
0	http://www.post-gazette.com/local/south/2013/0...
1	http://www.dailybulletin.com/article/zz/201301...
2	http://chronicle.northcoastnow.com/2013/02/14/...
3	http://www.dailydemocrat.com/20130106/aurora-s...
4	http://www.journalnow.com/news/local/article_d...

	incident_url_fields_missing ...	
↪ participant_age \		
0	False ...	0::
↪ 20		
1	False ...	0::
↪ 20		
2	False ... 0::25 1::31 2::33 3::34 4::	
↪ 33		
3	False ... 0::29 1::33 2::56 3::	
↪ 33		
4	False ... 0::18 1::46 2::14 3::	
↪ 47		

	participant_age_group \
0	0::Adult 18+ 1::Adult 18+ 2::Adult 18+ 3::A...
1	0::Adult 18+ 1::Adult 18+ 2::Adult 18+ 3::A...
2	0::Adult 18+ 1::Adult 18+ 2::Adult 18+ 3::A...
3	0::Adult 18+ 1::Adult 18+ 2::Adult 18+ 3::A...
4	0::Adult 18+ 1::Adult 18+ 2::Teen 12-17 3::...

	participant_gender \
0	0::Male 1::Male 3::Male 4::Female
1	0::Male
2	0::Male 1::Male 2::Male 3::Male 4::Male

3	0::Female 1::Male 2::Male 3::Male
4	0::Female 1::Male 2::Male 3::Female

	participant_name \
0	0::Julian Sims
1	0::Bernard Gillis
2	0::Damien Bell 1::Desmen Noble 2::Herman Sea...
3	0::Stacie Philbrook 1::Christopher Ratliffe ...
4	0::Danielle Imani Jameison 1::Maurice Eugene ...

	participant_relationship \
0	NaN
1	NaN
2	NaN
3	NaN
4	3::Family

	participant_status \
0	0::Arrested 1::Injured 2::Injured 3::Injure...
1	0::Killed 1::Injured 2::Injured 3::Injured
2	0::Injured, Unharmed, Arrested 1::Unharmed, A...
3	0::Killed 1::Killed 2::Killed 3::Killed
4	0::Injured 1::Injured 2::Killed 3::Killed

	participant_type \
0	0::Victim 1::Victim 2::Victim 3::Victim 4::...
1	0::Victim 1::Victim 2::Victim 3::Victim 4::...
2	0::Subject-Suspect 1::Subject-Suspect 2::Vic...
3	0::Victim 1::Victim 2::Victim 3::Subject-Su...
4	0::Victim 1::Victim 2::Victim 3::Subject-Su...

	sources
	↪state_house_district \
0	http://pittsburgh.cbslocal.com/2013/01/01/4-pe... ↪ NaN
1	http://losangeles.cbslocal.com/2013/01/01/man-... ↪ 62.0
2	http://www.morningjournal.com/general-news/201... ↪ 56.0
3	http://denver.cbslocal.com/2013/01/06/officer-... ↪ 40.0
4	http://myfox8.com/2013/01/08/update-mother-sho... ↪ 62.0

	state_senate_district
0	NaN
1	35.0
2	13.0
3	28.0

4 27.0

[5 rows x 29 columns]

```
[104]: total_count = data.shape[0]
print('Всего строк: {}'.format(total_count))
```

Всего строк: 162867

2. 1. Обработка пропусков в данных

2.1. 1.1. Простые стратегии - удаление или заполнение нулями

```
[105]: # Удаление колонок, содержащих пустые значения
data_new_1 = data.dropna(axis=1, how='any')
(data.shape, data_new_1.shape)
```

```
[105]: ((162867, 29), (162867, 8))
```

```
[106]: # Удаление строк, содержащих пустые значения
data_new_2 = data.dropna(axis=0, how='any')
(data.shape, data_new_2.shape)
```

```
[106]: ((162867, 29), (450, 29))
```

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

```
[107]:
```

	incident_id	date	state	city_or_county \			
0	461105	2013-01-01	Pennsylvania	McKeesport			
1	460726	2013-01-01	California	Hawthorne			
2	478855	2013-01-01	Ohio	Lorain			
3	478925	2013-01-05	Colorado	Aurora			
4	478959	2013-01-07	North Carolina	Greensboro			

		address	n_killed	n_injured
0	1506 Versailles Avenue and Coursin Street		0	4
1	13500 block of Cerise Avenue		1	3
2	1776 East 28th Street		1	3
3	16000 block of East Ithaca Place		4	0
4	307 Mourning Dove Terrace		2	2

	incident_url \
0	http://www.gunviolencearchive.org/incident/461105
1	http://www.gunviolencearchive.org/incident/460726
2	http://www.gunviolencearchive.org/incident/478855
3	http://www.gunviolencearchive.org/incident/478925
4	http://www.gunviolencearchive.org/incident/478959

source_url \

```

0 http://www.post-gazette.com/local/south/2013/0...
1 http://www.dailybulletin.com/article/zz/201301...
2 http://chronicle.northcoastnow.com/2013/02/14/...
3 http://www.dailydemocrat.com/20130106/aurora-s...
4 http://www.journalnow.com/news/local/article_d...

    incident_url_fields_missing ...
↪ participant_age \
0          False ... 0::
↪ 20
1          False ... 0::
↪ 20
2          False ... 0::25||1::31||2::33||3::34||4::
↪ 33
3          False ... 0::29||1::33||2::56||3::
↪ 33
4          False ... 0::18||1::46||2::14||3::
↪ 47

                                participant_age_group \
0  0::Adult 18+||1::Adult 18+||2::Adult 18+||3::A...
1  0::Adult 18+||1::Adult 18+||2::Adult 18+||3::A...
2  0::Adult 18+||1::Adult 18+||2::Adult 18+||3::A...
3  0::Adult 18+||1::Adult 18+||2::Adult 18+||3::A...
4  0::Adult 18+||1::Adult 18+||2::Teen 12-17||3::...

                                participant_gender \
0          0::Male||1::Male||3::Male||4::Female
1          0::Male
2  0::Male||1::Male||2::Male||3::Male||4::Male
3          0::Female||1::Male||2::Male||3::Male
4          0::Female||1::Male||2::Male||3::Female

                                participant_name \
0          0::Julian Sims
1          0::Bernard Gillis
2  0::Damien Bell||1::Desmen Noble||2::Herman Sea...
3  0::Stacie Philbrook||1::Christopher Ratliffe||...
4  0::Danielle Imani Jameison||1::Maurice Eugene ...

    participant_relationship \
0          NaN
1          NaN
2          NaN
3          NaN
4          3::Family

                                participant_status \
0  0::Arrested||1::Injured||2::Injured||3::Injure...

```

```

1      0::Killed||1::Injured||2::Injured||3::Injured
2 0::Injured, Unharmed, Arrested||1::Unharmed, A...
3      0::Killed||1::Killed||2::Killed||3::Killed
4      0::Injured||1::Injured||2::Killed||3::Killed

```

```

                                participant_type \
0  0::Victim||1::Victim||2::Victim||3::Victim||4:...
1  0::Victim||1::Victim||2::Victim||3::Victim||4:...
2  0::Subject-Suspect||1::Subject-Suspect||2::Vic...
3  0::Victim||1::Victim||2::Victim||3::Subject-Su...
4  0::Victim||1::Victim||2::Victim||3::Subject-Su...

```

```

                                sources
→state_house_district \
0  http://pittsburgh.cbslocal.com/2013/01/01/4-pe...
→  NaN
1  http://losangeles.cbslocal.com/2013/01/01/man-...
→  62.0
2  http://www.morningjournal.com/general-news/201...
→  56.0
3  http://denver.cbslocal.com/2013/01/06/officer-...
→  40.0
4  http://myfox8.com/2013/01/08/update-mother-sho...
→  62.0

```

```

state_senate_district
0      NaN
1      35.0
2      13.0
3      28.0
4      27.0

```

[5 rows x 29 columns]

```

[108]: # Заполнение всех пропущенных значений нулями
# В данном случае это некорректно, так как нулями заполняются в
→ том числе категориальные колонки
data_new_3 = data.fillna(0)
data_new_3.head()

```

```

[108]: incident_id      date      state city_or_county \
0      461105  2013-01-01  Pennsylvania  Mckeesport
1      460726  2013-01-01  California    Hawthorne
2      478855  2013-01-01      Ohio      Lorain
3      478925  2013-01-05  Colorado      Aurora
4      478959  2013-01-07 North Carolina  Greensboro

                                address  n_killed  n_injured
→ \

```


0	1506 Versailles Avenue and Coursin Street	0	4
1	13500 block of Cerise Avenue	1	3
2	1776 East 28th Street	1	3
3	16000 block of East Ithaca Place	4	0
4	307 Mourning Dove Terrace	2	2

	incident_url \
0	http://www.gunviolencearchive.org/incident/461105
1	http://www.gunviolencearchive.org/incident/460726
2	http://www.gunviolencearchive.org/incident/478855
3	http://www.gunviolencearchive.org/incident/478925
4	http://www.gunviolencearchive.org/incident/478959

	source_url \
0	http://www.post-gazette.com/local/south/2013/0...
1	http://www.dailybulletin.com/article/zz/201301...
2	http://chronicle.northcoastnow.com/2013/02/14/...
3	http://www.dailydemocrat.com/20130106/aurora-s...
4	http://www.journalnow.com/news/local/article_d...

	incident_url_fields_missing ...	
	→ participant_age \	
0	False ...	0::
	→ 20	
1	False ...	0::
	→ 20	
2	False ... 0::25 1::31 2::33 3::34 4::	
	→ 33	
3	False ... 0::29 1::33 2::56 3::	
	→ 33	
4	False ... 0::18 1::46 2::14 3::	
	→ 47	

	participant_age_group \
0	0::Adult 18+ 1::Adult 18+ 2::Adult 18+ 3::A...
1	0::Adult 18+ 1::Adult 18+ 2::Adult 18+ 3::A...
2	0::Adult 18+ 1::Adult 18+ 2::Adult 18+ 3::A...
3	0::Adult 18+ 1::Adult 18+ 2::Adult 18+ 3::A...
4	0::Adult 18+ 1::Adult 18+ 2::Teen 12-17 3::...

	participant_gender \
0	0::Male 1::Male 3::Male 4::Female
1	0::Male
2	0::Male 1::Male 2::Male 3::Male 4::Male
3	0::Female 1::Male 2::Male 3::Male
4	0::Female 1::Male 2::Male 3::Female

	participant_name \
0	0::Julian Sims

```

1                                0::Bernard Gillis
2  0::Damien Bell||1::Desmen Noble||2::Herman Sea...
3  0::Stacie Philbrook||1::Christopher Ratliffe||...
4  0::Danielle Imani Jameison||1::Maurice Eugene ...

    participant_relationship \
0                                0
1                                0
2                                0
3                                0
4                                3::Family

                                participant_status \
0  0::Arrested||1::Injured||2::Injured||3::Injure...
1      0::Killed||1::Injured||2::Injured||3::Injured
2  0::Injured, Unharmed, Arrested||1::Unharmed, A...
3      0::Killed||1::Killed||2::Killed||3::Killed
4      0::Injured||1::Injured||2::Killed||3::Killed

                                participant_type \
0  0::Victim||1::Victim||2::Victim||3::Victim||4:...
1  0::Victim||1::Victim||2::Victim||3::Victim||4:...
2  0::Subject-Suspect||1::Subject-Suspect||2::Vic...
3  0::Victim||1::Victim||2::Victim||3::Subject-Su...
4  0::Victim||1::Victim||2::Victim||3::Subject-Su...

                                sources
→state_house_district \
0  http://pittsburgh.cbslocal.com/2013/01/01/4-pe...
→  0.0
1  http://losangeles.cbslocal.com/2013/01/01/man-...
→  62.0
2  http://www.morningjournal.com/general-news/201...
→  56.0
3  http://denver.cbslocal.com/2013/01/06/officer-...
→  40.0
4  http://myfox8.com/2013/01/08/update-mother-sho...
→  62.0

    state_senate_district
0                                0.0
1                                35.0
2                                13.0
3                                28.0
4                                27.0

[5 rows x 29 columns]

```

2.2. 1.2. “Внедрение значений” - импьютация (imputation)

2.2.1. 1.2.1. Обработка пропусков в числовых данных

```
[109]: # Выберем числовые колонки с пропущенными значениями
# Цикл по колонкам датасета
num_cols = []
for col in data.columns:
    # Количество пустых значений
    temp_null_count = data[data[col].isnull()].shape[0]
    dt = str(data[col].dtype)
    if temp_null_count>0 and (dt=='float64' or dt=='int64'):
        num_cols.append(col)
        temp_perc = round((temp_null_count / total_count) * 100.
        ↪0, 2)
        print('Колонка {}. Тип данных {}. Количество пустых
        ↪значений {}, {}%.'.format(col, dt, temp_null_count, temp_perc))
```

Колонка congressional_district. Тип данных float64. Количество
↪пустых значений

4865, 2.99%.

Колонка latitude. Тип данных float64. Количество пустых значений
↪4715, 2.9%.

Колонка longitude. Тип данных float64. Количество пустых значений
↪4715, 2.9%.

Колонка n_guns_involved. Тип данных float64. Количество пустых
↪значений 99299,
60.97%.

Колонка state_house_district. Тип данных float64. Количество
↪пустых значений
24163, 14.84%.

Колонка state_senate_district. Тип данных float64. Количество
↪пустых значений
20659, 12.68%.

```
[110]: # Фильтр по колонкам с пропущенными значениями
data_num = data[num_cols]
data_num
```

```
[110]: congressional_district  latitude  longitude  □
        ↪n_guns_involved  \
0                14.0    40.3467    -79.8559    □
        ↪NaN
1                43.0    33.9090   -118.3330    □
        ↪NaN
2                9.0    41.4455    -82.1377    □
        ↪2.0
3                6.0    39.6518   -104.8020    □
        ↪NaN
```

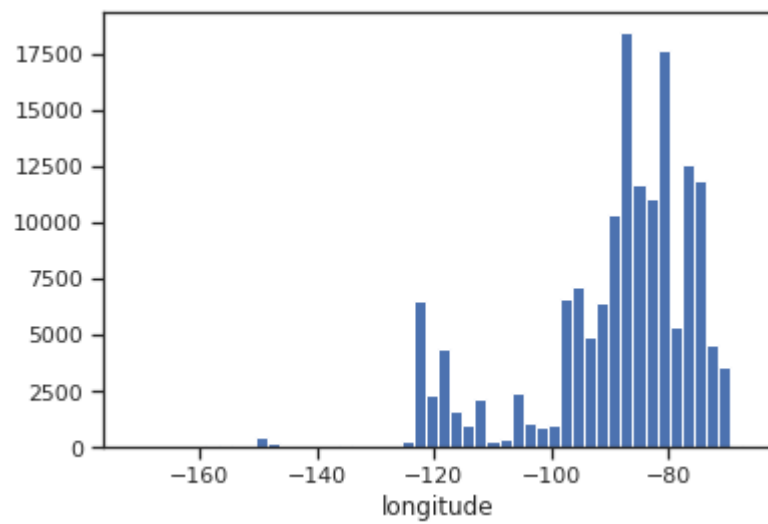
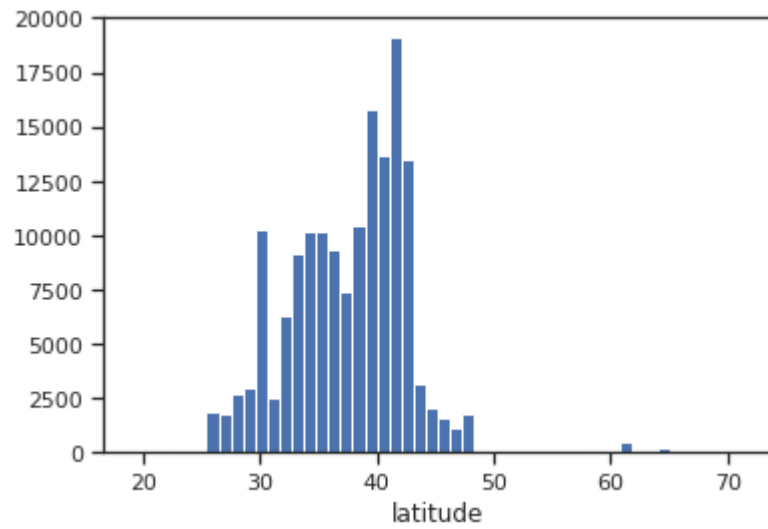
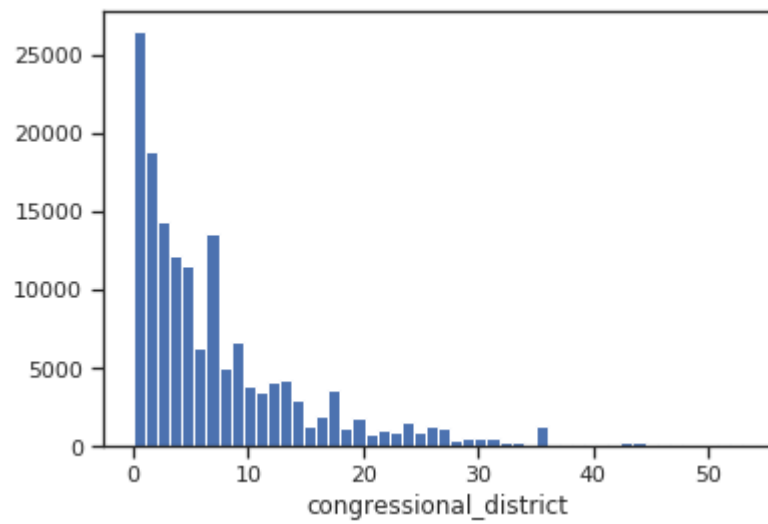
4	6.0	36.1140	-79.9569	□
↪2.0				
...	
162862	13.0	33.7938	-84.5894	□
↪1.0				
162863	13.0	37.7338	-122.1790	□
↪1.0				
162864	NaN	NaN	NaN	□
↪1.0				
162865	1.0	34.2190	-88.7378	□
↪1.0				
162866	8.0	35.0708	-89.6713	□
↪1.0				

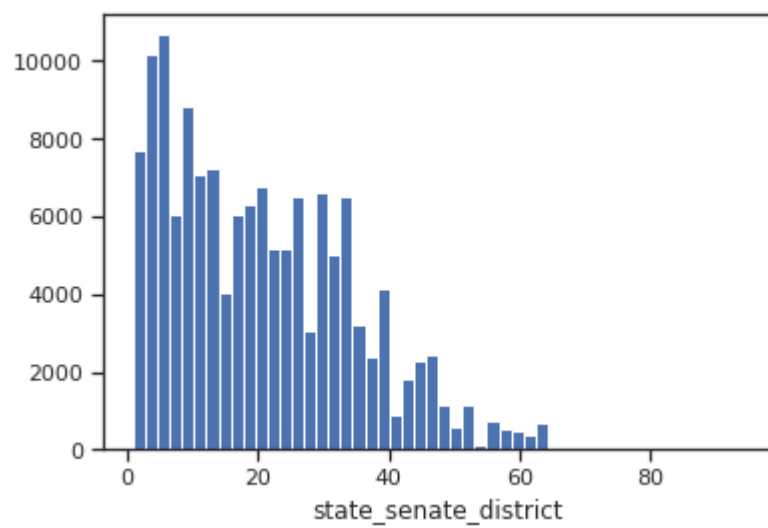
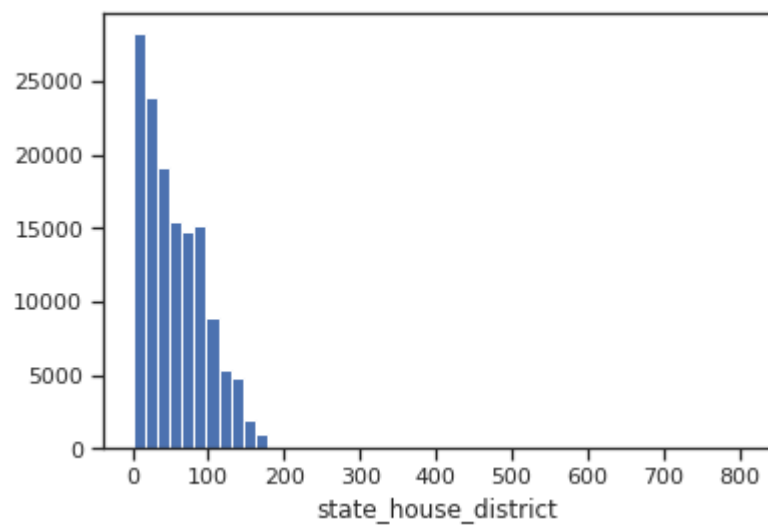
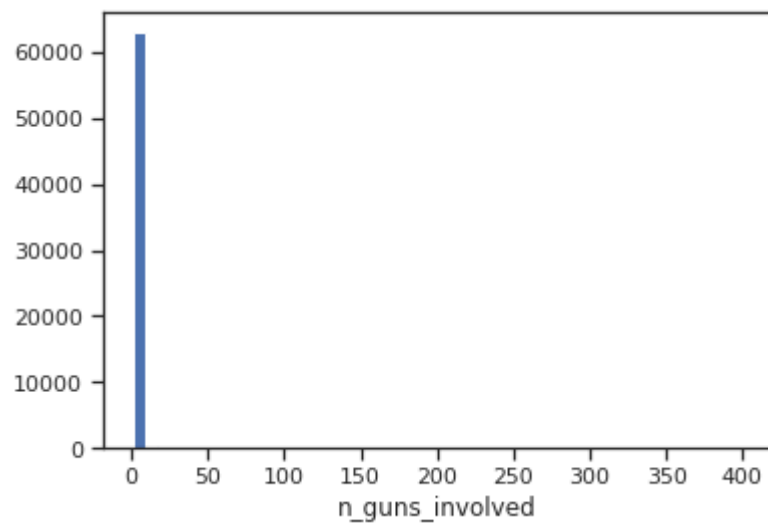
	state_house_district	state_senate_district
0	NaN	NaN
1	62.0	35.0
2	56.0	13.0
3	40.0	28.0
4	62.0	27.0
...
162862	39.0	38.0
162863	18.0	9.0
162864	NaN	NaN
162865	16.0	7.0
162866	95.0	32.0

[162867 rows x 6 columns]

```
[111]: # Гистограмма по признакам
for col in data_num:
    plt.hist(data[col], 50)
    plt.xlabel(col)
    plt.show()
```

```
/home/dan/anaconda3/lib/python3.7/site-packages/numpy/lib/
↪histograms.py:839:
RuntimeWarning: invalid value encountered in greater_equal
    keep = (tmp_a >= first_edge)
/home/dan/anaconda3/lib/python3.7/site-packages/numpy/lib/
↪histograms.py:840:
RuntimeWarning: invalid value encountered in less_equal
    keep &= (tmp_a <= last_edge)
```





```
[112]: # Фильтр по пустым значениям поля n_guns_involved
data[data['n_guns_involved'].isnull()]
```

```
[112]:
```

	incident_id	date	state	city_or_county	\
0	461105	2013-01-01	Pennsylvania	McKeesport	
1	460726	2013-01-01	California	Hawthorne	
3	478925	2013-01-05	Colorado	Aurora	
5	478948	2013-01-07	Oklahoma	Tulsa	
7	479374	2013-01-21	Louisiana	New Orleans	
...	
161927	729430	2016-12-16	Wisconsin	Madison	
162166	730843	2016-12-18	Florida	Naples	
162373	729453	2016-12-19	California	Brawley	
162740	730974	2016-12-21	Arkansas	Fayetteville	
162801	732054	2016-12-22	Florida	Palm Harbor	

	address	n_killed	\
0	1506 Versailles Avenue and Coursin Street	0	
1	13500 block of Cerise Avenue	1	
3	16000 block of East Ithaca Place	4	
5	6000 block of South Owasso	4	
7	LaSalle Street and Martin Luther King Jr. Boul...	0	
...	
161927	Fourth Street	0	
162166	Pine Ridge Rd and Airport Pulling Road	0	
162373	500 block of North Imperial	1	
162740	800 South School Avenue	1	
162801	252 Whisper Lake Road	1	

	n_injured	incident_url	\
0	4	http://www.gunviolencearchive.org/incident/461105	
1	3	http://www.gunviolencearchive.org/incident/460726	
3	0	http://www.gunviolencearchive.org/incident/478925	

5	0	http://www.gunviolencearchive.org/incident/
↪478948		
7	5	http://www.gunviolencearchive.org/incident/
↪479374		
...
161927	0	http://www.gunviolencearchive.org/incident/
↪729430		
162166	0	http://www.gunviolencearchive.org/incident/
↪730843		
162373	0	http://www.gunviolencearchive.org/incident/
↪729453		
162740	0	http://www.gunviolencearchive.org/incident/
↪730974		
162801	0	http://www.gunviolencearchive.org/incident/
↪732054		

	source_url \
0	http://www.post-gazette.com/local/south/2013/0...
1	http://www.dailybulletin.com/article/zz/201301...
3	http://www.dailydemocrat.com/20130106/aurora-s...
5	http://usnews.nbcnews.com/_news/2013/01/07/163...
7	http://www.nola.com/crime/index.ssf/2013/01/no...

...	...
161927	http://www.nbc15.com/content/news/2-teens-arre...
162166	http://www.naplesnews.com/story/news/crime/201...
162373	http://www.kyma.com/news/fatal-officer-involve...
162740	http://www.4029tv.com/article/officer-involved...
162801	http://web.tampabay.com/news/publicsafety/crim...

	incident_url_fields_missing
↪participant_age \			
0	False	...	0::20
1	False	...	0::20
3	False	...	0::29 1::33 2::56 3::33
5	False	...	0::23 1::23 2::33 3::55
7	False	...	NaN
...
161927	False	...	0::18 1::18
162166	False	...	0::24
162373	False	...	NaN
162740	False	...	0::25
162801	False	...	0::55

	participant_age_group \
0	0::Adult 18+ 1::Adult 18+ 2::Adult 18+ 3::A...
1	0::Adult 18+ 1::Adult 18+ 2::Adult 18+ 3::A...
3	0::Adult 18+ 1::Adult 18+ 2::Adult 18+ 3::A...
5	0::Adult 18+ 1::Adult 18+ 2::Adult 18+ 3::A...
7	NaN


```

...
161927      0::Adult 18+||1::Adult 18+
162166      0::Adult 18+
162373      0::Adult 18+
162740      0::Adult 18+
162801      0::Adult 18+

                                participant_gender \
0      0::Male||1::Male||3::Male||4::Female
1      0::Male
3      0::Female||1::Male||2::Male||3::Male
5      0::Female||1::Female||2::Female||3::Female||4:...
7      0::Male||1::Male||2::Male||3::Male||4::Male
...
161927      0::Male||1::Male
162166      0::Male
162373      0::Male
162740      0::Male
162801      0::Male

                                participant_name \
0      0::Julian Sims
1      0::Bernard Gillis
3      0::Stacie Philbrook||1::Christopher Ratliffe||...
5      0::Rebeika Powell||1::Kayetie Melchor||2::Mist...
7      NaN
...
161927      0::Taylor Loving||1::Theron Walker
162166      0::Sean Blackwell
162373      NaN
162740      0::Benjamin Ortiz
162801      0::Stanley Eversole

                                participant_relationship \
0      NaN
1      NaN
3      NaN
5      NaN
7      NaN
...
161927      NaN
162166      NaN
162373      NaN
162740      NaN
162801      NaN

                                participant_status \
0      0::Arrested||1::Injured||2::Injured||3::Injure...
1      0::Killed||1::Injured||2::Injured||3::Injured
3      0::Killed||1::Killed||2::Killed||3::Killed

```

5	0::Killed 1::Killed 2::Killed 3::Killed 4::...
7	0::Injured 1::Injured 2::Injured 3::Injured...
...	...
161927	0::Unharmed, Arrested 1::Unharmed, Arrested
162166	0::Unharmed, Arrested
162373	0::Killed
162740	0::Killed
162801	0::Killed
	participant_type \
0	0::Victim 1::Victim 2::Victim 3::Victim 4::...
1	0::Victim 1::Victim 2::Victim 3::Victim 4::...
3	0::Victim 1::Victim 2::Victim 3::Subject-Su...
5	0::Victim 1::Victim 2::Victim 3::Victim 4::...
7	0::Victim 1::Victim 2::Victim 3::Victim 4::...
...	...
161927	0::Subject-Suspect 1::Subject-Suspect
162166	0::Subject-Suspect
162373	0::Subject-Suspect
162740	0::Subject-Suspect
162801	0::Subject-Suspect
	sources \
0	http://pittsburgh.cbslocal.com/2013/01/01/4-pe...
1	http://losangeles.cbslocal.com/2013/01/01/man-...
3	http://denver.cbslocal.com/2013/01/06/officer-...
5	http://www.kjrh.com/news/local-news/4-found-sh...
7	http://www.huffingtonpost.com/2013/01/21/new-o...
...	...
161927	http://www.nbc15.com/content/news/2-teens-arre...
162166	http://www.naplesnews.com/story/news/crime/201...
162373	http://www.kyma.com/news/fatal-officer-involve...
162740	http://www.4029tv.com/article/officer-involved...
162801	http://www.nbcmiami.com/news/local/Deputies-Sh...
	state_house_district state_senate_district
0	NaN NaN
1	62.0 35.0
3	40.0 28.0
5	72.0 11.0
7	93.0 5.0
...	...
161927	76.0 26.0
162166	106.0 23.0
162373	56.0 40.0
162740	85.0 4.0
162801	NaN NaN

[99299 rows x 29 columns]

```
[113]: # Запоминаем индексы строк с пустыми значениями
flt_index = data[data['n_guns_involved'].isnull()].index
flt_index
```

```
[113]: Int64Index([      0,      1,      3,      5,      7,      8,
    ↪9,     14,
           17,     19,
           ...
    160630, 160803, 160878, 161236, 161836, 161927,
    ↪162166, 162373,
           162740, 162801],
           dtype='int64', length=99299)
```

```
[114]: # Проверяем что выводятся нужные строки
data[data.index.isin(flt_index)]
```

```
[114]:
```

	incident_id	date	state	city_or_county	\
0	461105	2013-01-01	Pennsylvania	McKeesport	
1	460726	2013-01-01	California	Hawthorne	
3	478925	2013-01-05	Colorado	Aurora	
5	478948	2013-01-07	Oklahoma	Tulsa	
7	479374	2013-01-21	Louisiana	New Orleans	
...	
161927	729430	2016-12-16	Wisconsin	Madison	
162166	730843	2016-12-18	Florida	Naples	
162373	729453	2016-12-19	California	Brawley	
162740	730974	2016-12-21	Arkansas	Fayetteville	
162801	732054	2016-12-22	Florida	Palm Harbor	

	n_killed	address
0		1506 Versailles Avenue and Coursin Street
↪0		
1		13500 block of Cerise Avenue
↪1		
3		16000 block of East Ithaca Place
↪4		
5		6000 block of South Owasso
↪4		
7		LaSalle Street and Martin Luther King Jr. Boul...
...		...
161927		Fourth Street
↪0		
162166		Pine Ridge Rd and Airport Pulling Road
↪0		
162373		500 block of North Imperial
↪1		
162740		800 South School Avenue
↪1		

162801

252 Whisper Lake Road

□

→1

	n_injured		
→incident_url \			□
0	4	http://www.gunviolencearchive.org/incident/461105	
1	3	http://www.gunviolencearchive.org/incident/460726	
3	0	http://www.gunviolencearchive.org/incident/478925	
5	0	http://www.gunviolencearchive.org/incident/478948	
7	5	http://www.gunviolencearchive.org/incident/479374	
...
161927	0	http://www.gunviolencearchive.org/incident/729430	
162166	0	http://www.gunviolencearchive.org/incident/730843	
162373	0	http://www.gunviolencearchive.org/incident/729453	
162740	0	http://www.gunviolencearchive.org/incident/730974	
162801	0	http://www.gunviolencearchive.org/incident/732054	

	source_url \
0	http://www.post-gazette.com/local/south/2013/01/07/1630106/aurora-south/
1	http://www.dailybulletin.com/article/zz/20130107/1630106/aurora-south/
3	http://www.dailydemocrat.com/20130106/aurora-south/
5	http://usnews.nbcnews.com/_news/2013/01/07/1630106/aurora-south/
7	http://www.nola.com/crime/index.ssf/2013/01/no...
...	...
161927	http://www.nbc15.com/content/news/2-teens-arrested-in-shooting-161927.html
162166	http://www.naplesnews.com/story/news/crime/2013-01-07/1630106/aurora-south/
162373	http://www.kyma.com/news/fatal-officer-involved-in-shooting-162373.html
162740	http://www.4029tv.com/article/officer-involved-in-shooting-162740.html
162801	http://web.tampabay.com/news/publicsafety/crime/162801.html

	incident_url_fields_missing		
→participant_age \			□
0	False	...	0::20
1	False	...	0::20
3	False	...	0::29 1::33 2::56 3::33
5	False	...	0::23 1::23 2::33 3::55
7	False	...	NaN
...
161927	False	...	0::18 1::18

162166	False	...	0::24
162373	False	...	NaN
162740	False	...	0::25
162801	False	...	0::55

	participant_age_group \
0	0::Adult 18+ 1::Adult 18+ 2::Adult 18+ 3::A...
1	0::Adult 18+ 1::Adult 18+ 2::Adult 18+ 3::A...
3	0::Adult 18+ 1::Adult 18+ 2::Adult 18+ 3::A...
5	0::Adult 18+ 1::Adult 18+ 2::Adult 18+ 3::A...
7	NaN

...	...
161927	0::Adult 18+ 1::Adult 18+
162166	0::Adult 18+
162373	0::Adult 18+
162740	0::Adult 18+
162801	0::Adult 18+

	participant_gender \
0	0::Male 1::Male 3::Male 4::Female
1	0::Male
3	0::Female 1::Male 2::Male 3::Male
5	0::Female 1::Female 2::Female 3::Female 4::...
7	0::Male 1::Male 2::Male 3::Male 4::Male

...	...
161927	0::Male 1::Male
162166	0::Male
162373	0::Male
162740	0::Male
162801	0::Male

	participant_name \
0	0::Julian Sims
1	0::Bernard Gillis
3	0::Stacie Philbrook 1::Christopher Ratliffe ...
5	0::Rebeika Powell 1::Kayetie Melchor 2::Mist...
7	NaN

...	...
161927	0::Taylor Loving 1::Theron Walker
162166	0::Sean Blackwell
162373	NaN
162740	0::Benjamin Ortiz
162801	0::Stanley Eversole

	participant_relationship \
0	NaN
1	NaN
3	NaN
5	NaN
7	NaN

...	...
161927	NaN
162166	NaN
162373	NaN
162740	NaN
162801	NaN

	participant_status \
0	0::Arrested 1::Injured 2::Injured 3::Injure...
1	0::Killed 1::Injured 2::Injured 3::Injured
3	0::Killed 1::Killed 2::Killed 3::Killed
5	0::Killed 1::Killed 2::Killed 3::Killed 4::...
7	0::Injured 1::Injured 2::Injured 3::Injured...
...	...
161927	0::Unharmed, Arrested 1::Unharmed, Arrested
162166	0::Unharmed, Arrested
162373	0::Killed
162740	0::Killed
162801	0::Killed

	participant_type \
0	0::Victim 1::Victim 2::Victim 3::Victim 4::...
1	0::Victim 1::Victim 2::Victim 3::Victim 4::...
3	0::Victim 1::Victim 2::Victim 3::Subject-Su...
5	0::Victim 1::Victim 2::Victim 3::Victim 4::...
7	0::Victim 1::Victim 2::Victim 3::Victim 4::...
...	...
161927	0::Subject-Suspect 1::Subject-Suspect
162166	0::Subject-Suspect
162373	0::Subject-Suspect
162740	0::Subject-Suspect
162801	0::Subject-Suspect

	sources \
0	http://pittsburgh.cbslocal.com/2013/01/01/4-pe...
1	http://losangeles.cbslocal.com/2013/01/01/man-...
3	http://denver.cbslocal.com/2013/01/06/officer-...
5	http://www.kjrh.com/news/local-news/4-found-sh...
7	http://www.huffingtonpost.com/2013/01/21/new-o...
...	...
161927	http://www.nbc15.com/content/news/2-teens-arre...
162166	http://www.naplesnews.com/story/news/crime/201...
162373	http://www.kyma.com/news/fatal-officer-involve...
162740	http://www.4029tv.com/article/officer-involved...
162801	http://www.nbcmiami.com/news/local/Deputies-Sh...

	state_house_district	state_senate_district
0	NaN	NaN
1	62.0	35.0
3	40.0	28.0

5	72.0	11.0
7	93.0	5.0
...
161927	76.0	26.0
162166	106.0	23.0
162373	56.0	40.0
162740	85.0	4.0
162801	NaN	NaN

[99299 rows x 29 columns]

```
[115]: # фильтр по колонке
data_num[data_num.index.isin(flt_index)]['n_guns_involved']
```

```
[115]: 0      NaN
1      NaN
3      NaN
5      NaN
7      NaN
      ..
161927 NaN
162166 NaN
162373 NaN
162740 NaN
162801 NaN
Name: n_guns_involved, Length: 99299, dtype: float64
```

Будем использовать встроенные средства импутации библиотеки scikit-learn - <https://scikit-learn.org/stable/modules/impute.html#impute>

```
[116]: data_num_guns = data_num[['n_guns_involved']]
data_num_guns.head()
```

```
[116]: n_guns_involved
0      NaN
1      NaN
2      2.0
3      NaN
4      2.0
```

```
[117]: from sklearn.impute import SimpleImputer
from sklearn.impute import MissingIndicator
```

```
[118]: # Фильтр для проверки заполнения пустых значений
indicator = MissingIndicator()
mask_missing_values_only = indicator.fit_transform(data_num_guns)
mask_missing_values_only
```

```
[118]: array([[ True],
       [ True],
       [False],
       ...,
```

```
[False],  
[False],  
[False]])
```

С помощью класса SimpleImputer можно проводить импьютацию различными показателями центра распределения

```
[119]: strategies=['mean', 'median', 'most_frequent']
```

```
[120]: def test_num_impute(strategy_param):  
        imp_num = SimpleImputer(strategy=strategy_param)  
        data_num_imp = imp_num.fit_transform(data_num_guns)  
        return data_num_imp[mask_missing_values_only]
```

```
[121]: strategies[0], test_num_impute(strategies[0])
```

```
[121]: ('mean',  
        array([1.5237069, 1.5237069, 1.5237069, ..., 1.5237069, 1.5237069,  
               1.5237069]))
```

```
[122]: strategies[1], test_num_impute(strategies[1])
```

```
[122]: ('median', array([1., 1., 1., ..., 1., 1., 1.]))
```

```
[123]: strategies[2], test_num_impute(strategies[2])
```

```
[123]: ('most_frequent', array([1., 1., 1., ..., 1., 1., 1.]))
```

```
[124]: # Более сложная функция, которая позволяет задавать колонку и вид  
        ↪ импьютации  
def test_num_impute_col(dataset, column, strategy_param):  
    temp_data = dataset[[column]]  
  
    indicator = MissingIndicator()  
    mask_missing_values_only = indicator.fit_transform(temp_data)  
  
    imp_num = SimpleImputer(strategy=strategy_param)  
    data_num_imp = imp_num.fit_transform(temp_data)  
  
    filled_data = data_num_imp[mask_missing_values_only]  
  
    return column, strategy_param, filled_data.size,   
    ↪ filled_data[0], filled_data[filled_data.size-1]
```

```
[125]: test_num_impute_col(data, 'n_guns_involved', strategies[0])
```

```
[125]: ('n_guns_involved', 'mean', 99299, 1.5237068965517242, 1.  
        ↪ 5237068965517242)
```

```
[126]: test_num_impute_col(data, 'n_guns_involved', strategies[1])
```

```
[126]: ('n_guns_involved', 'median', 99299, 1.0, 1.0)
```



```
[127]: test_num_impute_col(data, 'n_guns_involved', strategies[2])
```

```
[127]: ('n_guns_involved', 'most_frequent', 99299, 1.0, 1.0)
```

2.2.2. 1.2.2. Обработка пропусков в категориальных данных

```
[128]: # Выберем категориальные колонки с пропущенными значениями
# Цикл по колонкам датасета
cat_cols = []
for col in data.columns:
    # Количество пустых значений
    temp_null_count = data[data[col].isnull()].shape[0]
    dt = str(data[col].dtype)
    if temp_null_count > 0 and (dt == 'object'):
        cat_cols.append(col)
        temp_perc = round((temp_null_count / total_count) * 100,
        ↪ 0, 2)
        print('Колонка {}. Тип данных {}. Количество пустых
        ↪ значений {}, {}%.'.format(col, dt, temp_null_count, temp_perc))
```

Колонка address. Тип данных object. Количество пустых значений
↪ 12303, 7.55%.

Колонка source_url. Тип данных object. Количество пустых значений
↪ 276, 0.17%.

Колонка gun_stolen. Тип данных object. Количество пустых значений
↪ 99311, 60.98%.

Колонка gun_type. Тип данных object. Количество пустых значений
↪ 99299, 60.97%.

Колонка incident_characteristics. Тип данных object. Количество
↪ пустых значений
242, 0.15%.

Колонка location_description. Тип данных object. Количество пустых
↪ значений
140476, 86.25%.

Колонка notes. Тип данных object. Количество пустых значений
↪ 56008, 34.39%.

Колонка participant_age. Тип данных object. Количество пустых
↪ значений 63464,
38.97%.

Колонка participant_age_group. Тип данных object. Количество
↪ пустых значений
27678, 16.99%.

Колонка participant_gender. Тип данных object. Количество пустых
↪ значений 23832,
14.63%.

Колонка participant_name. Тип данных object. Количество пустых
↪ значений 84207,
51.7%.

Колонка participant_relationship. Тип данных object. Количество
 ↳пустых значений
 152618, 93.71%.
 Колонка participant_status. Тип данных object. Количество пустых
 ↳значений 18510,
 11.37%.
 Колонка participant_type. Тип данных object. Количество пустых
 ↳значений 16327,
 10.02%.
 Колонка sources. Тип данных object. Количество пустых значений
 ↳516, 0.32%.

Класс SimpleImputer можно использовать для категориальных признаков со стратегиями “most_frequent” или “constant”.

```
[129]: cat_temp_data = data[['gun_stolen']]
       cat_temp_data.head()
```

```
[129]:          gun_stolen
0                NaN
1                NaN
2  0::Unknown|1::Unknown
3                NaN
4  0::Unknown|1::Unknown
```

```
[130]: cat_temp_data['gun_stolen'].unique()[0:10]
```

```
[130]: array([nan, '0::Unknown|1::Unknown', '0::Unknown',
              '0::Unknown|1::Unknown|2::Unknown|3::Unknown',
              '0::Not-stolen|1::Unknown', '0::Unknown|1::Unknown|2::
              ↳Unknown',
              '0::Stolen|1::Stolen', '0::Not-stolen', '0::Stolen',
              '0::Stolen|1::Stolen|2::Unknown|3::Unknown'],
              ↳dtype=object)
```

```
[131]: cat_temp_data[cat_temp_data['gun_stolen'].isnull()].shape
```

```
[131]: (99311, 1)
```

```
[132]: # Импутация наиболее частыми значениями
       imp2 = SimpleImputer(missing_values=np.nan,
       ↳strategy='most_frequent')
       data_imp2 = imp2.fit_transform(cat_temp_data)
       data_imp2
```

```
[132]: array(['0::Unknown',
              '0::Unknown',
              '0::Unknown|1::Unknown',
              ...,
              '0::Unknown',
              '0::Unknown',
              '0::Unknown'], dtype=object)
```

```
[133]: # Пустые значения отсутствуют
np.unique(data_imp2)[0:5]
```

```
[133]: array(['0::Not-stolen', '0::Not-stolen|1::Not-stolen',
          '0::Not-stolen|1::Not-stolen|2::Not-stolen',
          '0::Not-stolen|1::Not-stolen|2::Not-stolen|3::
          ↪Not-stolen',
          '0::Not-stolen|1::Not-stolen|2::Not-stolen|3::
          ↪Not-stolen|4::Not-
          stolen'],
          dtype=object)
```

```
[134]: # Импутация константой
imp3 = SimpleImputer(missing_values=np.nan, strategy='constant',
          ↪fill_value='!!!')
data_imp3 = imp3.fit_transform(cat_temp_data)
data_imp3
```

```
[134]: array(['!!!',
          '!!!',
          '0::Unknown|1::Unknown'],
          ...,
          ['0::Unknown'],
          ['0::Unknown'],
          ['0::Unknown']], dtype=object)
```

```
[135]: np.unique(data_imp3)[0:5]
```

```
[135]: array(['!!!', '0::Not-stolen', '0::Not-stolen|1::Not-stolen',
          '0::Not-stolen|1::Not-stolen|2::Not-stolen',
          '0::Not-stolen|1::Not-stolen|2::Not-stolen|3::
          ↪Not-stolen'],
          dtype=object)
```

```
[136]: data_imp3[data_imp3=='!!!'].size
```

```
[136]: 99311
```

3. 2. Преобразование категориальных признаков в числовые

```
[137]: cat_enc = pd.DataFrame({'c1':data_imp2.T[0]})
cat_enc
```

```
[137]:
```

	c1
0	0::Unknown
1	0::Unknown
2	0::Unknown 1::Unknown
3	0::Unknown
4	0::Unknown 1::Unknown

```
...
162862          0::Unknown
162863          0::Unknown
162864          0::Unknown
162865          0::Unknown
162866          0::Unknown
```

```
[162867 rows x 1 columns]
```

3.1. 2.1. Кодирование категорий целочисленными значениями - label encoding

```
[138]: from sklearn.preprocessing import LabelEncoder, OneHotEncoder
```

```
[139]: le = LabelEncoder()
cat_enc_le = le.fit_transform(cat_enc['c1'])
```

```
[140]: cat_enc['c1'].unique()[0:5]
```

```
[140]: array(['0::Unknown', '0::Unknown|1::Unknown',
              '0::Unknown|1::Unknown|2::Unknown|3::Unknown',
              '0::Not-stolen|1::Unknown', '0::Unknown|1::Unknown|2::
→Unknown'],
              dtype=object)
```

```
[141]: np.unique(cat_enc_le)[0:10]
```

```
[141]: array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
```

```
[142]: le.inverse_transform([0, 1, 2, 3])
```

```
[142]: array(['0::Not-stolen', '0::Not-stolen|1::Not-stolen',
              '0::Not-stolen|1::Not-stolen|2::Not-stolen',
              '0::Not-stolen|1::Not-stolen|2::Not-stolen|3::
→Not-stolen'],
              dtype=object)
```

3.2. 2.2. Кодирование категорий наборами бинарных значений - one-hot encoding

```
[143]: ohe = OneHotEncoder()
cat_enc_ohe = ohe.fit_transform(cat_enc[['c1']])
```

```
[144]: cat_enc.shape
```

```
[144]: (162867, 1)
```

```
[145]: cat_enc_ohe.shape
```

```
[145]: (162867, 277)
```

```
[146]: cat_enc_ohe
```

```
[146]: <162867x277 sparse matrix of type '<class 'numpy.float64'>'
      with 162867 stored elements in Compressed Sparse Row
      ↪format>
```

```
[147]: cat_enc_ohe.todense()[0:10]
```

```
[147]: matrix([[0., 0., 0., ..., 0., 0., 0.],
             [0., 0., 0., ..., 0., 0., 0.],
             [0., 0., 0., ..., 0., 0., 0.],
             ...,
             [0., 0., 0., ..., 0., 0., 0.],
             [0., 0., 0., ..., 0., 0., 0.],
             [0., 0., 0., ..., 0., 0., 0.]])
```

```
[148]: cat_enc.head(10)
```

```
[148]:
```

	c1
0	0::Unknown
1	0::Unknown
2	0::Unknown 1::Unknown
3	0::Unknown
4	0::Unknown 1::Unknown
5	0::Unknown
6	0::Unknown 1::Unknown
7	0::Unknown
8	0::Unknown
9	0::Unknown

4. 3. Масштабирование данных

Термины “масштабирование” и “нормализация” часто используются как синонимы. Масштабирование предполагает изменение диапазона измерения величины, а нормализация - изменение распределения этой величины.

Если признаки лежат в различных диапазонах, то необходимо их нормализовать. Как правило, применяют два подхода: - MinMax масштабирование:

$$x = \frac{x - \min(X)}{\max(X) - \min(X)}$$

В этом случае значения лежат в диапазоне от 0 до 1. - Масштабирование данных на основе Z-оценки:

$$x = \frac{x - AVG(X)}{\sigma(X)}$$

В этом случае большинство значений попадает в диапазон от -3 до 3.

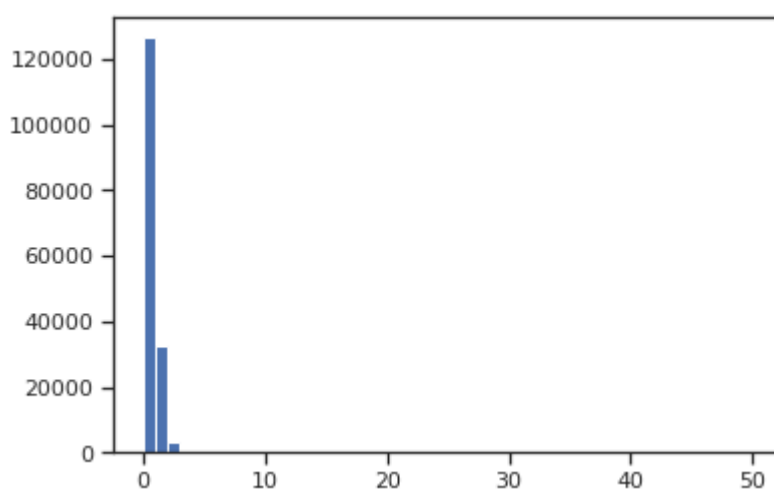
где X - матрица объект-признак, $AVG(X)$ - среднее значение, σ - среднеквадратичное отклонение.

```
[149]: from sklearn.preprocessing import MinMaxScaler, StandardScaler, Normalizer
```

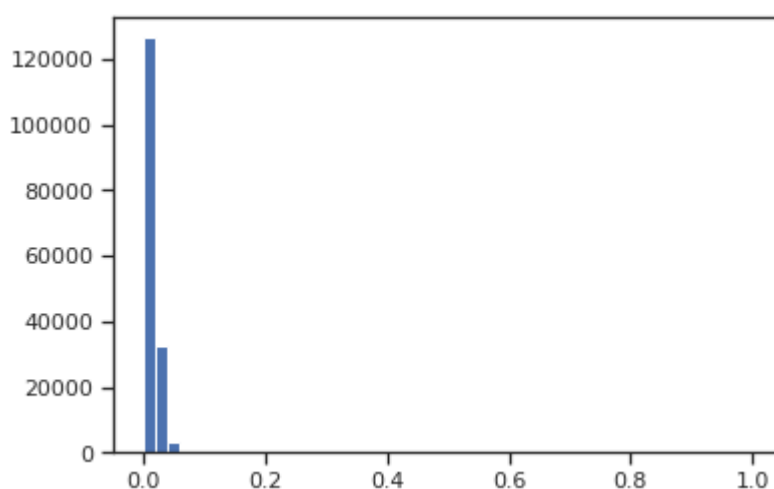
4.1. 3.1. MinMax масштабирование

```
[150]: sc1 = MinMaxScaler()  
sc1_data = sc1.fit_transform(data[['n_killed']])
```

```
[151]: plt.hist(data['n_killed'], 50)  
plt.show()
```



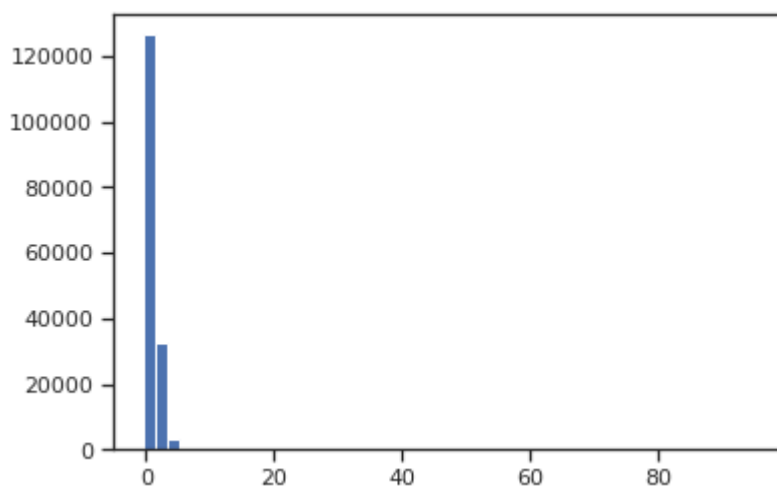
```
[152]: plt.hist(sc1_data, 50)  
plt.show()
```



4.2. 3.2. Масштабирование данных на основе Z-оценки - StandardScaler

```
[153]: sc2 = StandardScaler()  
sc2_data = sc2.fit_transform(data[['n_killed']])
```

```
[154]: plt.hist(sc2_data, 50)  
plt.show()
```



4.3. 3.3. Нормализация данных

```
[155]: sc3 = Normalizer()  
sc3_data = sc3.fit_transform(data[['n_killed']])
```

```
[156]: plt.hist(sc3_data, 50)  
plt.show()
```

