# In [1]: 1 import pandas as pd 2 import numpy as np 3 from matplotlib import pyplot as plt 4 %matplotlib inline 5 import seaborn as sns 6 sns.set() 7 import random

2 data= pd.read\_csv("C:\\Users\\ADMIN\\Desktop\\LETSUPGRADE\\DATA SCIENCE\\NOT

3 print(data)

	I	Id g	roupId	ma	tchId	assist	s	boosts	\		
0	2f262dd9795e6	_	•	 5db3a49e			0	0	`		
1	a32847cf5bf34			5223f05c			0	0			
2	1b1900a999039	96 edf80d65	23380a 1	cadec453	4 <b>f</b> 30a		0	3			
3	f589dd03b60bf	2 804ab5e5	585558 c4	4a5676dc	91604		0	0			
4	c23c4cc5b78b3		9ed920 co	d595700a	01bfa		0	0			
		•									
9995	ef4f474acd8e8	35 2eca2a83	91f75d 49	92ecdfae	90b46		0	3			
9996	cf0bf82fb4d80	e 2eaf2765	f93adb 14	4bffd71e	96320		0	0			
9997	a0a31a0b1dcbe	e1 8d50c64c	cc5071 14	47e4bbb6	2e3bb		0	0			
9998	f6874657399d6	9 d31843d7	e62ccb 6	52567dcf	280f5		0	0			
9999	90359b0b8f8b0	d 61d5b1bb	8da43f 2	58bfa48d	88014		0	0			
	•		shotKills	heals	killP:			roadKi	11s	\	
0	0.0	0	0	0			• •		0		
1	163.2	1	1	0			• •		0		
2	278.7	2	1	8			• •		0		
3	191.9	1	0	0			• •		0		
4	100.0	1	0	0		87 .	• •		0		
• • •	• • •	• • •	• • •	• • •			• •		• • •		
9995	204.5	1	0	0			• •		0		
9996	0.0	0	0	0			• •		0		
9997	0.0	0	0	0			• •		0		
9998	0.0	0	0	0			• •		0		
9999	0.0	0	0	0		78 .	• •		0		
	swimDistance	teamKills	vehicleD	astrovs	walkD:	istance	WE	anonsAc	auir	Δd	\
а	swimDistance a a	teamKills	vehicleDe	-	walkD			aponsAc	quir		\
0 1	0.0	0	vehicleDe	0	walkD	0.00		aponsAc	quir	0	\
1	0.0 0.0	0 0	vehicleDe	0 0		0.00 132.70		eaponsAc		0 2	\
1 2	0.0 0.0 0.0	0 0 0	vehicleDe	0 0 0		0.00 132.70 3591.00		aponsAc		0 2 10	\
1 2 3	0.0 0.0 0.0 0.0	0 0 0 0	vehicleDe	0 0 0 0		0.00 132.70 3591.00 332.70		aponsAc		0 2 10 3	\
1 2 3 4	0.0 0.0 0.0 0.0	0 0 0	vehicleDe	0 0 0		0.00 132.70 3591.00 332.70 252.70		eaponsAc		0 2 10	\
1 2 3 4	0.0 0.0 0.0 0.0	0 0 0 0	vehicleDe	0 0 0 0	3	0.00 132.70 3591.00 332.70 252.70		eaponsAc		0 2 10 3	\
1 2 3 4	0.0 0.0 0.0 0.0 0.0	0 0 0 0	vehicleDe	0 0 0 0	3	0.00 132.70 3591.00 332.70 252.70		eaponsAc		0 2 10 3 3	\
1 2 3 4  9995	0.0 0.0 0.0 0.0 0.0	0 0 0 0 	vehicleDe	0 0 0 0 0	3	0.00 132.70 3591.00 332.70 252.70 		eaponsAc		0 2 10 3 3  4 6	\
1 2 3 4  9995 9996	0.0 0.0 0.0 0.0 0.0  0.0	0 0 0 0 	vehicleDe	0 0 0 0 0	3	0.00 132.70 3591.00 332.70 252.70  1648.00 897.10		eaponsAc		0 2 10 3 3	\
1 2 3 4  9995 9996 9997	0.0 0.0 0.0 0.0 0.0 0.0	0 0 0 0  0	vehicleDe	0 0 0 0 	3	0.00 132.70 3591.00 332.70 252.70  1648.00 897.10 188.20		eaponsAc		0 2 10 3 3  4 6 2	\
1 2 3 4  9995 9996 9997 9998	0.0 0.0 0.0 0.0  0.0 0.0	0 0 0 0  0	vehicleDe	0 0 0 0 0 0	3	0.00 132.70 3591.00 332.70 252.70  1648.00 897.10 188.20 108.10		eaponsAc		0 2 10 3 3  4 6 2 5	\
1 2 3 4  9995 9996 9997 9998	0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0	0 0 0 0  0	vehicleDe	0 0 0 0 0  0 0	: :	0.00 132.70 3591.00 332.70 252.70  1648.00 897.10 188.20 108.10		eaponsAc		0 2 10 3 3  4 6 2 5	\
1 2 3 4  9995 9996 9997 9998	0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0	0 0 0 0  0 0	Unnamed:	0 0 0 0 0  0	: :	0.00 132.70 3591.00 332.70 252.70  1648.00 897.10 188.20 108.10		eaponsAc		0 2 10 3 3  4 6 2 5	\
1 2 3 4  9995 9996 9997 9998 9999	0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 winPoints wi	0 0 0 0  0 0 0 0	Unnamed:	0 0 0 0  0 0 0	S	0.00 132.70 3591.00 332.70 252.70  1648.00 897.10 188.20 108.10		eaponsAc		0 2 10 3 3  4 6 2 5	\
1 2 3 4  9995 9996 9997 9998 9999	0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 winPoints wi	0 0 0 0  0 0 0 0	Unnamed:	0 0 0 0  0 0 0 29 kil	s 0	0.00 132.70 3591.00 332.70 252.70  1648.00 897.10 188.20 108.10		eaponsAc		0 2 10 3 3  4 6 2 5	\
1 2 3 4  9995 9996 9997 9998 9999	0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 winPoints wi 1470 1531 0	0 0 0 0 0  0 0 0 1nPlacePerc 0.0000 0.2222 0.8571 0.3462	Unnamed:	0 0 0 0 0 0 0 0 29 kil NaN NaN	s 0 1	0.00 132.70 3591.00 332.70 252.70  1648.00 897.10 188.20 108.10		eaponsAc		0 2 10 3 3  4 6 2 5	
1 2 3 4  9995 9996 9997 9998 9999	0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 winPoints wi 1470 1531 0	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	Unnamed:	0 0 0 0 0 0 0 0 29 kil	s 0 1	0.00 132.70 3591.00 332.70 252.70  1648.00 897.10 188.20 108.10		eaponsAc		0 2 10 3 3  4 6 2 5	
1 2 3 4  9995 9996 9997 9998 9999 0 1 2 3 4 	0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 winPoints wi 1470 1531 0 0	0 0 0 0 0 0 0 0 0 1nPlacePerc 0.0000 0.2222 0.8571 0.3462 0.0690	Unnamed:	0 0 0 0 0 0 0 0 0 29 kil NaN NaN NaN	s 0 1 1 0 0	0.00 132.70 3591.00 332.70 252.70  1648.00 897.10 188.20 108.10		eaponsAc		0 2 10 3 3  4 6 2 5	
1 2 3 4  9995 9996 9997 9998 9999 0 1 2 3 4 	0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 winPoints wi 1470 1531 0 0 1557 	0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0	Unnamed:	0 0 0 0 0 0 0 0 0 29 kil NaN NaN NaN NaN	s 0 1 1 0 0	0.00 132.70 3591.00 332.70 252.70  1648.00 897.10 188.20 108.10		eaponsAc		0 2 10 3 3  4 6 2 5	
1 2 3 4  9995 9996 9997 9998 9999 0 1 2 3 4  9995 9996	0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 winPoints wi 1470 1531 0 0 1557  1471 1500	0 0 0 0 0 0 0 0 0 0 1nPlacePerc 0.0000 0.2222 0.8571 0.3462 0.0690  0.8333 0.7174	Unnamed:	0 0 0 0 0 0 0 0 0 29 kill NaN NaN NaN NaN NaN	s 0 1 1 0 0	0.00 132.70 3591.00 332.70 252.70  1648.00 897.10 188.20 108.10		eaponsAc		0 2 10 3 3  4 6 2 5	
1 2 3 4  9995 9996 9997 0 1 2 3 4  9995 9996 9997	0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 winPoints wi 1470 1531 0 0 1557  1471 1500 1434	0 0 0 0 0 0 0 0 0 0 1nPlacePerc 0.0000 0.2222 0.8571 0.3462 0.0690  0.8333 0.7174 0.2083	Unnamed:	0 0 0 0 0 0 0 0 0 0 29 kil NaN NaN NaN NaN NaN	s 0 1 1 0 0 0	0.00 132.70 3591.00 332.70 252.70  1648.00 897.10 188.20 108.10		eaponsAc		0 2 10 3 3  4 6 2 5	
1 2 3 4  9995 9996 9997 9998 9999 0 1 2 3 4  9995 9996	0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 winPoints wi 1470 1531 0 0 1557  1471 1500	0 0 0 0 0 0 0 0 0 0 1nPlacePerc 0.0000 0.2222 0.8571 0.3462 0.0690  0.8333 0.7174	Unnamed:	0 0 0 0 0 0 0 0 0 29 kill NaN NaN NaN NaN NaN	s 0 1 1 0 0	0.00 132.70 3591.00 332.70 252.70  1648.00 897.10 188.20 108.10		eaponsAc		0 2 10 3 3  4 6 2 5	

<class 'pandas.core.frame.DataFrame'> RangeIndex: 10000 entries, 0 to 9999 Data columns (total 29 columns): Ιd 10000 non-null object groupId 10000 non-null object 10000 non-null object matchId assists 10000 non-null int64 10000 non-null int64 boosts damageDealt 10000 non-null float64 **DBNOs** 10000 non-null int64 headshotKills 10000 non-null int64 heals 10000 non-null int64 killPlace 10000 non-null int64 killPoints 10000 non-null int64 kills 10000 non-null int64 10000 non-null int64 killStreaks longestKill 10000 non-null float64 matchDuration 10000 non-null int64 matchType 10000 non-null object 10000 non-null int64 maxPlace 10000 non-null int64 numGroups 10000 non-null int64 rankPoints revives 10000 non-null int64 rideDistance 10000 non-null float64 roadKills 10000 non-null int64 10000 non-null float64 swimDistance teamKills 10000 non-null int64 10000 non-null int64 vehicleDestroys walkDistance 10000 non-null float64

dtypes: float64(6), int64(19), object(4)

10000 non-null int64

10000 non-null int64

10000 non-null float64

memory usage: 2.2+ MB

weaponsAcquired

winPoints

winPlacePerc

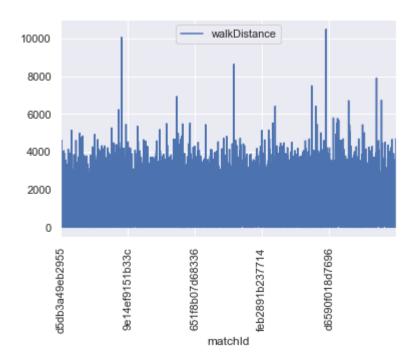
### Out[4]:

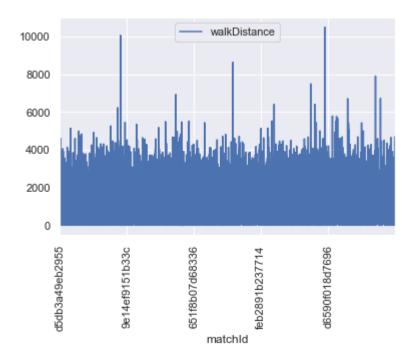
	assists	boosts	damageDealt	DBNOs	headshotKills	heals	killPlace	killPoints	kills	killSt
0	0	0	0.0	0	0	0	92	1126	0	
1	0	0	163.2	1	1	0	42	1309	1	
2	0	3	278.7	2	1	8	16	0	2	
3	0	0	191.9	1	0	0	31	0	1	
4	0	0	100.0	1	0	0	87	1332	0	
9995	0	3	204.5	1	0	0	17	1033	2	
9996	0	0	0.0	0	0	0	49	1000	0	
9997	0	0	0.0	0	0	0	83	1334	0	
9998	0	0	0.0	0	0	0	82	1335	0	
9999	0	0	0.0	0	0	0	78	0	0	

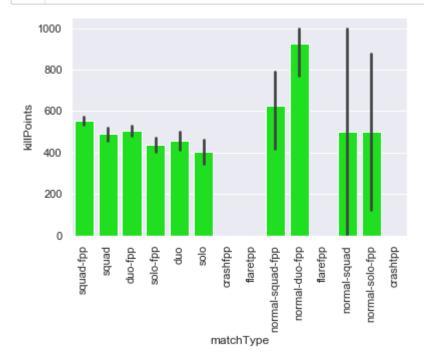
10000 rows × 25 columns

```
In [7]:
          1 # Question 4:The average person kills how many players?
          2 import pandas as pd
          3 from matplotlib import pyplot as plt
          4 import seaborn as sns
          5 sns.set_style("darkgrid")
          6 sns.set(color_codes=True)
             df = pd.read_csv("C:\\Users\\ADMIN\\Desktop\\LETSUPGRADE\\DATA SCIENCE\\NOTE
          7
            print(df["kills"])
          9
                0
        0
        1
                1
                2
        2
        3
                1
        4
                0
        9995
                2
        9996
                0
        9997
                0
        9998
                0
        9999
        Name: kills, Length: 10000, dtype: int64
```

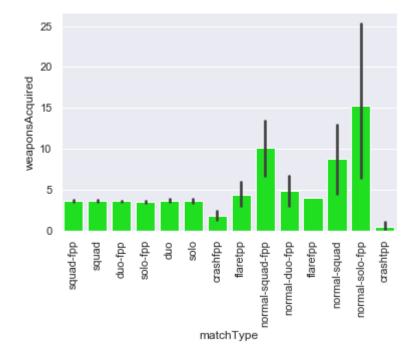
#### 

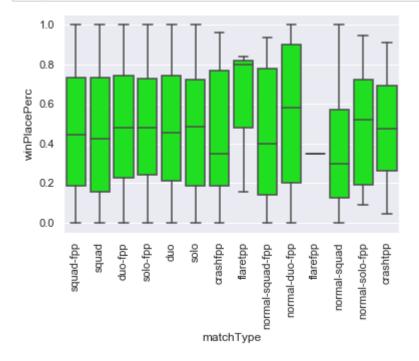




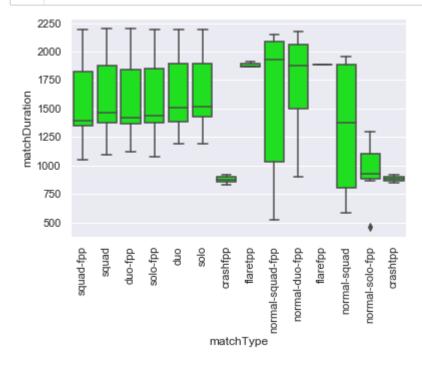


In [20]: 1 #Question 15. Plot a barplot of 'matchType' vs 'weaponsAcquired'. Write your
2 import seaborn as sns
3 from matplotlib import pyplot as plt
4 import pandas as pd
5 df = pd.read\_csv("C:\\Users\\ADMIN\\Desktop\\LETSUPGRADE\\DATA SCIENCE\\NOTE
6 sns.barplot(x="matchType", y="weaponsAcquired", data=df,color="lime")
7 plt.xticks(rotation=90)
8 plt.show()





```
In [22]: 1 #Question 18. Plot a boxplot of 'matchType' vs 'matchDuration'. Write your i
    import seaborn as sns
    from matplotlib import pyplot as plt
    import pandas as pd
    df = pd.read_csv("C:\\Users\\ADMIN\\Desktop\\LETSUPGRADE\\DATA SCIENCE\\NOTE
        sns.boxplot(x="matchType", y="matchDuration", data=df,color="lime")
        plt.xticks(rotation=90)
        plt.show()
```

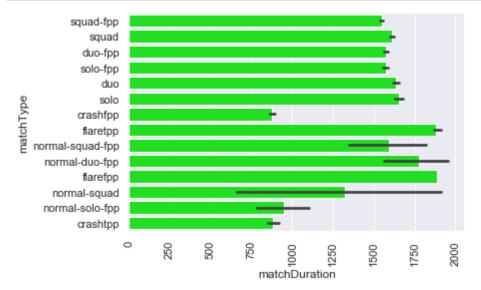


```
In [23]:
           1 #Question16. Find the Categorical columns.
           2 list(data.columns.values)
Out[23]: ['Id',
           'groupId',
           'matchId',
           'assists',
           'boosts',
           'damageDealt',
           'DBNOs',
           'headshotKills',
           'heals',
           'killPlace',
           'killPoints',
           'kills',
           'killStreaks',
           'longestKill',
           'matchDuration',
           'matchType',
           'maxPlace',
           'numGroups',
           'rankPoints',
           'revives',
           'rideDistance',
           'roadKills',
           'swimDistance',
           'teamKills',
```

'vehicleDestroys',
'walkDistance',
'weaponsAcquired',

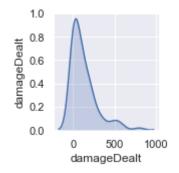
'winPoints',
'winPlacePerc']

```
In [39]: 1 #Question 19. Change the orientation of the above plot to horizontal.
2 import seaborn as sns
3 from matplotlib import pyplot as plt
4 import pandas as pd
5 df = pd.read_csv("C:\\Users\\ADMIN\\Desktop\\LETSUPGRADE\\DATA SCIENCE\\NOTE
6 sns.barplot(x="matchDuration",y="matchType",data=df,color="lime")
7 plt.xticks(rotation=90)
8 plt.show()
```



```
In [43]: 1 #Question 22. Take a sample of size 50 from the column damageDealt for 100
2 col=[ "damageDealt"]
3 sns.pairplot(data[col].sample(100), size =2.5, kind="reg", diag_kind="kde")
```

Out[43]: <seaborn.axisgrid.PairGrid at 0xb83be33248>



```
In [44]: 1 average = df['damageDealt'].mean()
2 print(average)
```

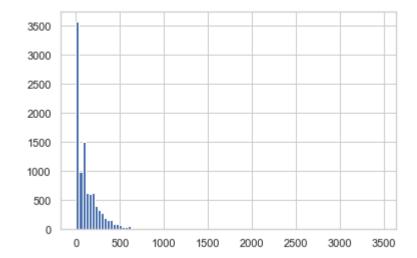
## 129.2112641000002

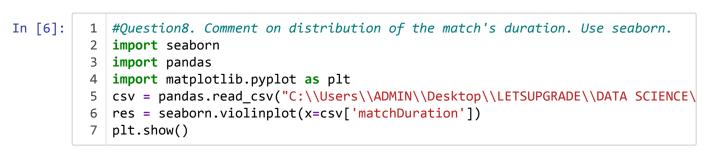
```
In [51]:

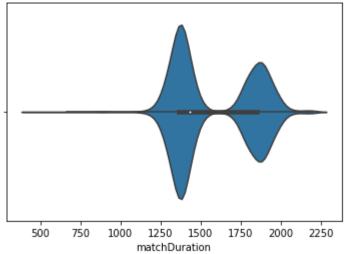
1 import pandas as pd
2 import numpy as np
3 import matplotlib.pyplot as plt
4 import seaborn as sns
```

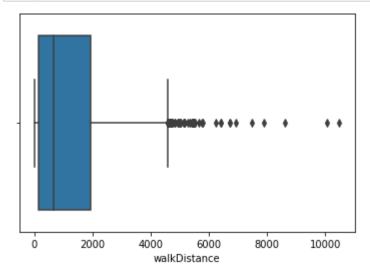
```
In [53]: 1 data_url = 'C:\\Users\\ADMIN\\Desktop\\LETSUPGRADE\\DATA SCIENCE\\NOTES\\pub
2 gapminder = pd.read_csv("C:\\Users\\ADMIN\\Desktop\\LETSUPGRADE\\DATA SCIENC
3 gapminder.head(n=3)
4 gapminder['damageDealt'].hist(bins=100)
```

Out[53]: <matplotlib.axes.\_subplots.AxesSubplot at 0xb83656abc8>









# In [9]:

- 1 import pandas as pd
- 2 import numpy as np
- 3 from matplotlib import pyplot as plt
- 4 %matplotlib inline
- 5 import seaborn as sns
- 6 sns.set()
- 7 **import** random

# In [10]:

- 1 #Question 20 Add a new column called 'KILL' which contains the sum of follow
- 2 teamKills, roadKills.
- data= pd.read\_csv("C:\\Users\\ADMIN\\Desktop\\LETSUPGRADE\\DATA SCIENCE\\NOT
- 4 data.head()

# Out[10]:

	ld	groupld	matchld	assists	boosts	damageDealt	DBNOs	head
0	2f262dd9795e60	78437bcd91d40e	d5db3a49eb2955	0	0	0.0	0	
1	a32847cf5bf34b	85b7ce5a12e10b	65223f05c7fdb4	0	0	163.2	1	
2	1b1900a9990396	edf80d6523380a	1cadec4534f30a	0	3	278.7	2	
3	f589dd03b60bf2	804ab5e5585558	c4a5676dc91604	0	0	191.9	1	
4	c23c4cc5b78b35	b3e2cd169ed920	cd595700a01bfa	0	0	100.0	1	

5 rows × 31 columns

4

#### In [19]: 1 #Question21. Round off column 'winPlacePerc' to 2 decimals. 2 df = pd.read\_csv("C:\\Users\\ADMIN\\Desktop\\LETSUPGRADE\\DATA SCIENCE\\NOTE 3 print(df["winPlacePerc"]) 4 df.round() 0 0.0000 1 0.2222 2 0.8571 3 0.3462 4 0.0690 9995 0.8333 9996 0.7174 9997 0.2083 9998 0.2449 9999 0.1875 Name: winPlacePerc, Length: 10000, dtype: float64

# Out[19]:

	ld	groupld	matchld	assists	boosts	damageDealt	DBNOs	h
0	2f262dd9795e60	78437bcd91d40e	d5db3a49eb2955	0	0	0.0	0	
1	a32847cf5bf34b	85b7ce5a12e10b	65223f05c7fdb4	0	0	163.0	1	
2	1b1900a9990396	edf80d6523380a	1cadec4534f30a	0	3	279.0	2	
3	f589dd03b60bf2	804ab5e5585558	c4a5676dc91604	0	0	192.0	1	
4	c23c4cc5b78b35	b3e2cd169ed920	cd595700a01bfa	0	0	100.0	1	
9995	ef4f474acd8e85	2eca2a8391f75d	492ecdfae90b46	0	3	204.0	1	
9996	cf0bf82fb4d80e	2eaf2765f93adb	14bffd71e96320	0	0	0.0	0	
9997	a0a31a0b1dcbe1	8d50c64ccc5071	147e4bbb62e3bb	0	0	0.0	0	
9998	f6874657399d69	d31843d7e62ccb	662567dcf280f5	0	0	0.0	0	
9999	90359b0b8f8b0d	61d5b1bb8da43f	258bfa48d88014	0	0	0.0	0	

10000 rows × 31 columns

```
In [20]: 1 # Question 7: Print all the columns of the dataframe.
2 numerical_cols = data.describe().columns
3 df = data[numerical_cols]
4 df
```

#### Out[20]:

	assists	boosts	damageDealt	DBNOs	headshotKills	heals	killPlace	killPoints	kills	killSt
0	0	0	0.0	0	0	0	92	1126	0	
1	0	0	163.2	1	1	0	42	1309	1	
2	0	3	278.7	2	1	8	16	0	2	
3	0	0	191.9	1	0	0	31	0	1	
4	0	0	100.0	1	0	0	87	1332	0	
9995	0	3	204.5	1	0	0	17	1033	2	
9996	0	0	0.0	0	0	0	49	1000	0	
9997	0	0	0.0	0	0	0	83	1334	0	
9998	0	0	0.0	0	0	0	82	1335	0	
9999	0	0	0.0	0	0	0	78	0	0	

10000 rows × 27 columns