

	<pre>Request_Closing_In_Hr 0 dtype: int64 s1 = sample_data[sample_data['Complaint Type'] == complaints_type_top5_names[0]].Request_Closing_In_Hr s1.head() 1</pre>
Out[56]: In [57]: Out[57]:	<pre>3 7.75 4 3.45 5 1.89 6 1.96 8 8.55 Name: Request_Closing_In_Hr, dtype: float64 s3 = sample_data[sample_data['Complaint Type'] == complaints_type_top5_names[2]].Request_Closing_In_Hr s3.head() 0 0.92 12 2.48 19 0.78 38 0.49 54 1.50 Name: Request_Closing_In_Hr, dtype: float64</pre>
In [58]: Out[58]: In [96]:	<pre>s4 = sample_data[sample_data['Complaint Type'] == complaints_type_top5_names[3]].Request_Closing_In_Hr s4.head() 17 0.85 18 2.93 22 1.26 29 2.50 30 1.99 Name: Request_Closing_In_Hr, dtype: float64 print(s1.isnull().sum()) print(s2.isnull().sum()) print(s3.isnull().sum()) print(s4.isnull().sum())</pre>
In [97]: Dut[97]: In [167	<pre>stats.f_oneway(s1, s2, s3, s4) F_onewayResult(statistic=1055.487168711559, pvalue=0.0) We can see pvalue is less than 0.05 so we reject null hypothesis and average response time is not same ### Try ChiSquare Test for second one - # Are the type of complaint or service requested and location related</pre>
In [71]: Out[71]:	<pre># H0 : 2 categories - Complain Type and Location is independent means not related # Ha : 2 categories - Complain Type and Location is dependent means related top_10_city = data['City'].value_counts()[:10] top_10_city BROOKLYN</pre>
In [72]: Out[72]: In [81]: Out[81]:	<pre>top10_city_names Index(['BROOKLYN', 'NEW YORK', 'BRONX', 'STATEN ISLAND', 'JAMAICA', 'ASTORIA',</pre>
In [172	ch2, p_value, df, exp_frq = stats.chi2_contingency(pd.crosstab(sample_data_city['Complaint Type'], sample_data_print(ch2) print(p_value) 50406.906171855495 0.0 Second way to check Null and Alternate and then provide a statistical test to accept
In [84]: Out[84]:	or reject the Null Hypothesis along with the corresponding 'p-value'. date_data = pd.read_csv("311_Service_Requests_from_2010_to_Present.csv", parse_dates=["Created Date", "Closed date_data["Request_Closing_Time"] = date_data["Closed Date"] - date_data["Created Date"] date_data['Request_Closing_Time'] = date_data.Request_Closing_Time.astype('int64') // 10**9 date_data C:\Users\abman\AppData\Local\Temp\ipykernel_11320\793570299.py:1: DtypeWarning: Columns (48,49) have mixed ty s. Specify dtype option on import or set low_memory=False. date_data = pd.read_csv("311_Service_Requests_from_2010_to_Present.csv", parse_dates=["Created Date", "Closed Date"]) Unique Created Closed Appendix Agency Complaint Descriptors Incident Incident Unique Created Closed Appendix Parameters Incident Incident Unique Created Closed Incident Incident Incident Unique Created Closed Incident Incide
	Name
	23:57:46 07:43:00 Department Parking AVENUE 4 32306529
In [85]:	300695 30283424 03-29 03-29 NYPD City Police Department
In [181	from matplotlib import style import seaborn as sns data['Complaint Type'].value_counts().head(10).plot(kind='barh',alpha=0.6,figsize=(10,5)); Noise-Park- Homeless Encampment- Traffic- Animal Abuse- Noise-Vehicle- Derelict Vehicle-
In [86]: Out[86]:	Noise - Commercial - Noise - Street/Sidewalk - Illegal Parking - Blocked Driveway - 0 10000 20000 30000 40000 50000 60000 70000 80000 meanclose1= date_data['Request_Closing_Time'].mean() meanclose1 -66361400.168607704
In [19]: Out[19]: In [87]: Out[87]: In [94]:	<pre>from scipy.stats import normaltest meanclose= date_data['Request_Closing_Time'].mean() stat, p = normaltest(date_data['Request_Closing_Time']) meanclose -5742847521.905661 from scipy.stats import ttest_1samp tset, pval = ttest_1samp(date_data['Request_Closing_Time'], date_data['Request_Closing_Time'].mean()) pval 1.0 from scipy import stats from statsmodels.stats import weightstats as stests</pre>
In [88]:	<pre>ztest ,pval = stests.ztest(date_data['Request_Closing_Time'], x2=None, value=date_data['Request_Closing_Time print(float(pval)) 1.0 contingency_table=pd.crosstab(data["Complaint Type"],data["Location Type"]) print('contingency_table :-\n',contingency_table)#Observed Values Observed_Values = contingency_table.values print("Observed Values :-\n",Observed_Values) b=stats.chi2_contingency(contingency_table) Expected_Values = b[3] print("Expected Values :-\n",Expected_Values)</pre>
	Contingency_table :- Location Type
	Noise - House of Worship 0 0 0 0 Noise - Park 0 0 0 0 Noise - Street/Sidewalk 0 0 0 0 Noise - Vehicle 0 0 0 0 Panhandling 0 0 0 0 Posting Advertisement 0 0 0 0 Squeegee 0 0 0 0 Traffic 0 0 0 0 Urinating in Public 0 21 0 0 Vending 0 0 0 0 Location Type Highway House and Store House of Worship Park \ Complaint Type Animal Abuse 0 93 0 0 Animal in a Park 0 0 0 0 1 Bike/Roller/Skate Chronic 0 0 0 0 0
	Blocked Driveway 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
	Posting Advertisement 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
	Graffiti 0 0 0 Homeless Encampment 353 0 0 Illegal Fireworks 8 0 0 Illegal Parking 0 0 0 Noise - Commercial 0 0 0 Noise - House of Worship 0 0 0 Noise - Park 4041 0 0 Noise - Street/Sidewalk 0 0 0 Noise - Vehicle 0 0 0 Panhandling 6 0 0 Posting Advertisement 0 7 0 Squeegee 0 0 0 Traffic 0 0 0 Urinating in Public 38 0 0 Vending 106 0 0
	Location Type Residential Building/House Roadway Tunnel \ Complaint Type Animal Abuse 5085 0 Animal in a Park 0 0 0 Bike/Roller/Skate Chronic 26 0 Blocked Driveway 0 0 0 Derelict Vehicle 0 5 Disorderly Youth 77 0 Drinking 291 0 Ferry Complaint 0 0 0 Graffiti 56 0 Homeless Encampment 983 1 Illegal Fireworks 33 0 Illegal Parking 0 0 0 Noise - Commercial 0 0 0 Noise - House of Worship 0 0
	Blocked Driveway 0 77007 0 Derelict Vehicle 0 17614 0 Disorderly Youth 8 201 0 Drinking 90 434 0 Ferry Complaint 0 0 0 Graffiti 32 25 0 Homeless Encampment 512 2541 0 Illegal Fireworks 2 125 0 Illegal Parking 0 75326 0 Noise - Commercial 18598 0 0 Noise - House of Worship 0 0 0 Noise - Park 0 0 0 Noise - Street/Sidewalk 0 48601 0 Noise - Vehicle 0 17080 0 Panhandling 60 225 0 Posting Advertisement 6 582 0
	Graffiti 0 0 0 Homeless Encampment 0 0 Illegal Fireworks 0 0 Illegal Parking 0 0 0 Noise - Commercial 0 0 0 Noise - House of Worship 0 0 Noise - Park 0 0 0 Noise - Street/Sidewalk 0 0 Noise - Vehicle 0 0 Panhandling 0 0 Posting Advertisement 0 0 Squeegee 0 0 0 Traffic 0 0 0 Urinating in Public 0 0 Observed Values :-
	Observed Values :- [[0 0 62 0 0 93 0 0 123 110 227 5085
	[0 0 0 0 1 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
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	1.42064831e-03 1.09389920e-01] [5.12411542e-01 4.44773219e+03 1.58847578e+01 2.56205771e-01 5.50842408e+01 2.38271367e+01 2.38015161e+02 2.56205771e-01 1.22287015e+03 2.99760752e+01 5.81587100e+01 1.78319217e+03 8.96720199e+00 5.22172982e+03 6.38718425e+04 8.71099622e+00 2.56205771e-01 1.97278444e+01] [1.17843942e-01 1.02288541e+03 3.65316219e+00 5.89219708e-02 1.26682237e+01 5.47974329e+00 5.47385109e+01 5.89219708e-02 2.81234567e+02 6.89387058e+00 1.33752874e+01 4.10096917e+02 2.06226898e+00 1.20088869e+03 1.46891884e+04 2.00334701e+00 5.89219708e-02 4.53699175e+00] [1.90306986e-03 1.65186464e+01 5.89951658e-02 9.51534932e-04 2.04580010e-01 8.84927487e-02 8.83975952e-01 9.51534932e-04 4.54167623e+00 1.11329587e-01 2.15998430e-01 6.62268313e+00 3.33037226e-02 1.93932335e+01 2.37216707e+02 3.23521877e-02 9.51534932e-04 7.32681898e-02]
	[8.51058167e-03 7.38718489e+01 2.63828032e-01 4.25529083e-03 9.14887529e-01 3.95742048e-01 3.95316518e+00 4.25529083e-03 2.03105031e+01 4.97869028e-01 9.65951019e-01 2.96168242e+01 1.48935179e-01 8.67270825e+01 1.06083975e+03 1.44679888e-01 4.25529083e-03 3.27657394e-01] [1.33081809e-05 1.15515010e-01 4.12553607e-04 6.65409044e-06 1.43062944e-03 6.1883041le-04 6.18165001e-03 6.65409044e-06 3.17599736e-02 7.78528581e-04 1.51047853e-03 4.63124694e-02 2.32893165e-04 1.35617017e-01 1.65885809e+00 2.26239075e-04 6.65409044e-06 5.12364964e-04] [7.51912219e-04 6.52659806e+00 2.33092788e-02 3.75956110e-04 8.08305636e-02 3.49639182e-02 3.49263226e-01 3.75956110e-04 1.79443851e+00 4.39868648e-02 8.53420369e-02 2.61665452e+00 1.31584638e-02 7.66236147e+00 9.37254822e+01 1.27825077e-02 3.75956110e-04 2.89486204e-02] [2.93245766e-02 2.54537324e+02 9.09061873e-01 1.46622883e-02
	[2.93245766e-02 2.54537324e+02 9.09061873e-01 1.46622883e-02 3.15239198e+00 1.36359281e+00 1.36212658e+01 1.46622883e-02 6.99831019e+01 1.71548773e+00 3.32833944e+00 1.02049526e+02 5.13180090e-01 2.98832097e+02 3.65529380e+03 4.98517801e-01 1.46622883e-02 1.12899620e+00] [1.11788719e-03 9.70326084e+00 3.46545030e-02 5.58943597e-04 1.20172873e-01 5.19817545e-02 5.19258601e-01 5.58943597e-04 2.66783779e+00 6.53964008e-02 1.26880196e-01 3.89024743e+00 1.95630259e-02 1.13918294e+01 1.39344080e+02 1.90040823e-02 5.58943597e-04 4.30386569e-02] [5.01226016e-01 4.35064182e+03 1.55380065e+01 2.50613008e-01 5.38817967e+01 2.33070098e+01 2.32819485e+02 2.50613008e-01 1.19617589e+03 2.93217219e+01 5.68891528e+01 1.74426654e+03 8.77145528e+00 5.10774372e+03 6.24775723e+04 8.52084227e+00 2.50613008e-01 1.92972016e+01] [2.36692651e-01 2.05449221e+03 7.33747218e+00 1.18346325e-01 2.54444600e+01 1.10062083e+01 1.09943736e+02 1.18346325e-01
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	7.71783240e+02 1.89186338e+01 3.67053835e+01 1.12541616e+03 5.65942036e+00 3.29556133e+03 4.03110811e+04 5.49772264e+00 1.61697725e-01 1.24507248e+01] [1.13651865e-01 9.86498185e+02 3.52320780e+00 5.68259323e-02 1.22175754e+01 5.28481171e+00 5.27912911e+01 5.68259323e-02 2.71230175e+02 6.64863408e+00 1.28994866e+01 3.95508489e+02 1.98890763e+00 1.15816933e+03 1.41666481e+04 1.93208170e+00 5.68259323e-02 4.37559679e+00] [2.04280576e-03 1.77315540e+01 6.33269787e-02 1.02140288e-03 2.19601620e-01 9.49904680e-02 9.48883277e-01 1.02140288e-03 4.87515596e+00 1.19504137e-01 2.31858454e-01 7.10896406e+00 3.57491009e-02 2.08172121e+01 2.54634717e+02 3.47276980e-02 1.02140288e-03 7.86480219e-02] [4.31850469e-03 3.74846207e+01 1.33873645e-01 2.15925235e-03 4.64239254e-01 2.00810468e-01 2.00594543e+00 2.15925235e-03 1.03061114e+01 2.52632525e-01 4.90150283e-01 1.50283963e+01
	7.55738321e-02 4.40077221e+01 5.38299451e+02 7.34145798e-02 2.15925235e-03 1.66262431e-01] [2.66163617e-05 2.31030020e-01 8.25107214e-04 1.33081809e-05 2.86125889e-03 1.23766082e-03 1.23633000e-02 1.33081809e-05 6.35199473e-02 1.55705716e-03 3.02095706e-03 9.26249389e-02 4.65786331e-04 2.71234034e-01 3.31771618e+00 4.52478150e-04 1.33081809e-05 1.02472993e-03] [2.98968283e-02 2.59504470e+02 9.26801678e-01 1.49484142e-02 3.21390905e+00 1.39020252e+00 1.38870768e+01 1.49484142e-02 7.13487808e+01 1.74896446e+00 3.39329002e+00 1.04040963e+02 5.23194496e-01 3.04663629e+02 3.72662470e+03 5.08246082e-01 1.49484142e-02 1.15102789e+00] [3.93256745e-03 3.41346854e+01 1.21909591e-01 1.96628372e-03 4.22751001e-01 1.82864386e-01 1.82667758e+00 1.96628372e-03
In []:	9.38507221e+00 2.30055196e-01 4.46346405e-01 1.36853347e+01 6.88199303e-02 4.00748286e+01 4.90192566e+02 6.68536466e-02 1.96628372e-03 1.51403847e-01] [2.52855437e-02 2.19478519e+02 7.83851853e-01 1.26427718e-02 2.71819594e+00 1.17577778e+00 1.17451350e+01 1.26427718e-02 6.03439499e+01 1.47920430e+00 2.86990920e+00 8.79936919e+01 4.42497014e-01 2.57672333e+02 3.15183037e+03 4.29854242e-01 1.26427718e-02 9.73493431e-01]] no_of_rows=len(contingency_table.iloc[0:2,0]) print("no_of_rows:-",no_of_rows) no_of_columns=len(contingency_table.iloc[0,0:2]) print("no_of_columns:-",no_of_columns) ddof=(no_of_rows-1)*(no_of_columns-1) print("Degree of Freedom:-",ddof) alpha = 0.05
In []:	<pre>from scipy.stats import chi2 chi_square=sum([(o-e)**2./e for o,e in zip(Observed_Values, Expected_Values)]) chi_square_statistic=chi_square[0]+chi_square[1] print("chi-square statistic:-",chi_square_statistic)</pre>
In []:	<pre>print('Degree of Freedom: ',ddof) print('chi-square statistic:',chi_square_statistic) print('critical_value:',critical_value) print('p-value:',p_value) if chi_square_statistic>=critical_value: print("Reject H0,There is a relationship between 2 categorical variables") else: print("Retain H0,There is no relationship between 2 categorical variables") if p_value<=alpha: print("Reject H0,There is a relationship between 2 categorical variables") else: print("Retain H0,There is a relationship between 2 categorical variables") else: print("Retain H0,There is no relationship between 2 categorical variables")</pre>
In [90]:	<pre>else:</pre>
Out[90]:	sortedComplaintType sortedComplaintType count Complaint Type count Blocked Driveway 77044 Illegal Parking 75361 Noise - Street/Sidewalk 48612 Noise - Commercial 35577 Derelict Vehicle 17718 Noise - Vehicle 17083
In [91]: Out[91]:	6 Animal Abuse 7778 7 Traffic 4498 8 Homeless Encampment 4416 9 Noise - Park 4042 10 Vending 3802 11 Drinking 1280 data . head () Unique Created Closed Agency Agency Complaint Descriptor Location Type Incident Incident Lot Ferry
.1:	Name
	3 32305098 12-31 01-01 NYPD City Police Illegal Parking Overnight Street/Sidewalk 10461.0 BAISLEY NaN NaN Parking 23:57:46 07:43:00 Department Parking Street/Sidewalk 10461.0 BAISLEY NaN AVENUE 4 32306529 2015- 2016- New York 12-31 01-01 NYPD City Police Illegal Parking 23:56:58 03:24:00 Department Street/Sidewalk Street/Sidewalk 11373.0 ROAD NaN NaN Sidewalk Street/Sidewalk 11373.0 BIOCKED Street/Sidewalk 11373.0 ROAD NaN NaN NaN Parking Sidewalk Street/Sidewalk 11373.0 ROAD NaN NaN NaN NaN NaN SortedComplaintType = sortedComplaintType.head() plt.figure(figsize=(6,6)) plt.pie(sortedComplaintType['count'],labels=sortedComplaintType["Complaint Type"], autopct="%1.1f%%")
In [92]:	plt.show()
In [92]:	Blocked Driveway 30.3% Perelict Vehicle 19.1%
Out[91]:	Drinking 1280 Drinking 1280 Drinking Drinki