

▼ PE29 Vasu Kalariya

IMLA Lab Assi 2 Data Preprocessing

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

ds = pd.read_csv('FireData.csv')
ds
```

	_id	Area_of_Origin	Building_Status	Business_Impact	Civilian_Casualties	Count_of_Persons_Rescued	Estimated_Dollar_Loss	Estimated_Number_Of_Persons_Displaced	Exposures	Ext_agen
	0	190306	81 - Engine Area	NaN	NaN	0	0	1000.0	NaN	NaN

ds.isnull().sum()

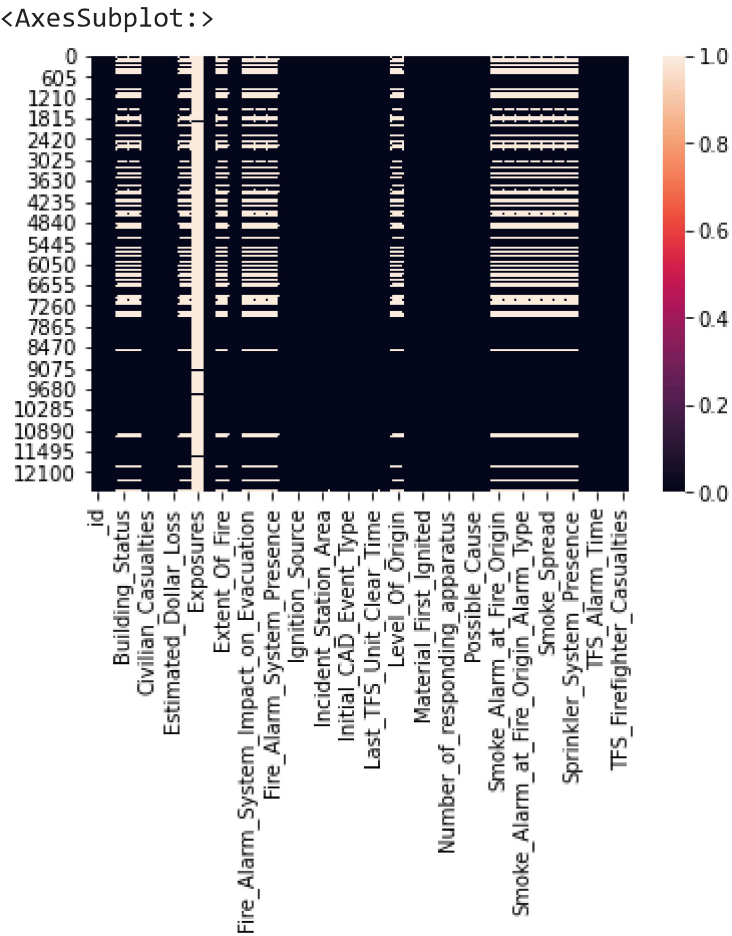
_id	0
Area_of_Origin	0
Building_Status	3482
Business_Impact	3484
Civilian_Casualties	0
Count_of_Persons_Rescued	0
Estimated_Dollar_Loss	1
Estimated_Number_Of_Persons_Displaced	3483
Exposures	12408
Ext_agent_app_or_defer_time	0
Extent_Of_Fire	3484
Final_Incident_Type	0
Fire_Alarm_System_Impact_on_Evacuation	3484
Fire_Alarm_System_Operation	3484
Fire_Alarm_System_Presence	3484
Fire_Under_Control_Time	1
Ignition_Source	0
Incident_Number	0
Incident_Station_Area	0
Incident_Ward	74
Initial_CAD_Event_Type	0
Intersection	1
Last_TFS_Unit_Clear_Time	0
Latitude	1
Level_Of_Origin	3484
Longitude	1
Material_First_Ignited	0
Method_Of_Fire_Control	0
Number_of_responding_apparatus	0
Number_of_responding_personnel	0
Possible_Cause	0
Property_Use	1
Smoke_Alarm_at_Fire_Origin	3484
Smoke_Alarm_at_Fire_Origin_Alarm_Failure	3484
Smoke_Alarm_at_Fire_Origin_Alarm_Type	3484
Smoke_Alarm_Impact_on_Persons_Evacuating_Impact_on_Evacuation	3484
Smoke_Spread	3484
Sprinkler_System_Operation	3484
Sprinkler_System_Presence	3484
Status_of_Fire_On_Arrival	0
TFS_Alarm_Time	0
TFS_Arrival_Time	0
TFS_Firefighter_Casualties	0

dtype: int64

ds.shape

(12687, 43)

```
sns.heatmap(ds.isnull())
```



```
ds.dropna(subset=['Building_Status'], inplace = True)
```

```
ds.isnull().sum()
```

_id	0
Area_of_Origin	0
Building_Status	0
Business_Impact	2
Civilian_Casualties	0
Count_of_Persons_Rescued	0
Estimated_Dollar_Loss	0
Estimated_Number_Of_Persons_Displaced	1
Exposures	8994
Ext_agent_app_or_defer_time	0
Extent_Of_Fire	2
Final_Incident_Type	0
Fire_Alarm_System_Impact_on_Evacuation	2
Fire_Alarm_System_Operation	2
Fire_Alarm_System_Presence	2
Fire_Under_Control_Time	0
Ignition_Source	0
Incident_Number	0
Incident_Station_Area	0

```
Incident_Ward 30
Initial_CAD_Event_Type 0
Intersection 0
Last_TFS_Unit_Clear_Time 0
Latitude 0
Level_Of_Origin 2
Longitude 0
Material_First_Ignited 0
Method_Of_Fire_Control 0
Number_of_responding_apparatus 0
Number_of_responding_personnel 0
Possible_Cause 0
Property_Use 0
Smoke_Alarm_at_Fire_Origin 2
Smoke_Alarm_at_Fire_Origin_Alarm_Failure 2
Smoke_Alarm_at_Fire_Origin_Alarm_Type 2
Smoke_Alarm_Impact_on_Persons_Evacuating_Impact_on_Evacuation 2
Smoke_Spread 2
Sprinkler_System_Operation 2
Sprinkler_System_Presence 2
Status_of_Fire_On_Arrival 0
TFS_Alarm_Time 0
TFS_Arrival_Time 0
TFS_Firefighter_Casualties 0
dtype: int64
```

```
ds['Exposures'].unique()

array([nan,  1.,  3.,  2.,  4.,  6.,  5.,  7.] )
```

```
Exposures1 = ds.iloc[:,8:9].values
Exposures2 = ds.iloc[:,8:9].values
Exposures3 = ds.iloc[:,8:9].values
Exposures1

array([[nan],
       [nan],
       [nan],
       ...,
       [nan],
       [nan],
       [nan]])
```

▼ Imputer for Missing values

```
from sklearn.impute import SimpleImputer
si1 = SimpleImputer(missing_values=np.nan, strategy='mean')
si1.fit(Exposures1)
Exposures1 = si1.transform(Exposures1)
Exposures1 = pd.DataFrame(Exposures1, columns=[ 'Exposures1' ])
Exposures1
```

Exposures1	
0	1.649289
1	1.649289
2	1.649289
3	1.649289
4	1.649289
...	...
9200	1.649289
9201	1.649289
9202	1.649289
9203	1.649289
9204	1.649289

9205 rows × 1 columns

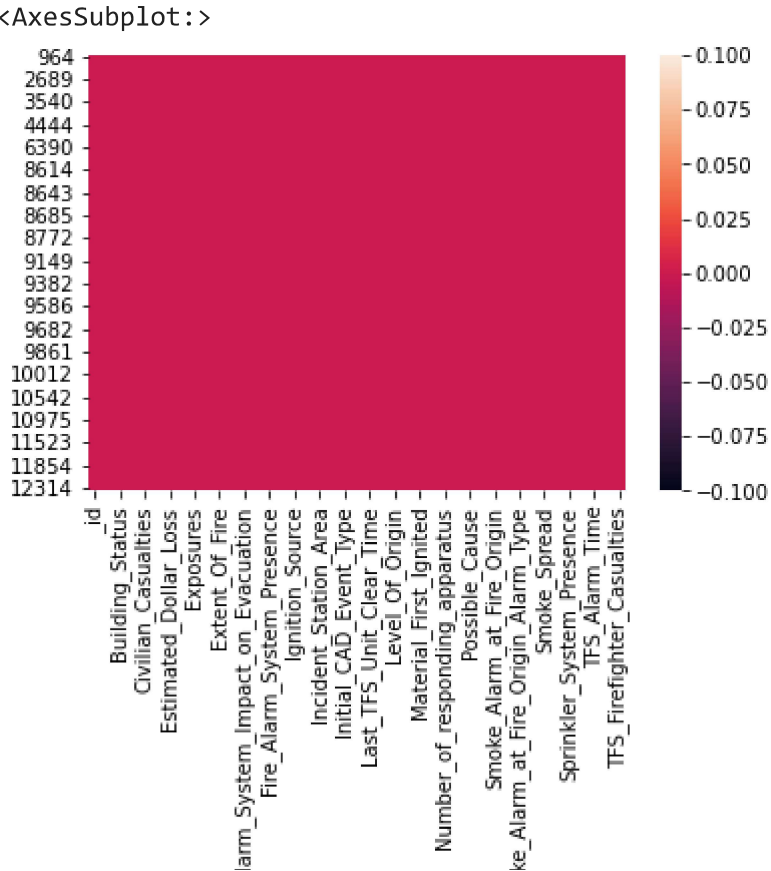
```
si2 = SimpleImputer(missing_values=np.nan,strategy='median')
si2.fit(Exposures2)
Exposures2 = si2.transform(Exposures2)
Exposures2 = pd.DataFrame(Exposures2, columns=['Exposures2'])
Exposures2
```

```
ds.isnull().sum()

_id 0
Area_of_Origin 0
Building_Status 0
Business_Impact 2
Civilian_Casualties 0
Count_of_Persons_Rescued 0
Estimated_Dollar_Loss 0
Estimated_Number_Of_Persons_Displaced 1
Exposures 8994
Ext_agent_app_or_defer_time 0
Extent_Of_Fire 2
Final_Incident_Type 0
Fire_Alarm_System_Impact_on_Evacuation 2
Fire_Alarm_System_Operation 2
Fire_Alarm_System_Presence 2
Fire_Under_Control_Time 0
Ignition_Source 0
Incident_Number 0
Incident_Station_Area 0
Incident_Ward 30
Initial_CAD_Event_Type 0
Intersection 0
Last_TFS_Unit_Clear_Time 0
Latitude 0
Level_Of_Origin 2
Longitude 0
Material_First_Ignited 0
Method_Of_Fire_Control 0
Number_of_responding_apparatus 0
Number_of_responding_personnel 0
Possible_Cause 0
Property_Use 0
Smoke_Alarm_at_Fire_Origin 2
Smoke_Alarm_at_Fire_Origin_Alarm_Failure 2
Smoke_Alarm_at_Fire_Origin_Alarm_Type 2
Smoke_Alarm_Impact_on_Persons_Evacuating_Impact_on_Evacuation 2
Smoke_Spread 2
Sprinkler_System_Operation 2
Sprinkler_System_Presence 2
Status_of_Fire_On_Arrival 0
TFS_Alarm_Time 0
TFS_Arrival_Time 0
TFS_Firefighter_Casualties 0
dtype: int64
```

```
ds.dropna(axis=0, how="any", thresh=None, subset=None, inplace=True)
```

```
sns.heatmap(ds.isnull())
```



```
ds = ds.reset_index()
```

```
ds['Estimated_Dollar_Loss'] = ds['Estimated_Dollar_Loss'].astype('int64')
ds['Estimated_Dollar_Loss'].unique()
```

```
array([[ 2000,    200,   5000,  70000, 100000,   7500,  10000,
         400,  90000,   1000,  20000,   2500,  25000,  50000,
        30000,  40000, 350000,    500,  60000,   9000, 1000000,
        80000, 500000, 400000, 5000000, 250000, 200000,  750000,
       800000, 125000,  700000,  34000, 300000,   9999, 2000000,
        75000, 1500000, 150000,   4000,   15000,   7000,    250,
       120000,    20, 175000,  18000,   65000,  12000,   35000,
        49800,  600000,   55000,    100, 1850000,   8000], dtype=int64)
```

▼ Normalisation using Max Min, L1 , L2 , Zscore

```
Estimated_Dollar_Loss = ds.iloc[:, 6:7].values

from sklearn import preprocessing
min_max = preprocessing.MinMaxScaler()
Estimated_Dollar_Loss= min_max.fit_transform(Estimated_Dollar_Loss)
Estimated_Dollar_Loss = pd.DataFrame(Estimated_Dollar_Loss,columns = ['Estimated_Dollar_Loss'])
Estimated_Dollar_Loss
```

Estimated_Dollar_Loss	
0	0.0
1	0.0
2	0.0
3	0.0
4	0.0
...	...
206	0.0
207	0.0
208	0.0
209	0.0
210	0.0

211 rows × 1 columns

Estimated_Dollar_Loss2 = ds.iloc[:, 6:7].values

```
from sklearn.preprocessing import Normalizer
Data_normalizer= Normalizer(norm='l1').fit(Estimated_Dollar_Loss2)
Estimated_Dollar_Loss2 = Data_normalizer.transform(Estimated_Dollar_Loss2)
Estimated_Dollar_Loss2 = pd.DataFrame(Estimated_Dollar_Loss2,columns = ['Estimated_Dollar_Loss2'])
Estimated_Dollar_Loss2
```

Estimated_Dollar_Loss2	
0	0.0
1	0.0
2	0.0
3	0.0
4	0.0
...	...
206	0.0
207	0.0
208	0.0
209	0.0
210	0.0

211 rows × 1 columns


```
Estimated_Dollar_Loss3 = ds.iloc[:, 6:7].values

from sklearn.preprocessing import Normalizer
Data_normalizer= Normalizer(norm='l2').fit(Estimated_Dollar_Loss3)
Estimated_Dollar_Loss3 = Data_normalizer.transform(Estimated_Dollar_Loss3)
Estimated_Dollar_Loss3 = pd.DataFrame(Estimated_Dollar_Loss3,columns = ['Estimated_Dollar_Loss3'])
Estimated_Dollar_Loss3
```

Estimated_Dollar_Loss3	
0	0.0
1	0.0
2	0.0
3	0.0
4	0.0
...	...
206	0.0
207	0.0
208	0.0
209	0.0
210	0.0

211 rows × 1 columns

```
Estimated_Number_Of_Persons_Displaced = ds.iloc[:, 7:8].values

from scipy import stats
Estimated_Number_Of_Persons_Displaced = stats.zscore(Estimated_Number_Of_Persons_Displaced)
Estimated_Number_Of_Persons_Displaced = pd.DataFrame(Estimated_Number_Of_Persons_Displaced,columns = ['Estimated_Number_Of_Persons_Displaced'])
Estimated_Number_Of_Persons_Displaced
```

Estimated_Number_Of_Persons_Displaced	
0	-0.467214
1	-0.471048
2	-0.460825
3	-0.322387
4	-0.258493

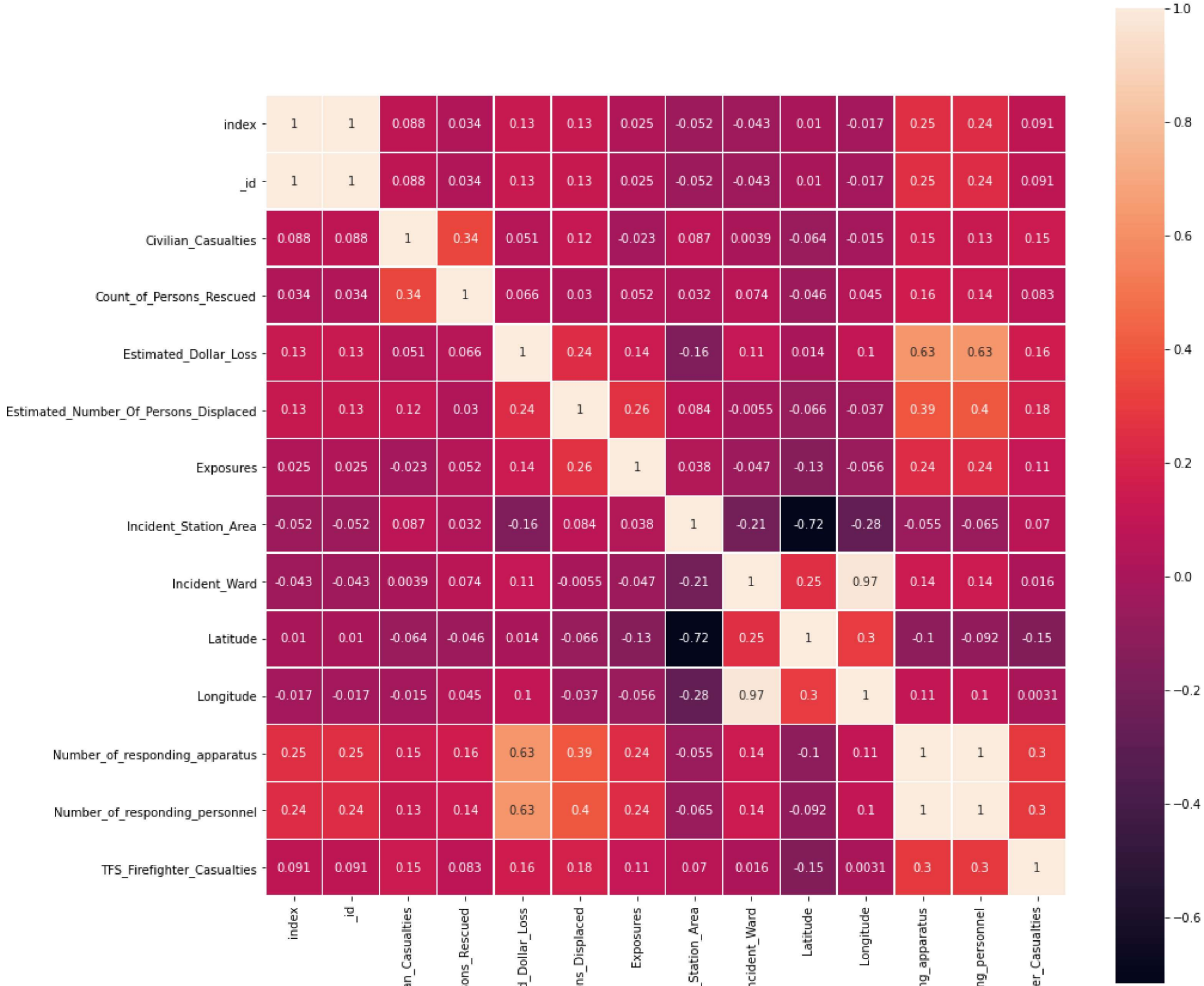
▼ Feature Selection using Correlation Coefficient

```
ds.corr(method = 'pearson')
```

	index	_id	Civilian_Casualties	Count_of_Persons_Rescued	Estimated_Dollar_Loss	Estimated_Number_Of_Persons_Displaced	Exposures	Incident_Stat
index	1.000000	1.000000	0.088213	0.034237	0.128786	0.125300	0.025089	
_id	1.000000	1.000000	0.088213	0.034237	0.128786	0.125300	0.025089	
Civilian_Casualties	0.088213	0.088213	1.000000	0.336775	0.050592	0.120813	-0.022841	
Count_of_Persons_Rescued	0.034237	0.034237	0.336775	1.000000	0.066351	0.030255	0.051705	
Estimated_Dollar_Loss	0.128786	0.128786	0.050592	0.066351	1.000000	0.236972	0.135748	
Estimated_Number_Of_Persons_Displaced	0.125300	0.125300	0.120813	0.030255	0.236972	1.000000	0.255827	
Exposures	0.025089	0.025089	-0.022841	0.051705	0.135748	0.255827	1.000000	
Incident_Station_Area	-0.051774	-0.051774	0.087232	0.032444	-0.160756	0.083949	0.037605	
Incident_Ward	-0.043413	-0.043413	0.003947	0.074149	0.109913	-0.005542	-0.046863	
Latitude	0.010110	0.010110	-0.064315	-0.046431	0.013641	-0.066485	-0.130792	
Longitude	-0.017260	-0.017260	-0.014582	0.044854	0.103475	-0.037393	-0.055659	
Number_of_responding_apparatus	0.249495	0.249495	0.152842	0.163182	0.629676	0.391738	0.241793	
Number_of_responding_personnel	0.239354	0.239354	0.134114	0.144415	0.630974	0.396794	0.241166	
TFS_Firefighter_Casualties	0.091333	0.091333	0.147323	0.083274	0.162295	0.176266	0.109110	

```
plt.subplots(figsize=(15,15))
sns.heatmap(ds.corr(), annot = True,annot_kws={'size': 10},linewidths=.5,square=True)
```

<AxesSubplot:>



▼ Encoding of categorical values

```
Final_Incident_Type = ds.loc[:, ['Final_Incident_Type']].values
```

```
from sklearn.preprocessing import LabelEncoder
le = LabelEncoder()
Final_Incident_Type = le.fit_transform(Final_Incident_Type)
Final_Incident_Type = pd.DataFrame(Final_Incident_Type,columns = ['Final_Incident_Type(Explosion)'])
Final_Incident_Type
```

C:\Users\kalar\anaconda3\lib\site-packages\sklearn\utils\validation.py:72: DataConversionWarning: A column-vector y was passed when a 1d array was expected. Please change the shape of y to (n_s
return f(**kwargs)

Final_Incident_Type(Explosion)	
0	0
1	0
2	0
3	0
4	0
...	...
206	0
207	0
208	0
209	0
210	0

211 rows × 1 columns

```
Method_Of_Fire_Control = pd.get_dummies(ds['Method_Of_Fire_Control'])
Method_Of_Fire_Control
```

	1 - Extinguished by fire department	3 - Extinguished by occupant	4 - Fire self extinguished	5 - Action taken unclassified
0	1	0	0	0
1	0	0	1	0
2	1	0	0	0

```
Method_Of_Fire_Control1 = ds.loc[:,['Method_Of_Fire_Control']].values
```

```
from sklearn.compose import ColumnTransformer
from sklearn.preprocessing import OneHotEncoder
```

```
ct = ColumnTransformer(transformers = [('encoder',OneHotEncoder(), [0])],remainder='passthrough')
Method_Of_Fire_Control1 = np.array(ct.fit_transform(Method_Of_Fire_Control1))
print(Method_Of_Fire_Control1)
```

```
(0, 0)      1.0
(1, 2)      1.0
(2, 0)      1.0
(3, 0)      1.0
(4, 0)      1.0
(5, 0)      1.0
(6, 0)      1.0
(7, 0)      1.0
(8, 0)      1.0
(9, 0)      1.0
(10, 1)     1.0
(11, 0)     1.0
(12, 1)     1.0
(13, 0)     1.0
(14, 1)     1.0
(15, 0)     1.0
(16, 0)     1.0
(17, 0)     1.0
(18, 0)     1.0
(19, 0)     1.0
(20, 0)     1.0
(21, 0)     1.0
(22, 0)     1.0
(23, 0)     1.0
(24, 0)     1.0
:          :
(186, 0)    1.0
(187, 0)    1.0
(188, 0)    1.0
(189, 0)    1.0
(190, 0)    1.0
(191, 3)    1.0
(192, 0)    1.0
(193, 0)    1.0
(194, 0)    1.0
(195, 0)    1.0
(196, 0)    1.0
(197, 0)    1.0
(198, 0)    1.0
```

```
(199, 0)      1.0
(200, 0)      1.0
(201, 0)      1.0
(202, 0)      1.0
(203, 0)      1.0
(204, 0)      1.0
(205, 0)      1.0
(206, 0)      1.0
(207, 0)      1.0
(208, 0)      1.0
(209, 0)      1.0
(210, 0)      1.0
```

```
Method_Of_Fire_Control2 = ds.loc[:,['Method_Of_Fire_Control']].values
Method_Of_Fire_Control2 = Method_Of_Fire_Control2.ravel()

from sklearn.feature_extraction.text import CountVectorizer
from sklearn.feature_extraction.text import TfidfVectorizer
vectorizer = CountVectorizer()
Method_Of_Fire_Control2 = vectorizer.fit_transform(Method_Of_Fire_Control2)
print(Method_Of_Fire_Control2)
```

```
(0, 3)      1
(0, 1)      1
(0, 4)      1
(0, 2)      1
(1, 3)      1
(1, 4)      1
(1, 6)      1
(2, 3)      1
(2, 1)      1
(2, 4)      1
(2, 2)      1
(3, 3)      1
(3, 1)      1
(3, 4)      1
(3, 2)      1
(4, 3)      1
(4, 1)      1
(4, 4)      1
(4, 2)      1
(5, 3)      1
(5, 1)      1
(5, 4)      1
(5, 2)      1
(6, 3)      1
(6, 1)      1
:          :
(204, 2)    1
(205, 3)    1
(205, 1)    1
(205, 4)    1
(205, 2)    1
(206, 3)    1
```

```
(206, 1)      1
(206, 4)      1
(206, 2)      1
(207, 3)      1
(207, 1)      1
(207, 4)      1
(207, 2)      1
(208, 3)      1
(208, 1)      1
(208, 4)      1
(208, 2)      1
(209, 3)      1
(209, 1)      1
(209, 4)      1
(209, 2)      1
(210, 3)      1
(210, 1)      1
(210, 4)      1
(210, 2)      1
```

▼ Feature Reduction using Variance Threshold

```
from sklearn import datasets
df = datasets.load_iris(as_frame=True)
X = df.data
y = df.target
print(X)
```

	sepal length (cm)	sepal width (cm)	petal length (cm)	petal width (cm)
0	5.1	3.5	1.4	0.2
1	4.9	3.0	1.4	0.2
2	4.7	3.2	1.3	0.2
3	4.6	3.1	1.5	0.2
4	5.0	3.6	1.4	0.2
..
145	6.7	3.0	5.2	2.3
146	6.3	2.5	5.0	1.9
147	6.5	3.0	5.2	2.0
148	6.2	3.4	5.4	2.3
149	5.9	3.0	5.1	1.8

[150 rows x 4 columns]

```
from sklearn.feature_selection import VarianceThreshold
```

```
selector = VarianceThreshold()
selector.fit_transform(X,y)
```

```
array([[5.1, 3.5, 1.4, 0.2],
       [4.9, 3. , 1.4, 0.2],
       [4.7, 3.2, 1.3, 0.2],
```

[4.6, 3.1, 1.5, 0.2],
[5. , 3.6, 1.4, 0.2],
[5.4, 3.9, 1.7, 0.4],
[4.6, 3.4, 1.4, 0.3],
[5. , 3.4, 1.5, 0.2],
[4.4, 2.9, 1.4, 0.2],
[4.9, 3.1, 1.5, 0.1],
[5.4, 3.7, 1.5, 0.2],
[4.8, 3.4, 1.6, 0.2],
[4.8, 3. , 1.4, 0.1],
[4.3, 3. , 1.1, 0.1],
[5.8, 4. , 1.2, 0.2],
[5.7, 4.4, 1.5, 0.4],
[5.4, 3.9, 1.3, 0.4],
[5.1, 3.5, 1.4, 0.3],
[5.7, 3.8, 1.7, 0.3],
[5.1, 3.8, 1.5, 0.3],
[5.4, 3.4, 1.7, 0.2],
[5.1, 3.7, 1.5, 0.4],
[4.6, 3.6, 1. , 0.2],
[5.1, 3.3, 1.7, 0.5],
[4.8, 3.4, 1.9, 0.2],
[5. , 3. , 1.6, 0.2],
[5. , 3.4, 1.6, 0.4],
[5.2, 3.5, 1.5, 0.2],
[5.2, 3.4, 1.4, 0.2],
[4.7, 3.2, 1.6, 0.2],
[4.8, 3.1, 1.6, 0.2],
[5.4, 3.4, 1.5, 0.4],
[5.2, 4.1, 1.5, 0.1],
[5.5, 4.2, 1.4, 0.2],
[4.9, 3.1, 1.5, 0.2],
[5. , 3.2, 1.2, 0.2],
[5.5, 3.5, 1.3, 0.2],
[4.9, 3.6, 1.4, 0.1],
[4.4, 3. , 1.3, 0.2],
[5.1, 3.4, 1.5, 0.2],
[5. , 3.5, 1.3, 0.3],
[4.5, 2.3, 1.3, 0.3],
[4.4, 3.2, 1.3, 0.2],
[5. , 3.5, 1.6, 0.6],
[5.1, 3.8, 1.9, 0.4],
[4.8, 3. , 1.4, 0.3],
[5.1, 3.8, 1.6, 0.2],
[4.6, 3.2, 1.4, 0.2],
[5.3, 3.7, 1.5, 0.2],
[5. , 3.3, 1.4, 0.2],
[7. , 3.2, 4.7, 1.4],
[6.4, 3.2, 4.5, 1.5],
[6.9, 3.1, 4.9, 1.5],
[5.5, 2.3, 4. , 1.3],
[6.5, 2.8, 4.6, 1.5],
[5.7, 2.8, 4.5, 1.3],
[6.3, 3.3, 4.7, 1.6],
[4.9, 2.4, 3.3, 1.],
[6.6, 2.9, 4.6, 1.3]

