# PE29 Vasu Kalariya

# **IMLA Lab Assi 2 Data Preprocessing**

In [46]: import pandas as pd import numpy as np import matplotlib.pyplot as plt import seaborn as sns

Out[47]:

	_id	Area_of_Origin	Building_Status	Business_Impact	Civilian_Casualties	Count_of_Persons_Rescued	Estimated_Dollar_Loss
0	190306	81 - Engine Area	NaN	NaN	0	0	1000.0
1	190307	22 - Sleeping Area or Bedroom (inc. patients r	01 - Normal (no change)	8 - Not applicable (not a business)	0	0	20.0
2	190308	81 - Engine Area	NaN	NaN	0	0	5000.0
3	190309	99 - Undetermined (formerly 98)	NaN	NaN	0	0	7500.0
4	190310	81 - Engine Area	NaN	NaN	0	0	10000.0
12682	202988	81 - Engine Area	NaN	NaN	0	0	500.0
12683	202989	81 - Engine Area	NaN	NaN	0	0	40000.0
12684	202990	27 - Laundry Area	01 - Normal (no change)	1 - No business interruption	0	0	2000.0
12685	202991	21 - Living Area (e.g. living, TV, recreation,	01 - Normal (no change)	8 - Not applicable (not a business)	1	0	25.0
12686	202992	97 - Other - unclassified	08 - Not Applicable	8 - Not applicable (not a business)	1	0	10.0

In [48]:	ds.isnull().sum()			
Out[48]:	_id	0		
	Area_of_Origin	0		
	Building_Status			
	Business_Impact	3484		
	Civilian_Casualties			
	Count_of_Persons_Rescued	0 1		
	Estimated_Dollar_Loss			
	Estimated_Number_Of_Persons_Displaced	3483 12408		
	Exposures			
	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		
	<pre>Initial_CAD_Event_Type</pre>	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 3484		
	Smoke_Alarm_at_Fire_Origin_Alarm_Type			
	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		

0

0

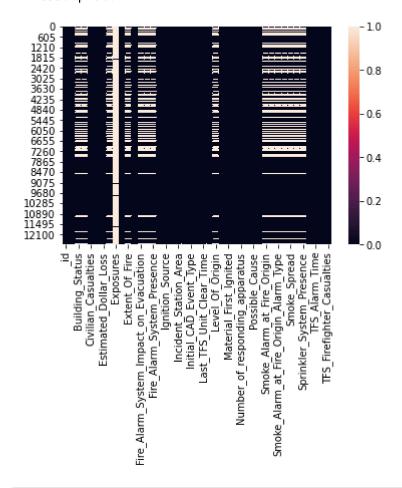
```
TFS_Arrival_Time
TFS_Firefighter_Casualties
dtype: int64
```

In [49]: ds.shape

Out[49]: (12687, 43)

In [50]: sns.heatmap(ds.isnull())

Out[50]: <AxesSubplot:>



In [51]: ds.dropna(subset=['Building\_Status'], inplace = True)

In [52]:	<pre>ds.isnull().sum()</pre>	
Out[52]:	_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 Incident Number	0 0
	Incident_Namber Incident_Station_Area	0
	Incident Ward	30
	Initial_CAD_Event_Type	9
	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
```

# **Imputer for Missing values**

```
In [55]: 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
```

#### Out[55]:

	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

```
In [56]: si2 = SimpleImputer(missing_values=np.nan,strategy='median')
    si2.fit(Exposures2)
    Exposures2 = si2.transform(Exposures2)
    Exposures2 = pd.DataFrame(Exposures2, columns=['Exposures2'])
    Exposures2
```

#### Out[56]:

	Exposures2
0	1.0
1	1.0
2	1.0
3	1.0
4	1.0
9200	1.0
9201	1.0
9202	1.0
9203	1.0
9204	1.0

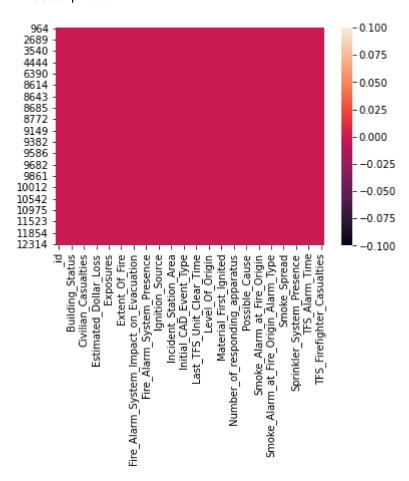
Out[57]: _id	0
Area_of_Origin	0
Building_Status	0
Business_Impact	2
Civilian_Casualties	0
Count_of_Persons_Rescued	0
Estimated_Dollar_Loss	0
Estimated_Number_Of_Persons_Displaced	1
Exposures	8994
Ext_agent_app_or_defer_time	0
Extent_Of_Fire	2
Final_Incident_Type	0
Fire_Alarm_System_Impact_on_Evacuation	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
<pre>Initial_CAD_Event_Type Intersection</pre>	0
Last_TFS_Unit_Clear_Time	0 0
Last_TF3_ONIt_Clear_Time 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
```

```
In [58]: ds.dropna(axis=0, how="any", thresh=None, subset=None, inplace=True)
```

```
In [59]: sns.heatmap(ds.isnull())
```

Out[59]: <AxesSubplot:>



```
In [60]: ds = ds.reset_index()
```

```
In [61]: ds['Estimated_Dollar_Loss'] = ds['Estimated_Dollar_Loss'].astype('int64')
         ds['Estimated_Dollar_Loss'].unique()
Out[61]: array([
                    2000,
                              200,
                                       5000,
                                               70000,
                                                       100000,
                                                                   7500,
                                                                            10000,
                     400,
                            90000,
                                       1000,
                                               20000,
                                                          2500,
                                                                  25000,
                                                                            50000,
                                                                   9000, 1000000,
                   30000,
                            40000,
                                     350000,
                                                  500,
                                                         60000,
                                                       250000,
                   80000,
                           500000,
                                     400000, 50000000,
                                                                 200000, 750000,
                                                        300000,
                                                                   9999, 2000000,
                  800000,
                           125000,
                                     700000,
                                               34000,
                   75000, 1500000,
                                     150000,
                                                4000.
                                                         15000,
                                                                   7000,
                                                                              250,
                  120000,
                                20,
                                     175000,
                                               18000,
                                                         65000,
                                                                  12000,
                                                                            35000,
                           600000,
                                                                   8000], dtype=int64)
                   49800,
                                      55000,
                                                 100, 1850000,
```

### Normalisation using Max Min, L1, L2, Zscore

```
In [62]: 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
```

#### Out[62]:

Estimated_Dollar_Loss
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0
0.0

```
In [63]: 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
```

#### Out[63]:

	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

```
In [64]: Estimated_Dollar_Loss3 = ds.iloc[:, 6:7].values

from sklearn.preprocessing import Normalizer
Data_normalizer= Normalizer(norm='12').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
```

#### Out[64]:

	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

```
In [65]: 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_Estimated_Number_Of_Persons_Displaced]
```

#### Out[65]:

	Estimated_Number_Of_Persons_Displaced
0	-0.467214
1	-0.471048
2	-0.460825
3	-0.322387
4	-0.258493
206	-0.301089
207	-0.466149
208	-0.454435
209	-0.450176
210	-0.450176

211 rows × 1 columns

### **Feature Selection using Correlation Coefficient**

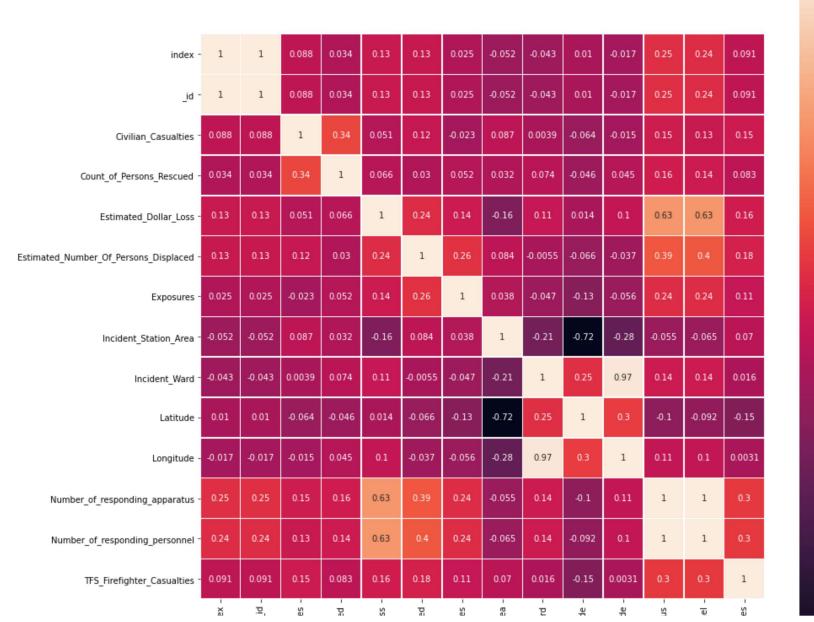
In [66]: ds.corr(method ='pearson')

Out[66]:

	index	_id	Civilian_Casualties	Count_of_Persons_Rescued	Estimated_Dollar_Loss	Est
index	1.000000	1.000000	0.088213	0.034237	0.128786	
_id	1.000000	1.000000	0.088213	0.034237	0.128786	
Civilian_Casualties	0.088213	0.088213	1.000000	0.336775	0.050592	
Count_of_Persons_Rescued	0.034237	0.034237	0.336775	1.000000	0.066351	
Estimated_Dollar_Loss	0.128786	0.128786	0.050592	0.066351	1.000000	
Estimated_Number_Of_Persons_Displaced	0.125300	0.125300	0.120813	0.030255	0.236972	
Exposures	0.025089	0.025089	-0.022841	0.051705	0.135748	
Incident_Station_Area	-0.051774	-0.051774	0.087232	0.032444	-0.160756	
Incident_Ward	-0.043413	-0.043413	0.003947	0.074149	0.109913	
Latitude	0.010110	0.010110	-0.064315	-0.046431	0.013641	
Longitude	-0.017260	-0.017260	-0.014582	0.044854	0.103475	
Number_of_responding_apparatus	0.249495	0.249495	0.152842	0.163182	0.629676	
Number_of_responding_personnel	0.239354	0.239354	0.134114	0.144415	0.630974	
TFS_Firefighter_Casualties	0.091333	0.091333	0.147323	0.083274	0.162295	

```
In [67]: plt.subplots(figsize=(15,15))
sns.heatmap(ds.corr(), annot = True,annot_kws={'size': 10},linewidths=.5,square=True)
```

Out[67]: <AxesSubplot:>



-1.0

- 0.8

- 0.6

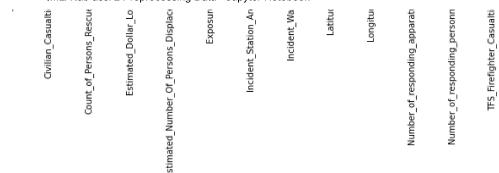
- 0.4

- 0.2

- 0.0

- -0.2

- -0.4



# **Encoding of categorical values**

```
In [68]: 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-vec tor y was passed when a 1d array was expected. Please change the shape of y to (n\_samples, ), for example usin g ravel().

return f(\*\*kwargs)

#### Out[68]:

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

In [69]: Method\_Of\_Fire\_Control = pd.get\_dummies(ds['Method\_Of\_Fire\_Control'])
Method\_Of\_Fire\_Control

Out[69]:

	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
3	1	0	0	0
4	1	0	0	0
206	1	0	0	0
207	1	0	0	0
208	1	0	0	0
209	1	0	0	0
210	1	0	0	0

```
In [70]: 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
```

1.0

(193, 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

```
IMLA lab assi 2 Preprocessing Data - Jupyter Notebook
In [71]:
         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, 3)
                          1
            (1, 4)
            (1, 6)
            (2, 3)
            (2, 1)
                          1
            (2, 4)
            (2, 2)
            (3, 3)
```

1

1

(204, 2) (205, 3) (205, 1) (205, 4) (205, 2) (206, 3)

(206, 1)

(3, 1) (3, 4) (3, 2) (4, 3) (4, 1) (4, 4) (4, 2) (5, 3) (5, 1) (5, 4) (5, 2) (6, 3) (6, 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**

```
In [77]: 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]

```
In [78]: from sklearn.feature_selection import VarianceThreshold
         selector = VarianceThreshold()
         selector.fit_transform(X,y)
Out[78]: 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],
```