

In [144...]

```
test_data = test_value.copy
test_data = test_data.set_i
test_data.head()
```

Out[144...]

	geo_level_1_id	geo_level_2_id	building_id
	300051	17	
	99355	6	
	890251	22	
	745817	26	
	421793	17	

In [145...]

```
test_data_num = test_data.s
test_data_cat = test_data.s
```

In [146...]

```
test_data_num = test_data_n
```

In [147...]

```
test_data_cat = pd.get_dumm
```

In [148...]

```
new_test_data = pd.concat([
```

In [149...]

```
scaler = StandardScaler()
scaled_test = new_test_data
data_train = df_new.copy()
scaled_col = ['geo_level_1_
              'count_floors,
              'height_perce

scaled_train = data_train[s
scaler.fit(scaled_train)
feature_scaled_test = scale
feature_scaled_test = scale
scaled_test[scaled_col] = f
data_test = scaled_test.cop
```

In [150...]

```
data_test.head()
```

0.30	0.39	4979
	2	0.60
0.88	0.71	29703
	3	0.56
0.20	0.29	17439

accuracy		
0.59	52121	
macro avg		
0.46	0.47	52121
weighted avg		
0.59	0.54	52121



Train F1 Score (Micro) for Logistic Regression Algorithm is : 0.591514773599386

Test F1 Score (Micro) for Logistic Regression Algorithm is : 0.5943861399435928

Decision Tree

In [124...

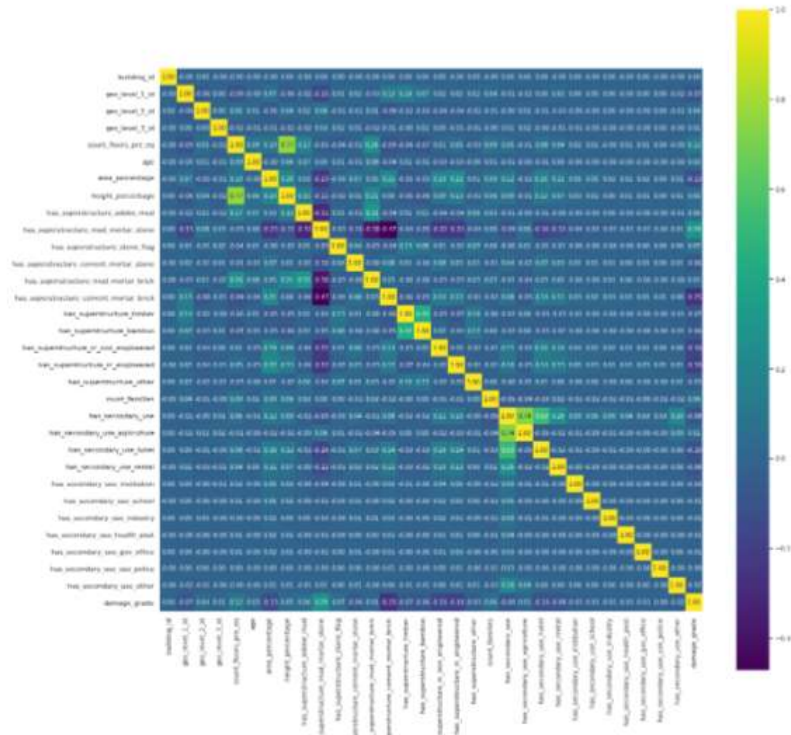
```
model_2 = DecisionTreeClass
model_2.fit(X_train, y_train)
pred_2 = model_2.predict(X_test)
train_pred_2 = model_2.predict(X_train)

print(classification_report(X_train, train_pred_2))
#make_confusion_matrix(pred_2, y_test)
train_score_2 = f1_score(y_train, train_pred_2)
score_2 = f1_score(y_test, pred_2)
print('Train F1 Score (Micro) :', train_score_2)
print('Test F1 Score (Micro) :', score_2)

with open('model_2', 'wb') as f:
    pickle.dump(model_2, f)
```

precision re

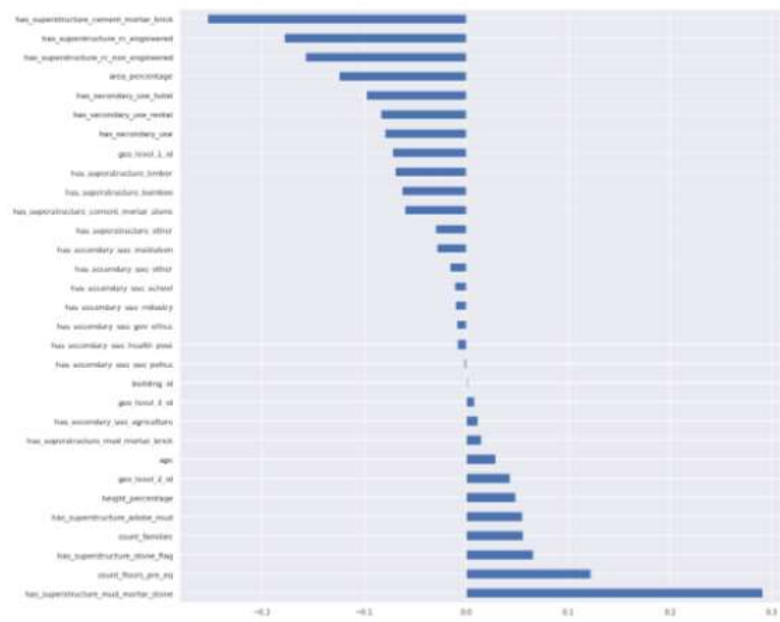
```
plt.show()
```



In [84]:

```
plt.figure(figsize=(16,16))

corr_matrix['damage_grade']
plt.show()
```



EDA (Recommendation)

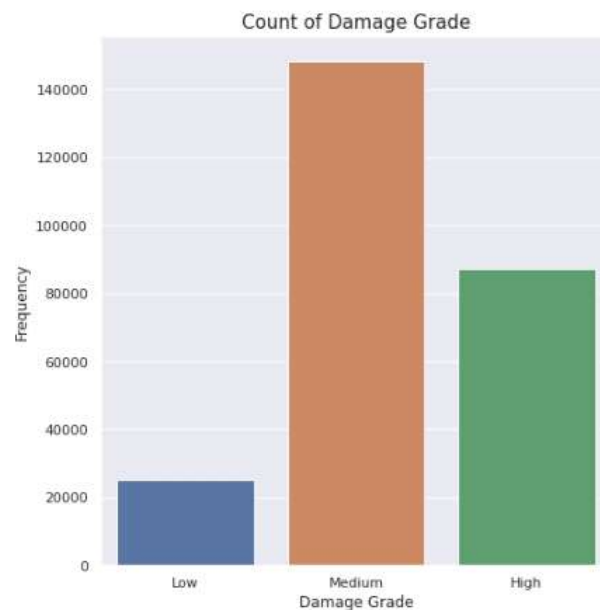
```
In [20]: train['damage_grade'] = tra
train['damage_grade'] = tra
```

```
In [21]: train['damage_grade'].unique
```

```
Out[21]: array(['High', 'Medium', 'Low'], dtype=object)
```

```
In [22]: plt.figure(figsize=(7,7))

sns.set_theme()
sns.countplot(train['damage_grade'],
plt.xlabel('Damage Grade',
plt.ylabel('Frequency', for
plt.title('Count of Damage
plt.tight_layout()
plt.show()
```



```
In [23]: print('Jumlah Bangunan deng
print('Jumlah Bangunan deng
print('Jumlah Bangunan deng
print('')
print('Persentase Bangunan
print('Persentase Bangunan
print('Persentase Bangunan
```

Jumlah Bangunan dengan tingka
t kerusakan rendah : 25124

Jumlah Bangunan dengan tingka

```
In [12]: train.isnull().sum()
```

```
Out[12]: building_id
0
geo_level_1_id
0
geo_level_2_id
0
geo_level_3_id
0
count_floors_pre_eq
0
age
0
area_percentage
0
height_percentage
0
land_surface_condition
0
foundation_type
0
roof_type
0
ground_floor_type
0
other_floor_type
0
position
0
plan_configuration
0
has_superstructure_adobe_mu
d                0
has_superstructure_mud_mort
ar_stone        0
has_superstructure_stone_fl
ag              0
has_superstructure_cement_m
ortar_stone     0
has_superstructure_mud_mort
ar_brick        0
has_superstructure_cement_m
ortar_brick     0
has_superstructure_timber
0
has_superstructure_bamboo
0
```



```
In [7]: train_label.head()
```

```
Out[7]:
```

	building_id	damage_grade
0	802906	3
1	28830	2
2	94947	3
3	590882	2
4	201944	3

```
In [8]: train_value.head()
```

```
Out[8]:
```

	building_id	geo_level_1_id	geo.
0	802906	6	
1	28830	8	
2	94947	21	
3	590882	22	
4	201944	11	

```
In [9]: test_value.head()
```

```
Out[9]:
```

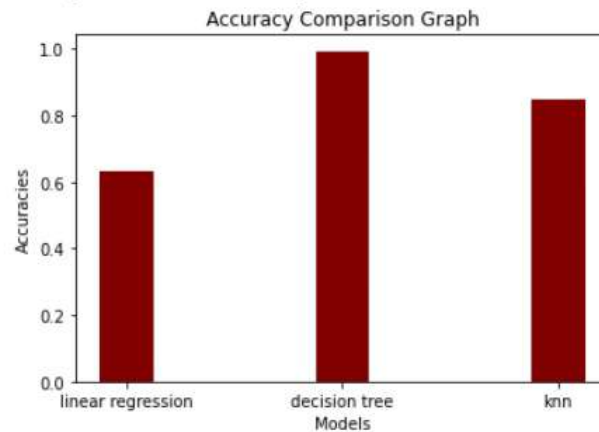
	building_id	geo_level_1_id	geo.
0	300051	17	
1	99355	6	
2	890251	22	
3	745817	26	
4	421793	17	

Description Data

Penjelasan Label

```
width = 0.25)  
plt.xlabel("Models")  
plt.ylabel("Accuracies")  
plt.title("Accuracy Compar
```

Out[455... Text(0.5, 1.0, 'Accuracy Co
mparison Graph')

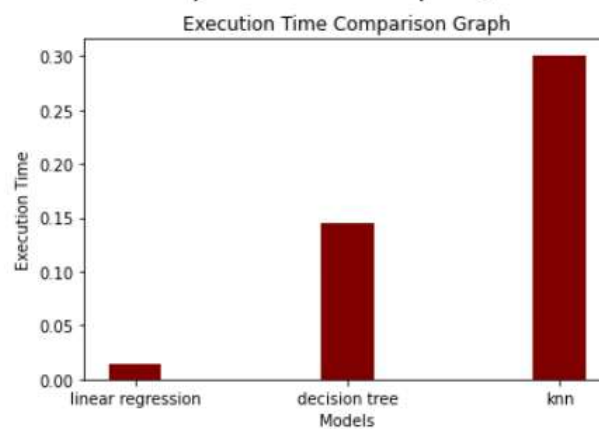


2. Execution Time

In [456...

```
times = [t1,t2,t3]  
plt.bar(models, times, color  
width = 0.25)  
plt.xlabel("Models")  
plt.ylabel("Execution Time  
plt.title("Execution Time
```

Out[456... Text(0.5, 1.0, 'Execution T
ime Comparison Graph')



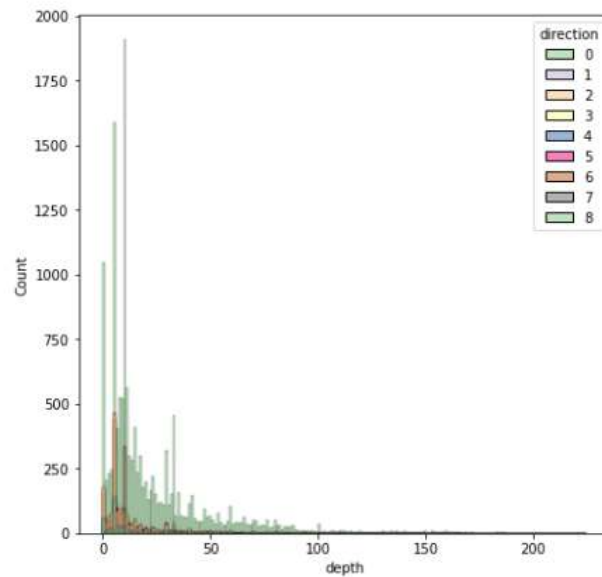
In []:



```
px.scatter(df, x='richter'
```

In [412...

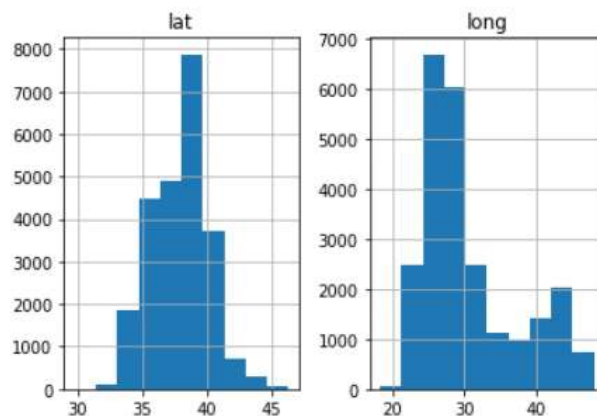
```
plt.figure(figsize=(7,7))
sns.histplot(data=df, x='di
plt.show()
```



In [413...

```
plt.figure(figsize=(7,7))
df[['lat', 'long']].hist()
plt.show()
```

<Figure size 504x504 with 0 Axes>



In [414...

```
plt.figure(figsize=(15,10))
sns.countplot(df.xm)
```



```
padding: 5px;
position: absolute;
top: 10px;
z-index: 5;
}
</style>
</head>

<body>
  <table border=0 cellpadding="1" style=
    <tr>
      <td><h1><p style="text-align:cen
    </tr>
  </table>

  <div id="map"></div>
  <script>

    var map, heatmap;

    function initMap() {
      map = new google.maps.Map(document.get
        zoom: 1.5,
        center: {lat: 0, lng: 0},
        mapTypeId: 'roadmap'
      });

      heatmap = new google.maps.visualization
        data: getPoints(),
        map: map
      });
    }

    function toggleHeatmap() {
      heatmap.setMap(heatmap.getMap() ? null
    }

    function changeGradient() {
      var gradient = [
        'rgba(0, 255, 255, 0.3)',
```


Out[24]:

	date	place	latitude
75	2020-08-18	Hawaii	19.282855
76	2020-08-18	Indonesia	-4.186166
77	2020-08-18	Nevada	38.177047
78	2020-08-18	Philippines	7.507142
79	2020-08-18	Puerto Rico	17.974638

In [25]:

```
import datetime as dt
# convert date to proper fo
days = list(set([d for d in
days.sort()

# Predict NaN outcome value
predict_day=days[2]
predict_day
```

Out[25]: '2020-08-14'

In [26]:

```
# place, date, lat and long
for i in range(0,7):
    live_set_tmp = live_set
    plt.scatter(live_set_tm
    plt.suptitle('Future Ea
    plt.xlabel('Longitude')
    plt.ylabel('Latitude')
    plt.grid()
    plt.show()
```

Future Earthquakes for 2020-08-12



In [12]:

```

from sklearn.metrics import
from sklearn.metrics import
from sklearn.metrics import
from sklearn.metrics import
from sklearn.metrics import

print(roc_auc_score(y_test,

fpr, tpr, _ = roc_curve(y_t
roc_auc = auc(fpr, tpr)
print('AUC:', np.round(roc_

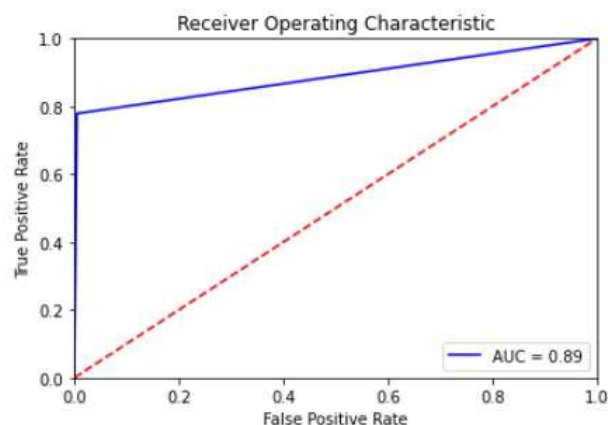
plt.title('Receiver Operati
plt.plot(fpr, tpr, 'b', lab
plt.legend(loc = 'lower rig
plt.plot([0, 1], [0, 1], 'r-
plt.xlim([0, 1])
plt.ylim([0, 1])
plt.ylabel('True Positive R
plt.xlabel('False Positive
plt.show()

print("Confusion Matrix: \n
print("\nRecall 'TP/TP+FN'

```

0.8869826795052952

AUC: 0.887



Confusion Matrix:

```

[[3595  18]
 [ 42 148]]

```

Recall 'TP/TP+FN' = 0.778947
3684210526

Adaboost Random Forest
Classifier

In [3]:

```
df_features.head()
```

Out[3]:

	date	depth	mag	place
0	2020-07-27	76.37	4.2	Papua New Guinea
1	2020-07-28	88.76	4.3	Papua New Guinea
2	2020-07-29	83.01	4.4	Papua New Guinea
3	2020-07-30	10.00	4.9	Papua New Guinea
4	2020-07-31	195.88	4.3	Papua New Guinea

In [4]:

```
engine = create_engine('sqli
df_predict = pd.read_sql_ta
```

In [5]:

```
# Live data to be predicted
#Hence NaN outcome that has

df_predict.head()
```

Out[5]:

	date	depth	mag	place
0	2020-08-03	140.81	4.4	Papua New Guinea
1	2020-08-04	172.48	5.4	Papua New Guinea
2	2020-08-05	94.65	5.0	Papua New Guinea
	2020-			Papua


```
    box-sizing: content-box;
}
input[type="search"]::-webkit-search-cancel-button,
input[type="search"]::-webkit-search-decoration,
input[type="search"]::-webkit-search-results-button,
input[type="search"]::-webkit-search-results-decoration {
    -webkit-appearance: none;
}
fieldset {
    border: 1px solid #c0c0c0;
    margin: 0 2px;
    padding: 0.35em 0.625em 0.75em;
}
legend {
    border: 0;
    padding: 0;
}
textarea {
    overflow: auto;
}
optgroup {
    font-weight: bold;
}
table {
    border-collapse: collapse;
    border-spacing: 0;
}
td,
th {
    padding: 0;
}
/*! Source: https://github.com/h5bp/html5-boilerplate/blob/master/src/css/main.css */
@media print {
    *,
    *:before,
    *:after {
        background: transparent !important;
        box-shadow: none !important;
        text-shadow: none !important;
    }
    a,
    a:visited {
```

```
    margin: 1em 40px;
  }
  hr {
    box-sizing: content-box;
    height: 0;
  }
  pre {
    overflow: auto;
  }
  code,
  kbd,
  pre,
  samp {
    font-family: monospace, monospace;
    font-size: 1em;
  }
  button,
  input,
  optgroup,
  select,
  textarea {
    color: inherit;
    font: inherit;
    margin: 0;
  }
  button {
    overflow: visible;
  }
  button,
  select {
    text-transform: none;
  }
  button,
  html input[type="button"],
  input[type="reset"],
  input[type="submit"] {
    -webkit-appearance: button;
    cursor: pointer;
  }
  button[disabled],
```

```
}
audio,
canvas,
progress,
video {
    display: inline-block;
    vertical-align: baseline;
}
audio:not([controls]) {
    display: none;
    height: 0;
}
[hidden],
template {
    display: none;
}
a {
    background-color: transparent;
}
a:active,
a:hover {
    outline: 0;
}
abbr[title] {
    border-bottom: 1px dotted;
}
b,
strong {
    font-weight: bold;
}
dfn {
    font-style: italic;
}
h1 {
    font-size: 2em;
    margin: 0.67em 0;
}
mark {
    background: #ff0;
    color: #000;
```

Code

Blame



```
<!DOCTYPE html>
<html>
<head><meta charset="utf-8" />

<title>ETL_USGS_EarthQuake</title>

<script src="https://cdnjs.cloudflare.com/a
<script src="https://cdnjs.cloudflare.com/a

<style type="text/css">
    /*!
    *
    * Twitter Bootstrap
    *
    */
    /*!
    * Bootstrap v3.3.7 (http://getbootstrap.co
    * Copyright 2011-2016 Twitter, Inc.
    * Licensed under MIT (https://github.com/t
    */
    /*! normalize.css v3.0.3 | MIT License | gi
html {
    font-family: sans-serif;
    -ms-text-size-adjust: 100%;
    -webkit-text-size-adjust: 100%;
}
body {
    margin: 0;
}
article,
aside,
details,
figcaption,
figure,
footer
```