FinalProject_WeatherData

April 17, 2022

[1]:
```
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import numpy as np
%matplotlib inline
```

[2]:
```
#import weather data
weather = pd.read_csv('canada_weather.csv')
weather.head()
```

[2]:

	Community	Weather station	Location	Elevation	January(Avg. high °C (°F))	January(Avg. low °C (°F))
0	Alberton, PE	NaN	46°51 00 N 064°01 00 W / 46.85000°N 64.01667°W…	3m (9.8ft)	-3.9 (25.0)	-12.5 (9.5)
1	Baker Lake, NU	YBK	64°17 56 N 096°04 40 W / 64.29889°N 96.07778°W…	18.6m (61ft)	-27.7 (-17.9)	-34.8 (-30.6)
2	Baie-Comeau, QC	YBC	49°08 00 N 068°12 00 W / 49.13333°N 68.20000°W…	22m (72ft)	-8.7 (16.3)	-19.9 (-3.8)
3	Calgary, AB	YYC	51°06 50 N 114°01 13 W / 51.11389°N 114.02028°…	1,084m (3,556ft)	-0.9 (30.4)	-13.2 (8.2)
4	Charlottetown, PE	YYG	46°17 19 N 063°07 43 W / 46.28861°N 63.12861°W…	49m (161ft)	-3.4 (25.9)	-12.1 (10.2)

	July(Avg. high °C (°F))	July(Avg. low °C (°F))	Annual(Avg. high °C (°F))
0	23.2 (73.8)	14.1 (57.4)	9.6 (49.3)
1	17.0 (62.6)	6.1 (43.0)	-7.3 (18.9)
2	20.9 (69.6)	10.3 (50.5)	6.6 (43.9)
3	23.2 (73.8)	9.8 (49.6)	10.8 (51.4)
4	23.3 (73.9)	14.1 (57.4)	9.9 (49.8)

```
  Annual(Avg. low °C (°F))
0               1.3 (34.3)
1             -15.2 (4.6)
2             -3.3 (26.1)
3             -1.9 (28.6)
4               1.3 (34.3)
```

[3]:
```
#split location and get latitude and longitude
# and then separate out latitude
weather['Latitude']=weather['Location'].apply(lambda x : x.split('/')[2]).
 ↪apply(lambda x : x.split(';')[0])
print(weather['Latitude'].head())
```

```
0     46.85000
1     64.29889
2     49.13333
3     51.11389
4     46.28861
Name: Latitude, dtype: object
```

[4]:
```
#split location and get latitude and longitude
# and then separate out latitude
weather['Longitude']=weather['Location'].apply(lambda x : x.split('/')[2]).
 ↪apply(lambda x : x.split(';')[1])
print(weather['Longitude'].head())
```

```
0               -64.01667 (Alberton)
1     -96.07778 (Baker Lake Airport)
2            -68.20000 (Baie-Comeau)
3               -114.02028 (Calgary)
4          -63.12861 (Charlottetown)
Name: Longitude, dtype: object
```

[5]:
```
#separate just the longitude
weather['Longitude']=weather['Longitude'].apply(lambda x : x.split(' (')[0].
 ↪strip())
```

[6]:
```
#converting to integer from string
weather['Longitude']=[float(long) for long in weather['Longitude']]
```

[7]:
```
#taking just the °C value for annual avg low
weather['Annual(Avg. low °C (°F))']=weather['Annual(Avg. low °C (°F))'].
 ↪apply(lambda x : x.split(' (')[0])
```

[8]:
```
#taking just the °C value for annual avg low
weather['Annual(Avg. high °C (°F))']=weather['Annual(Avg. high °C (°F))'].
 ↪apply(lambda x : x.split(' (')[0])
```

[9]:
```
#taking elevation in meters
weather['Elevation']=weather['Elevation'].apply(lambda x : x.split(' (')[0])
```

[10]:
```
#removing meter unit from elevation
weather['Elevation']=weather['Elevation'].str.replace("m","")
```

[11]:
```
#converting to integer from string
import re
weather['Elevation']=[int(float(re.sub(",", "", e))) for e in␣
 ↪weather['Elevation']]
```

[12]:
```
# renaming the column by removing °F
weather.rename(columns = {'Annual(Avg. low °C (°F))':'Annual(Avg. low °C)'},␣
 ↪inplace=True)
weather.rename(columns = {'Annual(Avg. high °C (°F))':'Annual(Avg. high °C)'},␣
 ↪inplace=True)
```

[13]:
```
weather['Annual(Avg. high °C)']=[float(re.sub("−", "-", high)) for high in␣
 ↪weather['Annual(Avg. high °C)']]
```

[14]:
```
weather['Annual(Avg. low °C)']=[float(re.sub("−", "-", low)) for low in␣
 ↪weather['Annual(Avg. low °C)']]
```

[15]:
```
# adding unit to the column Elevation
weather.rename(columns = {'Elevation':'Elevation(m)'}, inplace=True)
```

[16]:
```
#dropping unwanted columns
weather.drop(['Community','Weather station','Location','January(Avg. high °C␣
 ↪(°F))','January(Avg. low °C (°F))','July(Avg. high °C (°F))','July(Avg. low␣
 ↪°C (°F))'], axis=1, inplace=True)
```

[17]:
```
#rearranging columns
weather = weather[['Latitude','Longitude','Elevation(m)','Annual(Avg. high␣
 ↪°C)','Annual(Avg. low °C)']]
```

[18]:
```
weather.head()
```

[18]:

	Latitude	Longitude	Elevation(m)	Annual(Avg. high °C)	\
0	46.85000	-64.01667	3	9.6	
1	64.29889	-96.07778	18	-7.3	
2	49.13333	-68.20000	22	6.6	
3	51.11389	-114.02028	1084	10.8	
4	46.28861	-63.12861	49	9.9	

	Annual(Avg. low °C)
0	1.3
1	-15.2

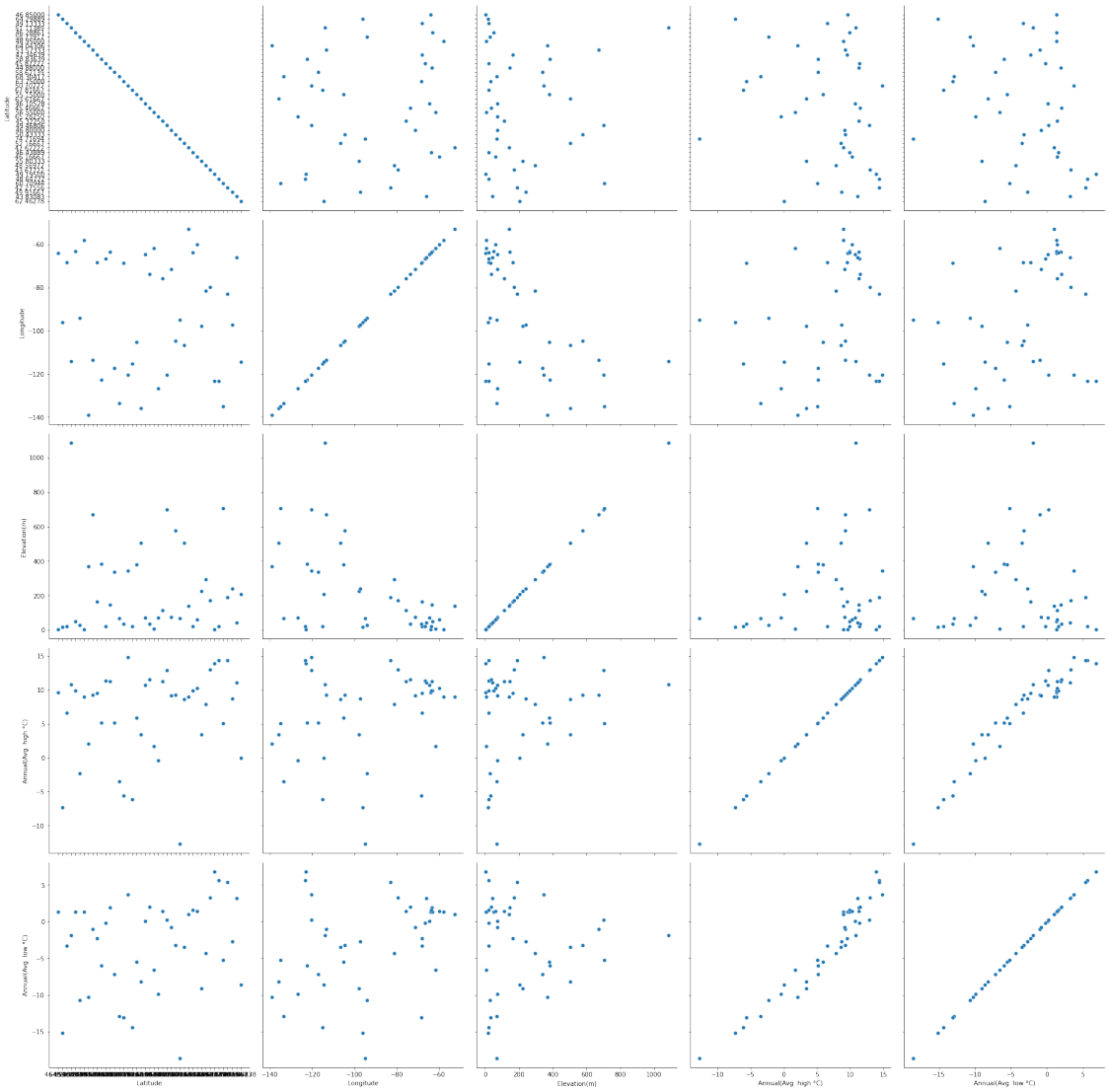
```
2                -3.3
3                -1.9
4                 1.3
```

[19]:
```
weather_subset=weather[['Elevation(m)','Annual(Avg. high °C)','Annual(Avg. low␣
↪°C)']]
print(weather_subset.head())
```

```
   Elevation(m)  Annual(Avg. high °C)  Annual(Avg. low °C)
0             3                   9.6                  1.3
1            18                  -7.3                -15.2
2            22                   6.6                 -3.3
3          1084                  10.8                 -1.9
4            49                   9.9                  1.3
```

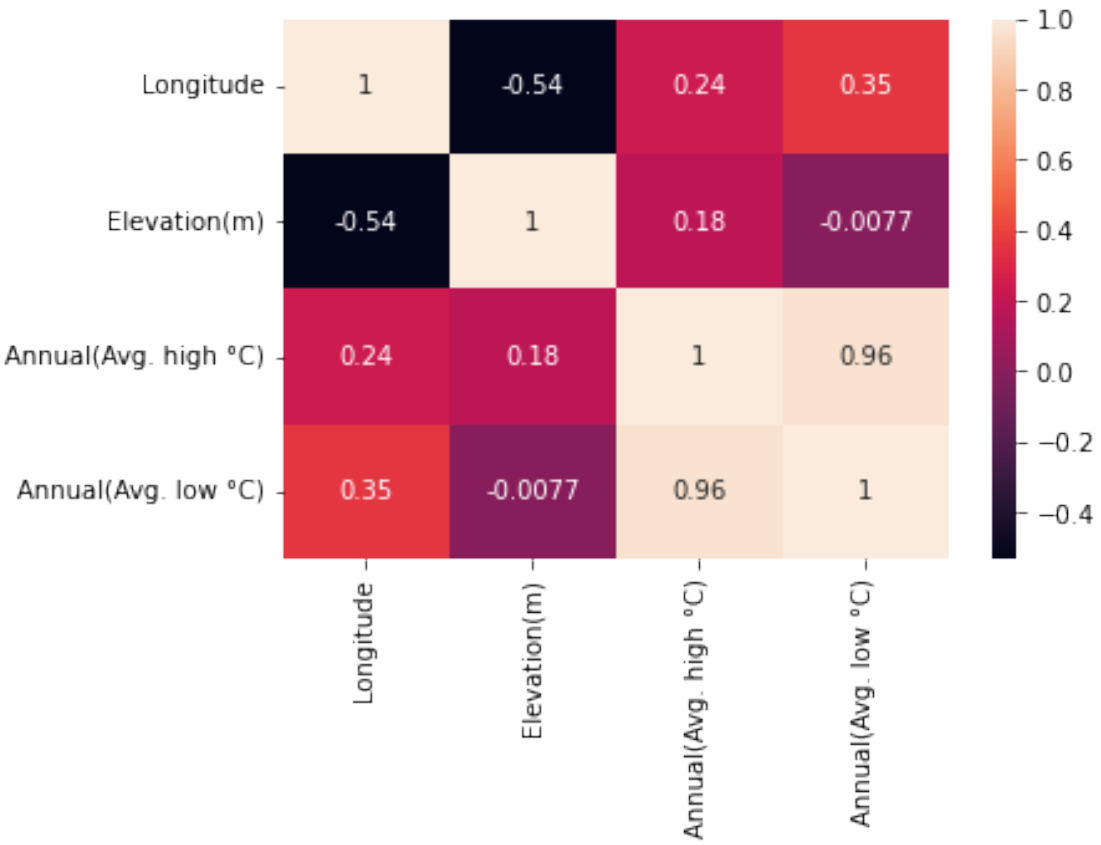
[20]:
```
x = sns.PairGrid(weather, vars=weather,  height = 5)
x.map(sns.scatterplot)
```

[20]:
```
<seaborn.axisgrid.PairGrid at 0x1e0584fad70>
```



[21]:
```
sns.heatmap(weather.corr(),annot=True)
```

[21]: <AxesSubplot:>



[22]:
```
X=weather[['Latitude','Longitude','Elevation(m)']].values
y=weather['Annual(Avg. high °C)'].values
print(X)
print(y)
```

```
[[' 46.85000' -64.01667 3]
 [' 64.29889' -96.07778 18]
 [' 49.13333' -68.2 22]
 [' 51.11389' -114.02028 1084]
 [' 46.28861' -63.12861 49]
 [' 58.73917' -94.06639 29]
 [' 48.95000' -57.95 5]
 [' 64.04306' -139.12778 370]
 [' 53.57333' -113.51833 671]
 [' 47.34639' -68.18778 163]
 [' 58.83639' -122.59722 382]
 [' 45.87222' -66.52778 21]
 [' 44.88000' -63.5 145]
 [' 58.62139' -117.16472 338]
 [' 68.30417' -133.48278 68]
 [' 63.75000' -68.55 34]
```

```
 [' 50.70222' -120.44194 345]
 [' 67.81667' -115.14389 23]
 [' 55.15000' -105.26667 379]
 [' 63.61667' -135.86667 504]
 [' 46.10528' -64.68389 71]
 [' 45.46667' -73.75 36]
 [' 56.55000' -61.68333 6]
 [' 65.28250' -126.80028 73]
 [' 45.32250' -75.66917 114]
 [' 49.46806' -120.51139 700]
 [' 46.80000' -71.38333 74]
 [' 50.43333' -104.66667 578]
 [' 74.71694' -94.96944 68]
 [' 52.16667' -106.71667 504]
 [' 47.62222' -52.74278 141]
 [' 46.43889' -63.83167 20]
 [' 46.16667' -60.04806 62]
 [' 55.80333' -97.8625 224]
 [' 48.56972' -81.37667 295]
 [' 43.67722' -79.63056 173]
 [' 49.19500' -123.18194 4]
 [' 48.64722' -123.42583 20]
 [' 60.70944' -135.06889 706]
 [' 42.27556' -82.95556 190]
 [' 49.91667' -97.23333 239]
 [' 43.83083' -66.08861 43]
 [' 62.46278' -114.44028 206]]
[  9.6  -7.3   6.6  10.8   9.9  -2.3   9.    2.1   9.3   9.5   5.2  11.4
  11.3   5.2  -3.5  -5.6  14.8  -6.1   5.9   3.4  10.7  11.5   1.7  -0.4
  11.3  12.9   9.2   9.3 -12.7   8.6   9.    9.9  10.3   3.4   7.9  13.
  13.9  14.4   5.1  14.4   8.7  11.1   0. ]
```

[23]:
```
from sklearn.model_selection import train_test_split
print(X.shape,y.shape)
```

```
(43, 3) (43,)
```

[24]:
```
X_train, X_test, y_train, y_test = train_test_split(X, y)
print(X_train.shape, X_test.shape, y_train.shape, y_test.shape)
```

```
(32, 3) (11, 3) (32,) (11,)
```

[25]:
```
#Polynomial Regression for Annual high temperature
X = weather[['Latitude','Longitude','Elevation(m)']]
y_high = weather['Annual(Avg. high °C)']

X_high_train, X_high_test, y_high_train, y_high_test = train_test_split(X,␣
 ↪y_high, test_size=0.2, random_state=101)
```

```
print(X_high_train.shape,X_high_test.shape,X.shape)
```

```
(34, 3) (9, 3) (43, 3)
```

[26]:
```
## builiding model
from sklearn.linear_model import LinearRegression
from sklearn.preprocessing import PolynomialFeatures
from sklearn.pipeline import make_pipeline
poly_model = make_pipeline(PolynomialFeatures(degree=3,include_bias=True),
                           LinearRegression())
# evaluation
poly_model.fit(X_high_train, y_high_train)
yfit_high = poly_model.predict(X_high_test)
yfit_high_training = poly_model.predict(X_high_train)

sns.scatterplot(X_high_train['Elevation(m)'],yfit_high_training,color='r') #␣
 ↪ground_truth
sns.scatterplot(X_high_test['Elevation(m)'],yfit_high,color='g') # prediction
```

```
C:\Users\priya\AppData\Local\Programs\Python\Python310\lib\site-
packages\seaborn\_decorators.py:36: FutureWarning: Pass the following variables
as keyword args: x, y. From version 0.12, the only valid positional argument
will be `data`, and passing other arguments without an explicit keyword will
result in an error or misinterpretation.
  warnings.warn(
C:\Users\priya\AppData\Local\Programs\Python\Python310\lib\site-
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  warnings.warn(
```
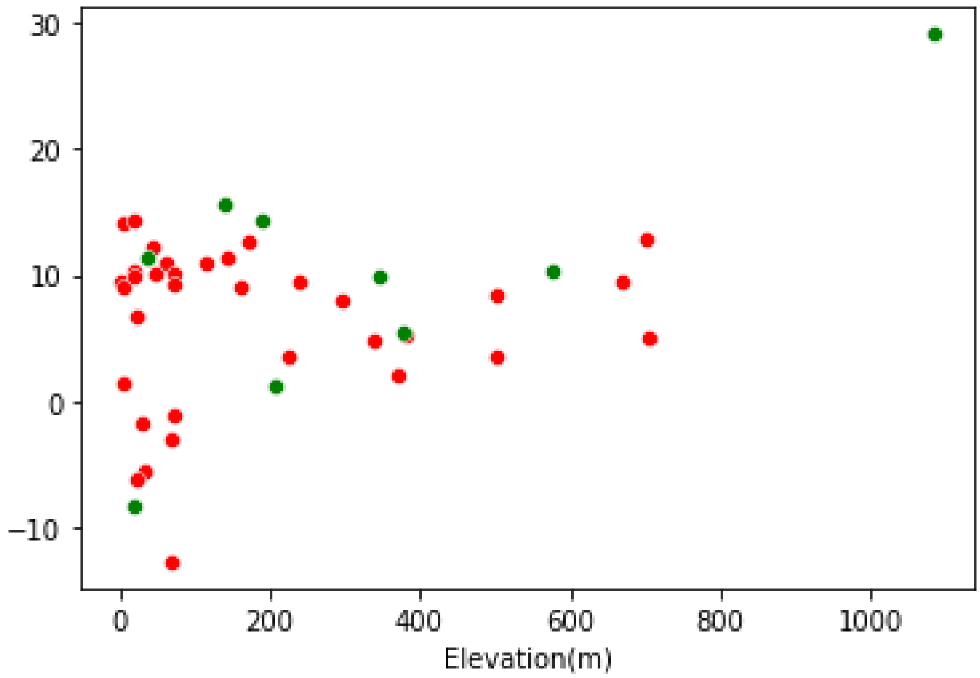

[26]:
```
<AxesSubplot:xlabel='Elevation(m)'>
```



[27]:
```
sns.lineplot(X.iloc[:,2],y,color='r')
sns.lineplot(X_high_train.iloc[:,2],yfit_high_training,color='g')
sns.scatterplot(X_high_test.iloc[:,2],yfit_high)

plt.legend(labels=['Original Function','Train Data','Test Data'])
```

```
C:\Users\priya\AppData\Local\Programs\Python\Python310\lib\site-
packages\seaborn\_decorators.py:36: FutureWarning: Pass the following variables
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  warnings.warn(
```
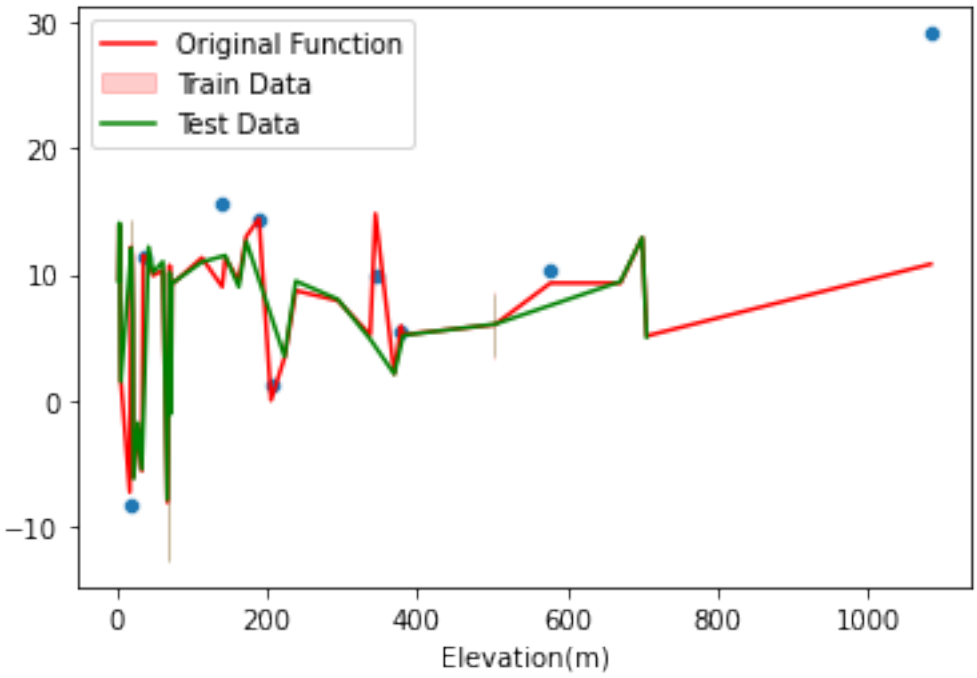
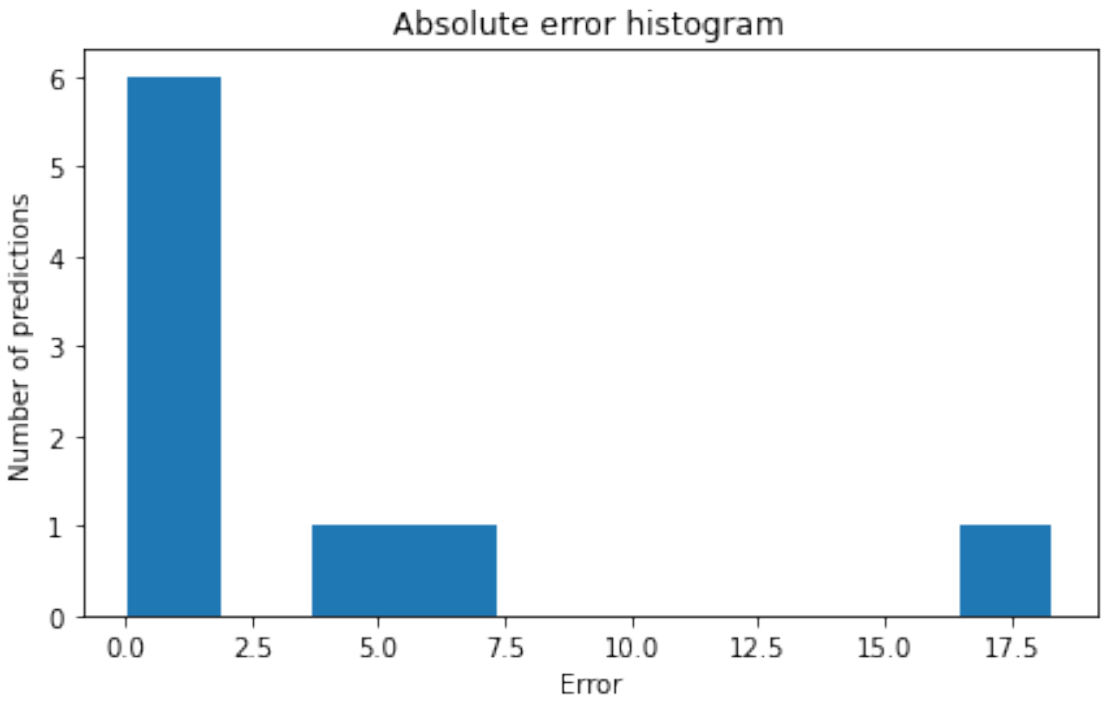

[27]: <matplotlib.legend.Legend at 0x1e05ccd64d0>



[28]:
```
from sklearn import metrics
print('MAE:', metrics.mean_absolute_error(y_high_test, yfit_high))
print('RMSE Test:', np.sqrt(metrics.mean_squared_error(y_high_test, yfit_high)))
print('RMSE Training:', np.sqrt(metrics.mean_squared_error(y_high_train,␣
 ↪yfit_high_training)))
```

```
MAE: 3.740632398582406
RMSE Test: 6.7079685692302675
RMSE Training: 0.4096556319574178
```

[29]:
```
absError=abs((yfit_high-y_high_test))
plt.figure(figsize=(7,4))
plt.xlabel("Error")
plt.ylabel("Number of predictions")
plt.title("Absolute error histogram")
plt.hist(absError)
plt.show()
```



[30]:
```
#Polynomial Regression for Annual low temperature
X = weather[['Latitude','Longitude','Elevation(m)']]
y_low = weather['Annual(Avg. low °C)']

X_low_train, X_low_test, y_low_train, y_low_test = train_test_split(X, y_low,␣
 ↪test_size=0.2, random_state=101)
print(X_low_train.shape,X_low_test.shape,X.shape)
```

```
(34, 3) (9, 3) (43, 3)
```

[31]:
```
## builiding model
from sklearn.preprocessing import PolynomialFeatures
from sklearn.pipeline import make_pipeline
poly_model = make_pipeline(PolynomialFeatures(degree=3,include_bias=True),
                           LinearRegression())
# evaluation
poly_model.fit(X_low_train, y_low_train)
yfit_low = poly_model.predict(X_low_test)
yfit_low_training = poly_model.predict(X_low_train)

sns.scatterplot(X_low_train['Elevation(m)'],yfit_low_training,color='r') #␣
 ↪ground_truth
sns.scatterplot(X_low_test['Elevation(m)'],yfit_low,color='g') # prediction
```

```
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```

```
packages\seaborn\_decorators.py:36: FutureWarning: Pass the following variables
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  warnings.warn(
```
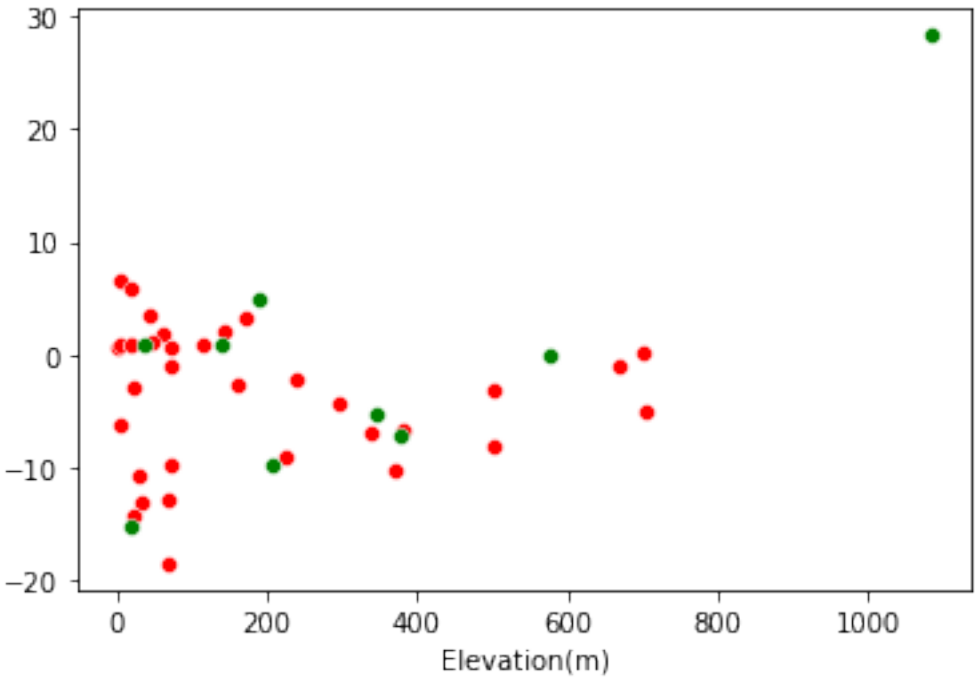

[31]: <AxesSubplot:xlabel='Elevation(m)'>



[32]:
```
sns.lineplot(X.iloc[:,2],y,color='r')
sns.lineplot(X_low_train.iloc[:,2],yfit_low_training,color='g')
sns.scatterplot(X_low_test.iloc[:,2],yfit_low)

plt.legend(labels=['Original Function','Train Data','Test Data'])
```

```
C:\Users\priya\AppData\Local\Programs\Python\Python310\lib\site-
packages\seaborn\_decorators.py:36: FutureWarning: Pass the following variables
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```

```
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result in an error or misinterpretation.
  warnings.warn(
```
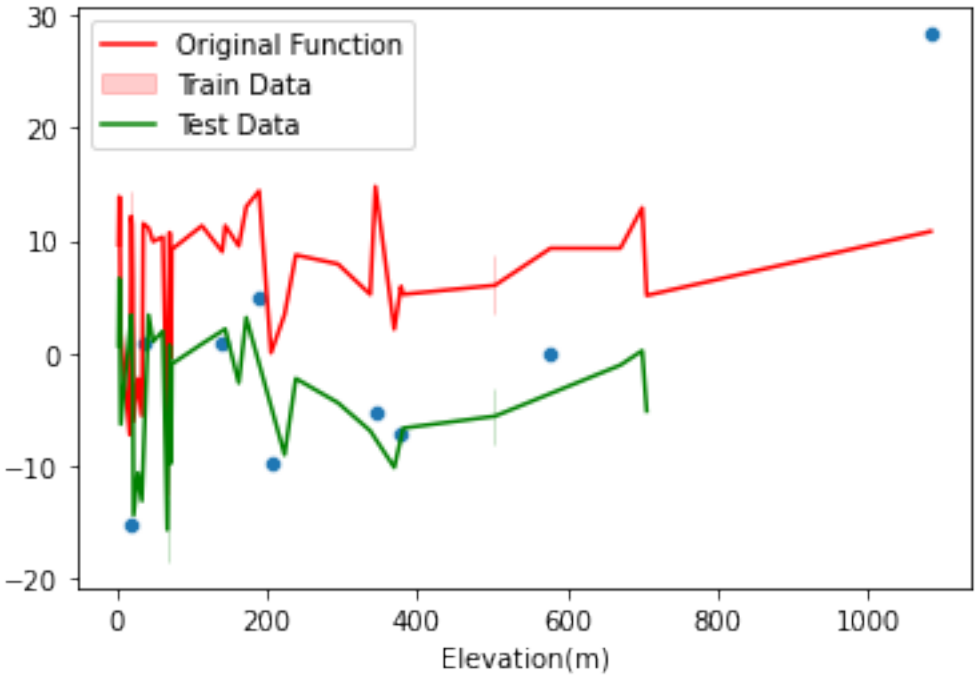
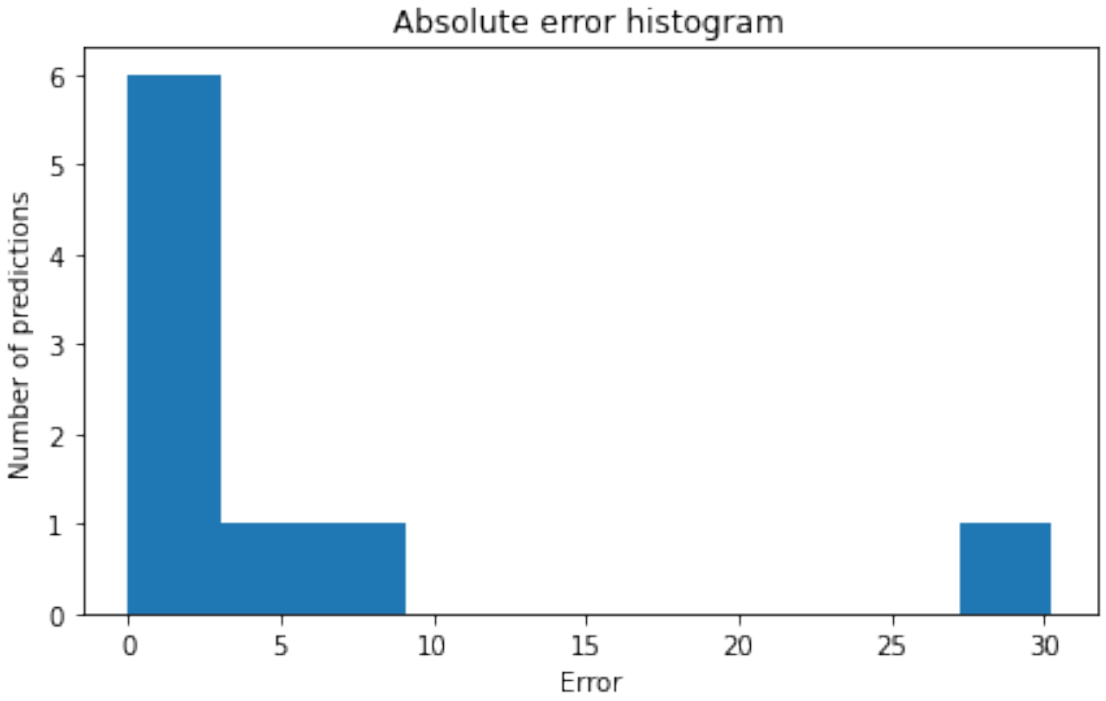

[32]: `<matplotlib.legend.Legend at 0x1e0603df100>`



[33]:
```
print('MAE:', metrics.mean_absolute_error(y_low_test, yfit_low))
print('RMSE Test:', np.sqrt(metrics.mean_squared_error(y_low_test, yfit_low)))
print('RMSE Training:', np.sqrt(metrics.mean_squared_error(y_low_train,␣
  ↪yfit_low_training)))
```

```
MAE: 5.244826449721522
RMSE Test: 10.600822745470317
RMSE Training: 0.38415976396258084
```

[34]:
```
absError1=abs((yfit_low-y_low_test))
plt.figure(figsize=(7,4))
plt.xlabel("Error")
plt.ylabel("Number of predictions")
plt.title("Absolute error histogram")
plt.hist(absError1)
plt.show()
```



[ ]:
```

```