# polynomialregression

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#### 1 Polynomial Regression

- Polynomial Regression is an extended version of linear regression where polynomial terms are introduced. When the regression line/good fit line isn't linear anymore and doesn't fit the data well(or when the data is non-linear), we opt for polynomial regression.
- Polynomial regression is also called as polynomial linear regression as still the target variable is in linear relationship with the coefficients.

Polynomial slope intercept formulae,  $Y=B0+B1X1+(B2X2)^{2+(B3X3)}3+(B4X4)^{4+\dots+(BNXN)}N$ 

```
[]: # Importing necessary libraries

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

```
[]: # Onboarding data

from google.colab import files
rawdata=files.upload()
```

<IPython.core.display.HTML object>

Saving Ice\_cream selling data.csv to Ice\_cream selling data (1).csv

```
[]: # DataFrame

df=pd.read_csv('Ice_cream selling data (1).csv')
df
```

```
[]:
         Temperature (°C)
                            Ice Cream Sales (units)
                 -4.662263
                                           41.842986
                 -4.316559
                                            34.661120
     1
     2
                 -4.213985
                                            39.383001
     3
                 -3.949661
                                            37.539845
     4
                 -3.578554
                                            32.284531
     5
                 -3.455712
                                           30.001138
     6
                 -3.108440
                                           22.635401
```

```
7
                 -3.081303
                                            25.365022
     8
                 -2.672461
                                            19.226970
     9
                 -2.652287
                                            20.279679
     10
                 -2.651498
                                            13.275828
     11
                 -2.288264
                                            18.123991
     12
                 -2.111870
                                            11.218294
     13
                 -1.818938
                                            10.012868
     14
                 -1.660348
                                            12.615181
                 -1.326379
     15
                                            10.957731
     16
                 -1.173123
                                             6.689123
     17
                 -0.773330
                                             9.392969
     18
                 -0.673753
                                             5.210163
     19
                 -0.149635
                                             4.673643
     20
                 -0.036156
                                             0.328626
     21
                 -0.033895
                                             0.897603
                  0.008608
     22
                                             3.165600
     23
                  0.149245
                                             1.931416
     24
                  0.688781
                                             2.576782
     25
                  0.693599
                                             4.625689
     26
                  0.874905
                                             0.789974
     27
                                             2.313806
                  1.024181
     28
                  1.240712
                                             1.292361
     29
                  1.359813
                                             0.953115
     30
                  1.740000
                                             3.782570
     31
                  1.850552
                                             4.857988
     32
                  1.999310
                                             8.943823
     33
                  2.075101
                                             8.170735
     34
                  2.318591
                                             7.412094
     35
                  2.471946
                                            10.336631
     36
                  2.784836
                                            15.996620
     37
                                            12.568237
                  2.831760
     38
                  2.959932
                                            21.342916
     39
                                            20.114413
                  3.020874
     40
                  3.211366
                                            22.839406
     41
                  3.270044
                                            16.983279
     42
                  3.316073
                                            25.142082
     43
                  3.335932
                                            26.104740
     44
                  3.610778
                                            28.912188
                                            17.843957
     45
                  3.704057
     46
                  4.130868
                                            34.530743
     47
                  4.133534
                                            27.698383
     48
                  4.899032
                                            41.514822
[]: # Shallow copy
     df_copy=df.copy()
```

### 2 Exploratory Data Analysis

```
[]: df.head()
        Temperature (°C)
                          Ice Cream Sales (units)
[]:
               -4.662263
                                         41.842986
     1
               -4.316559
                                         34.661120
               -4.213985
                                         39.383001
     3
               -3.949661
                                         37.539845
                                         32.284531
               -3.578554
[]: df.shape
[]: (49, 2)
[]: df.info()
    <class 'pandas.core.frame.DataFrame'>
    RangeIndex: 49 entries, 0 to 48
    Data columns (total 2 columns):
         Column
                                   Non-Null Count
                                                   Dtype
        _____
     0
         Temperature (°C)
                                   49 non-null
                                                   float64
         Ice Cream Sales (units)
                                                   float64
                                   49 non-null
    dtypes: float64(2)
    memory usage: 912.0 bytes
[]: df.describe()
[]:
            Temperature (°C)
                              Ice Cream Sales (units)
     count
                   49.000000
                                             49.000000
                    0.271755
                                             15.905308
     mean
     std
                    2.697672
                                             12.264682
    min
                   -4.662263
                                              0.328626
     25%
                   -2.111870
                                              4.857988
     50%
                    0.688781
                                             12.615181
     75%
                    2.784836
                                             25.142082
                    4.899032
                                             41.842986
     max
[]: df.isna().sum()
[]: Temperature (°C)
                                0
     Ice Cream Sales (units)
                                 0
     dtype: int64
[]: df.duplicated().sum()
```

[]:0

[]: df.columns

[]: Index(['Temperature (°C)', 'Ice Cream Sales (units)'], dtype='object')

### 3 Univariate Analysis

```
[]: sns.distplot(df['Temperature (°C)'],color='green')
```

<ipython-input-64-3bd4a6f783fa>:1: UserWarning:

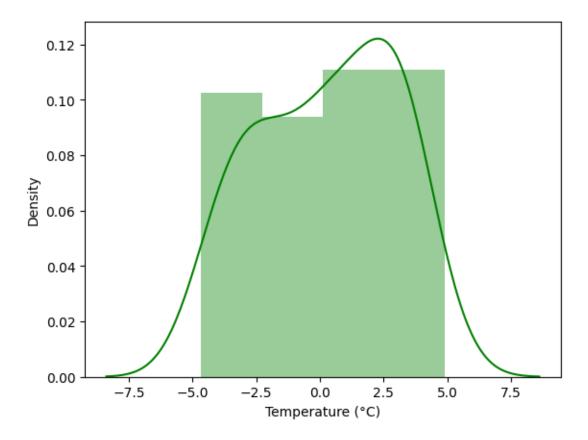
`distplot` is a deprecated function and will be removed in seaborn v0.14.0.

Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

For a guide to updating your code to use the new functions, please see https://gist.github.com/mwaskom/de44147ed2974457ad6372750bbe5751

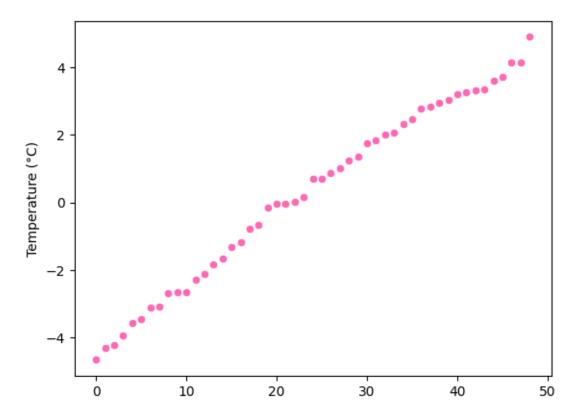
sns.distplot(df['Temperature (°C)'],color='green')

[]: <Axes: xlabel='Temperature (°C)', ylabel='Density'>



```
[]: sns.scatterplot(df['Temperature (°C)'],color='hotpink')
```

[]: <Axes: ylabel='Temperature (°C)'>



## 4 Splitting independent and dependent variable

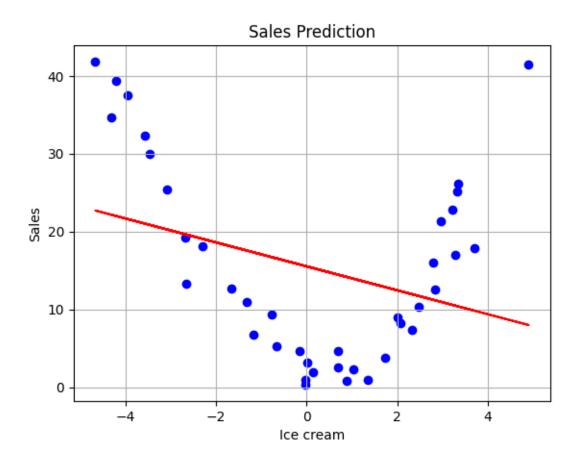
```
[]: X=df.iloc[:,[0]]
Y=df.iloc[:,[-1]]
```

# 5 Train test split

```
[]: from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(X,Y,test_size=0.2,random_state=3)
```

### 6 Model Building

```
[]: from sklearn.linear_model import LinearRegression
     lr=LinearRegression()
     lr.fit(x_train,y_train)
[]: LinearRegression()
[]: y_predict=lr.predict(x_test)
[]: y_predict
[]: array([[18.77660674],
            [10.87747649],
            [19.60829152],
            [ 9.16923147],
            [12.67856553],
            [13.61709043],
            [18.32579355],
            [ 9.16512885],
            [ 9.96963269],
            [20.31029702]])
[]: y_test
[]:
         Ice Cream Sales (units)
     12
                       11.218294
     39
                       20.114413
     9
                       20.279679
                       34.530743
     46
     31
                        4.857988
     28
                        1.292361
     13
                       10.012868
     47
                       27.698383
     44
                       28.912188
     6
                       22.635401
[]: plt.scatter(x_train,y_train,color='blue')
     plt.plot(x_train,lr.predict(x_train),color='red')
     plt.title('Sales Prediction')
     plt.xlabel('Ice cream')
     plt.ylabel('Sales')
     plt.grid()
     plt.show()
```



```
from sklearn.preprocessing import PolynomialFeatures

pf=PolynomialFeatures(degree=5)
X_poly=pf.fit_transform(X)
lrg=LinearRegression()
lrg.fit(X_poly,Y)

[]: LinearRegression()

[]: plt.scatter(X,Y,color='green')
plt.plot(X,lrg.predict(pf.fit_transform(X)),color='blue')
```

[]: # Fitting Polynomial Regression

plt.show()

