	<pre>import pandas as pd import numpy as np import seaborn as sns import matplotlib.pyplot as plt</pre>
2]:	UNIVARIATE REGRESSION: data = pd.read_csv('ex1data1.txt', header=None)
3]:	data.head() 0 1
	 0 6.1101 17.5920 1 5.5277 9.1302 2 8.5186 13.6620 3 7.0032 11.8540
4]:	4 5.8598 6.8233 # NOW WE NAME THE COLUMNS data.columns = ['Population', 'Profit']
5]:	data.head() Population Profit
5]:	Population Profit 0 6.1101 17.5920 1 5.5277 9.1302 2 8.5186 13.6620
6]:	3 7.0032 11.8540 4 5.8598 6.8233
6]:	<pre>data.isna().sum() Population 0 Profit 0 dtype: int64</pre>
7]:	<pre>data.info() <class 'pandas.core.frame.dataframe'=""> RangeIndex: 97 entries, 0 to 96 Data columns (total 2 columns):</class></pre>
	# Column Non-Null Count Dtype O Population 97 non-null float64 Profit 97 non-null float64 dtypes: float64(2) memory usage: 1.6 KB
8]: 8]:	data.describe() Population Profit
	count 97.000000 97.000000 mean 8.159800 5.839135 std 3.869884 5.510262 min 5.026900 -2.680700
	25% 5.707700 1.986900 50% 6.589400 4.562300 75% 8.578100 7.046700 max 22.203000 24.147000
9]:	<pre>X = data.iloc[:,:-1].values y = data.iloc[:,-1].values.reshape(-1,1)</pre>
[]: []:	X.shape, y.shape ((97, 1), (97, 1))
1]:	<pre>m = y.size ones = np.ones((m,1)) # ADDING A MATRIX FULL OF ONES X1 = np.c_[ones, X]</pre>
	X1.shape, y.shape ((97, 2), (97, 1))
3]:	<pre>n = X1.shape[1] theta1 = np.zeros((n,1)) # ADDING A MATRIX FULL OF ZEROES theta1.shape</pre> (2, 1)
4]:	<pre>def calCost(error): m = error.shape[0] squared_error = np.dot(error.T,error) return squared_error[0][0]/m</pre>
5]:	<pre>def GradientDescent(X,y,theta,alpha,epochs): cost_history = [] theta_history = [theta]</pre>
	<pre>m = y.size for i in range(epochs): h_theta = np.dot(X, theta) error = h_theta - y cost_history.append(calCost(error)) diff = np.dot(X.T. error)/m</pre>
67	<pre>diff = np.dot(X.T, error)/m theta = theta - (alpha*diff) theta_history.append(theta) return cost_history, theta_history</pre>
6]:	cost_history, theta_history = GradientDescent(X=X1, y=y, theta=theta1, alpha=0.001, epochs=2000)Plotting the cost history:
7]:	<pre>plt.figure(figsize=(8,4),dpi=200) plt.plot(cost_history);</pre>
	60 -
	50 -
	40 -
	30 -
	20 -
	10 -
8]:	0 250 500 750 1000 1250 1500 1750 2000 theta_history[-1] # the latest value of theta
8]: 9]:	<pre>array([[-1.12369018], [0.91454707]]) from sklearn.linear_model import LinearRegression</pre>
9]:	<pre>model = LinearRegression()</pre>
_	<pre>model.fit(X1,y) LinearRegression()</pre>
2]:	model.intercept_ , model.coef_ (array([-3.89578088]), array([[0.
,	TIME FOR MULTIVARIATE REGRESSION:
3]:	new_data = pd.read_csv('ex1data2.txt', header=None)
4]: 4]:	<pre>new_data.head() 0 1 2</pre>
	0 2104 3 399900
	1 1600 3 329900 2 2400 3 369000 3 1416 2 232000
5]:	1 1600 3 329900 2 2400 3 369000 3 1416 2 232000 4 3000 4 539900 new_data.isna().sum()
5]: 5]:	1 1600 3 329900 2 2400 3 369000 3 1416 2 232000 4 3000 4 539900 new_data.isna().sum() 0 0 0 1 0 0 0 0 0 0 0 0 0 0 0 0 0 0 0
5]: 5]:	1 1600 3 329900 2 2400 3 369000 3 1416 2 232000 4 3000 4 539900 new_data.isna().sum() 0 0 1 0 2 0 dtype: int64 new_data.info() <class 'pandas.core.frame.dataframe'=""> RangeIndex: 47 entries, 0 to 46 Data columns (total 3 columns): # Column Non-Null Count Dtype</class>
5]: 5]:	1 1600 3 329900 2 2400 3 369000 3 1416 2 232000 4 3000 4 539900 new_data.isna().sum() 0 0 1 0 2 0 dtype: int64 new_data.info() <class 'pandas.core.frame.dataframe'=""> RangeIndex: 47 entries, 0 to 46 Data columns (total 3 columns): # Column Non-Null Count Dtype</class>
5]: 5]:	1 1000 3 329900 2 2400 3 369000 3 1416 2 232000 4 3000 4 539900 new_data.isna().sum() 0 0 0 2 0 dtype: int64 new_data.info() <class' pandas.core.frame.dataframe'=""> RangeIndex: 47 entries, 0 to 48 Data columns (total 3 columns): ## Column Non-Null Count Dtype 0 0 4 7 non-null int64 1 1 47 non-null int64 1 1 1 7 non-null int64 1 1 1 7 non-null int64 1 1 1 8 non-null int64 1 1 1 1 8 non-null int64 1 1 1 1 8 non-null int64 1 1 1 1 8 non-null int64 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1</class'>
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5]: 5]: 7]: 7]:	1 1000 3 329900 2 2003 3 399900 3 1216 2 222000 4 3000 4 539900 new_data.isna().sum() 6 0 0 2 0 0 2 0 0 3 1216 5 0 2 0 0 4 3000 4 539900 cclass 'pandas.core.frame.DataFrame'> RangeIndex: 47 entries, 8 to 46 Data columns (total 3 columns): # Column Non-Null Lount Diype # Column Non-Null Lount Diype # Column Non-Null int64 1 1 47 non-null int64 1 1 47 non-null int64 1 1 47 non-null int64 2 2 47 non-null int64 4 types: int84(3) memory usage: 1.2 kB new_data.describe() new_data.describe() 1 2 count 47.00000 47.00000 47.00000 mean 200.65065 317013 340412.659574 std 794.70354 0.76092 12539.899586
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