Гэ

```
1 from google.colab import drive
2 drive.mount('/content/drive')
    Mounted at /content/drive

1 !cp "/content/drive/MyDrive/Colab Notebooks/steps_vs_cal.csv" "/content/"

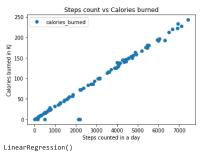
1 import pandas as pd
2 import numpy as np
3 import matplotlib.pyplot as plt
4 %matplotlib inline
5

1 dataset = pd.read_csv('steps_vs_cal.csv')
2 dataset = dataset[['step_count', 'calories_burned']]
3 dataset
4 # df.drop('date')
```

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index	step_count	calories_burned
0	5464	181
1	6041	197
2	25	0
3	5461	174
4	6915	223
5	4545	149
6	4340	140
7	1230	38
8	61	1
9	1258	40
10	3148	101
11	4687	152
12	4732	150
13	3519	113
14	1580	49
15	2822	86
16	181	6
17	3158	99
18	4383	143
19	3881	125
20	4037	129
21	202	6
22	292	9
23	330	10
24	2209	72
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```
1 # step_count = df['step_count']
2 # calories_burned = df['calories_burned']
3
4 # step_count = step_count.to_numpy()
5 # calories_burned = calories_burned.to_numpy()
6
7 dataset.plot(x='step_count', y='calories_burned', style='o')
8 plt.title('Steps count vs Calories burned')
9 plt.xlabel('Steps counted in a day')
10 plt.ylabel('Calories burned in KJ')
11 X = dataset.iloc[:, :-1].values
12 y = dataset.iloc[:, 1].values
13 plt.show()
14
15 from sklearn.model_selection import train_test_split
16 X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=0)
17
18 from sklearn.linear_model import LinearRegression
19 regressor = LinearRegression()
20 regressor.fit(X_train, y_train)
21
22
```



From the above graph we can recongnise a few outliers, but majority of the data follows a simple trend of a linear relation. We can simply say, there is a positive relationship between steps walked and calories burned. And judging from the linear regression coeficient we can see that it is 0.9892, it shows that the linear model is workable with the current dataset.

```
1 y_pred = regressor.predict(X_test)
2 df = pd.DataFrame({ 'Actual': y_test, 'Predicted': y_pred})
3 df
4
```

		1 to 20 of 20 entries Filter 🚨 🔞
index	Actual	Predicted
0	141	142.81528159788144
1	99	97.34145933124304
2	0	-3.6708807073598244
3	116	115.34497406354254
4	125	126.40617407434645
5	180	179.42021376576713
6	6	1.5109427211249002
7	21	17.122846640277594
8	220	217.25416815707547
9	25	22.038935533968232
10	35	34.12985686709926
11	47	43.563432852289395
12	39	36.38860041284901
13	113	112.3886773639583
14	38	36.35538359599975
15	17	13.402563153160358
16	9	5.198009391392876
17	72	68.87464729142631
18	131	130.62470981420262
19	1	-2.4750753007864263

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The following table shows calories burned in kJ when a certain amount of steps are taken, in predicted values and actual values.

```
1 y_pred = regressor.predict(dataset[['step_count']].to_numpy()) #to predict other amount of steps walked, please replace what is in the initial ()s !
2 df = pd.DataFrame({ 'Steps count': dataset['step_count'].to_numpy(), 'Actual': dataset['calories_burned'].to_numpy(), 'Predicted': y_pred})
3 df
```

Predicted	Actual	Steps count	ndex
176.99538613577	181	5464	0
196.1614894577	197	6041	1
-3.6708807073598	0	25	2
176.89573568522	174	5461	3
225.19298738404	223	6915	4
146.46913145130	149	4545	5
139.65968399720	140	4340	6
36.35538359599	38	1230	7
-2.4750753007864	1	61	8
37.285454467779	40	1258	9
100.06523831288	101	3148	10
151.18591944389	152	4687	11
152.6806762021	150	4732	12
112_3886773639	113	3519	13
47.98126949324	49	1580	14
89.23655602002	86	2822	15
1.5109427211249	6	181	16
100.39740648137	99	3158	17
141.08800712171	143	4383	18
124.41316506339	125	3881	19
129.59498849187	129	4037	20
2.2084958749593	6	202	21
5.198009391392	9	292	22
6.460248431664	10	330	23
68.87464729142	72	2209	24

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