

Linear Regression using error square method
matrix method

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

```
dataset = pd.read_csv('/content/sales.csv')
print(dataset.head())
weeks = dataset['x1(week)'].values
sales = dataset['y1(sales in thousands)'].values
```

```
X = weeks.reshape(-1, 1)
```

```
Y = sales.reshape(-1, 1)
```

```
X_b = np.c_[np.ones((len(X), 1)), X]
```

```
theta = np.linalg.pinv(X_b.T.dot(X_b)).dot(X_b.T).dot(Y)
```

```
b = theta[0]
```

```
m = theta[1]
```

```
print(f"The regression equation is:  $y = \{m[0]:.2f\}x + \{b[0]:.2f\}$ ")
```

```
predicted_sales_7 = m*7 + b
```

```
predicted_sales_9 = m*9 + b
```

```
print(f"Predicted sales for the 7th week: {predicted_sales_7[0]:.2f} thousand")
```

```
print(f"Predicted sales for the 9th week: {predicted_sales_9[0]:.2f} thousand")
```

```
plt.scatter(weeks, sales, color='blue', label='Data points')
plt.plot(weeks, m + weeks + b, color='red',
label=f'Linear Regression:  $y = \{m[0]:.2f\}x + \{b[0]:.2f\}$ ')

```

```
plt.xlabel('Weeks')
plt.ylabel('Sales (in thousands)')
plt.title('Sales data and Linear regression')
plt.legend()
plt.show()

```

Output:-

	x_i (week)	y_i (Sales in thousands)
0	0	1.2
1	1	2.8
2	2	2.6
3	3	3.2
4	4	3.8

The regression equation is: $y = 0.66x + 0.54$

Predicted Sales for the 7th week: 5.16 thousand

Predicted Sales for the 9th week: 6.48 thousand

Sales Data and Linear Regression

- Data points

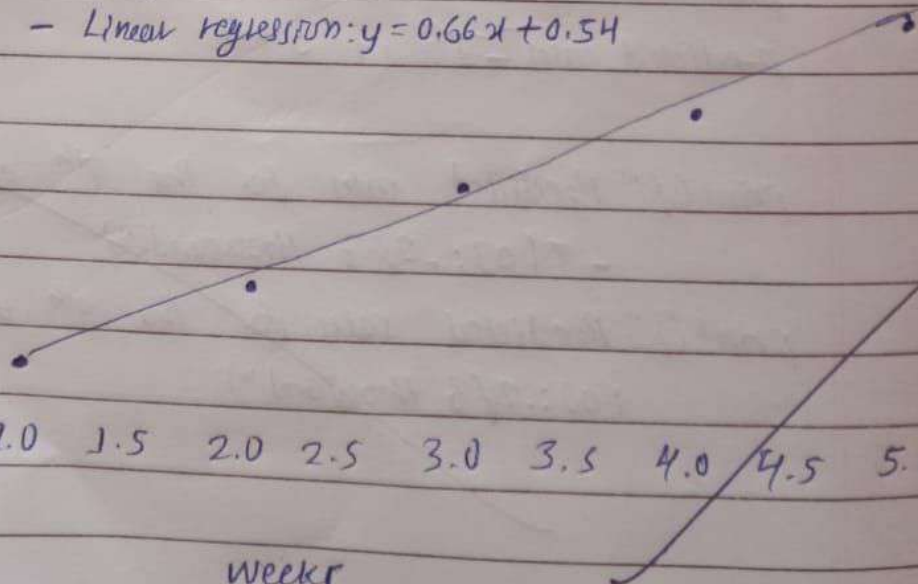
- Linear regression: $y = 0.66x + 0.54$

Sales (in thousands)

3.5
3.0
2.5
2.0
1.5

1.0 1.5 2.0 2.5 3.0 3.5 4.0 4.5 5.0

Weeks



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```

```
dataset = pd.read_csv('content/sales.csv')
print(dataset.head())
```

```
weeks = dataset['x1(weeks)'].values
sales = dataset['y1(sales in thousands)'].values
n = len(weeks)
```

```
sum_x = np.sum(weeks)
sum_y = np.sum(sales)
sum_x2 = np.sum(weeks**2)
sum_xy = np.sum(weeks * sales)
```

$$m = (n * \text{sum_xy} - \text{sum_x} * \text{sum_y}) / (n * \text{sum_x2} - \text{sum_x}^2)$$
$$b = (\text{sum_y} - m * \text{sum_x}) / n$$

```
print(f"The regression equation is:  $y = \{m:.2f\}x + \{b:.2f\}$ ")
```

```
print(f"Predicted sales for 7th week:  $\{predicted\_sales_7:1.2f\}$  thousand")
```

```
print(f"Predicted sales for 9th week:  $\{predicted\_sales_9:1.2f\}$  thousand")
```

```

plt.scatter(weeks, sales, color='blue', label='Data points')
plt.plot(weeks, m * weeks + b, color='red',
         label=f'Linear Regression: y = {m:.2f}x + {b:.2f}')
plt.xlabel('Weeks')
plt.ylabel('Sales (in thousands)')
plt.title('Sales Data and Linear Regression')
plt.legend()
plt.show()

```

Output:-

The regression equation is: $y = 0.66x + 0.54$

Predicted sales for the 7th week: 5.16 thousand

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Sales Data and Linear Regression

- Data points

- Linear regression: $y = 0.66x + 0.54$

