



Machine Learning is a subset of Artificial Intelligence which involves developing algorithms and models that allow computers to learn or make predictions.

Types of Machine Learning:

1. Supervised Learning : Labelled Data with a particular target
2. Unsupervised Learning : Unlabelled Data without any particular target
3. Reinforcement Learning : Trains using trial and error approach with feedbacks.
4. Deep Learning: Subset of ML. Used to create a neural network which can recognize complex patterns.

```
In [1]: import pandas as pd
        from sklearn.model_selection import train_test_split
        from sklearn.linear_model import Lasso
        import matplotlib.pyplot as plt
        import kagglehub
        import os
        from sklearn.metrics import r2_score
```

```
In [2]: path = kagglehub.dataset_download("tylermorse/retail-business-sales-20172019")
        print(os.listdir(path))
```

Downloading from [https://www.kaggle.com/api/v1/datasets/download/tylermorse/retail-business-sales-20172019?dataset\\_version\\_number=2...](https://www.kaggle.com/api/v1/datasets/download/tylermorse/retail-business-sales-20172019?dataset_version_number=2...)

100%|██████████| 10.7k/10.7k [00:00<00:00, 9.24MB/s]

Extracting files...

['business.retailsales2.csv', 'business.retailsales.csv']

```
In [3]: csv_path = os.path.join(path, 'business.retailsales.csv')
        df = pd.read_csv(csv_path)
        df.head()
```

```
Out[3]:
```

	Product Type	Net Quantity	Gross Sales	Discounts	Returns	Total Net Sales
0	Art & Sculpture	34	14935.0	-594.00	-1609.0	12732.00
1	Basket	13	3744.0	-316.80	0.0	3427.20
2	Basket	12	3825.0	-201.60	-288.0	3335.40
3	Basket	17	3035.0	-63.25	0.0	2971.75
4	Art & Sculpture	47	2696.8	-44.16	0.0	2652.64

```
In [4]: df.shape
```

```
Out[4]: (1775, 6)
```

```
In [5]: df.isnull().sum()
```

```
Out[5]:
```

	<b>0</b>
<b>Product Type</b>	8
<b>Net Quantity</b>	0
<b>Gross Sales</b>	0
<b>Discounts</b>	0
<b>Returns</b>	0
<b>Total Net Sales</b>	0

**dtype:** int64

```
In [6]: df = df.drop('Product Type', axis=1)
```

```
In [7]: x = df.iloc[:, :-1].values  
y = df.iloc[:, -1].values
```

```
In [8]: x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3,random_state=
```

```
In [9]: model=Lasso()  
model.fit(x_train,y_train)  
prediction=model.predict(x_test)  
print(prediction)  
print(y_test)
```

[8.64720730e+02 6.79976628e+01 7.56465528e+01 2.19936332e+02  
1.57590465e+02 7.19960488e+01 1.15978295e+02 2.36444633e+02  
1.74954489e+02 2.03942788e+02 6.00008907e+01 6.49988733e+01  
1.80178373e+01 2.80138024e+01 2.80138024e+01 2.88370525e+01  
5.57010979e+02 9.69334935e+01 7.59944348e+01 9.59863650e+01  
2.14151932e+02 1.40194513e+01 1.80178373e+01 4.80057326e+01  
2.12861131e+02 4.80057326e+01 6.12504989e+01 1.95032262e+02  
6.79976628e+01 5.85014959e+01 1.48586730e+02 1.09980716e+02  
1.34412124e+02 4.14422817e+01 2.95538241e+02 8.36237811e+01  
5.31245472e+02 3.80097675e+01 1.07217205e+02 1.18914114e+03  
3.38888317e+02 1.51780057e+02 1.87949244e+02 1.29223062e+02  
5.22368415e+01 1.67663553e+02 2.10043498e+02 1.44966594e+02  
5.80016977e+01 2.24559522e+02 4.86452698e+01 8.34262277e+02  
6.66527400e+01 2.71915351e+02 1.02667680e+02 4.28933034e+02  
4.91385706e+02 9.88972065e+01 2.29932297e+02 1.57590465e+02  
3.57875361e+01 1.31971839e+02 5.80016977e+01 7.79936279e+01  
1.44966594e+02 2.40154164e+01 1.34621775e+03 1.69405368e+02  
4.80057326e+01 5.40033117e+01 2.76002135e+02 4.80057326e+01  
3.06377996e+01 6.79976628e+01 5.49803180e+02 1.43966997e+02  
3.60105745e+01 1.17073657e+02 1.66594200e+02 1.51963770e+02  
4.18244406e+01 1.44966594e+02 3.32519854e+02 4.80057326e+01  
1.80178373e+01 1.84950454e+02 2.03647599e+02 6.96146969e+01  
1.80178373e+01 3.09900018e+02 7.49948383e+01 1.48237945e+02  
1.59960542e+02 1.84950454e+02 6.79976628e+01 3.41887106e+02  
1.64958524e+02 1.75954086e+02 2.50150129e+01 1.44966594e+02  
1.24974664e+02 6.79976628e+01 2.98608695e+02 3.48399442e+02  
1.48586730e+02 8.99887860e+01 9.79855580e+01 1.80178373e+01  
2.20162234e+01 8.75082110e+01 1.09980716e+02 1.74954489e+02  
1.74954489e+02 1.58446639e+02 1.49964576e+02 5.94497519e+01  
1.84950454e+02 3.69875808e+02 5.40033117e+01 3.08062889e+02  
2.35289229e+01 4.56407230e-02 1.74954489e+02 9.86103622e+01  
2.11204708e+02 2.43926648e+02 1.86232084e+02 2.80138024e+01  
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1.54962559e+02 3.00129955e+01 4.80057326e+01 4.40073466e+01  
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 2.71151033e+02 4.20081536e+01 2.40154164e+01 2.09021802e+02  
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 1.49964576e+02 1.98107273e+02 3.40113815e+01 6.99742656e+02  
 3.40113815e+01]

[	864.8	68.	75.59	220.	157.5	72.	116.	236.4	175.
	204.	60.	65.	18.	28.	28.	28.8	557.19	96.91
	76.	96.	214.2	14.	18.	48.	212.81	48.	61.2
	194.99	68.	58.5	148.5	110.	134.4	41.4	295.6	83.6
	531.	38.	107.2	1189.4	339.	151.8	188.	129.2	52.19
	167.6	210.	145.	58.	224.6	48.6	834.4	66.6	272.
	102.6	429.	491.4	98.8	230.	157.5	35.75	132.	58.
	78.	145.	24.	1346.4	169.4	48.	54.	276.	48.
	30.6	68.	550.	144.	36.	117.	166.5	152.	41.8
	145.	332.5	48.	18.	185.	203.58	69.6	18.	310.
	75.	148.	160.	185.	68.	342.	165.	176.	25.
	145.	125.	68.	298.59	348.4	148.5	90.	98.	18.
	22.	87.5	110.	175.	175.	158.4	150.	59.4	185.
	370.	54.	308.	23.48	0.	175.	98.6	211.2	244.
	186.21	28.	315.	220.	990.	61.2	155.	30.	48.
	44.	48.	176.38	16.	175.	737.	111.5	18.	782.
	851.5	175.	160.	118.2	192.	197.2	135.	127.6	24.
	228.	44.	766.52	10.5	397.59	14.	54.4	56.	140.
	36.	91.2	21.6	136.	1183.2	64.6	67.5	24.	18.
	78.4	38.	604.8	88.	185.	72.	63.8	66.	410.4
	56.	198.	166.5	129.2	18.	83.6	120.	166.	88.2
	1107.49	61.2	121.59	52.2	43.2	92.	62.5	372.61	90.
	22.	90.	1030.	125.	58.	185.	352.	68.	82.
	90.	890.4	56.	148.2	36.	114.01	64.	18.	56.
	38.	32.	522.	291.2	264.	115.	18.	204.	42.
	28.	34.	121.51	165.	44.	38.	38.	28.	58.
	32.	32.	48.	81.	69.6	34.	62.	98.	251.69
	857.5	185.	204.	123.2	520.	28.	34.84	38.	28.

52.2	251.6	200.	72.	28.	66.	186.19	196.	43.2
39.6	54.4	39.	216.	116.	158.	23.41	103.5	285.
58.	307.8	44.	145.	90.	110.	148.2	56.	818.4
14.	288.	957.	49.35	30.4	166.5	56.	185.	28.
44.	185.	34.	144.	127.4	88.	225.	79.2	54.
472.71	58.	145.	23.4	173.6	187.2	34.	273.68	372.8
120.	305.5	44.	19.8	50.4	116.	46.8	58.	96.
48.	48.	296.	27.2	36.54	187.2	79.8	182.	66.39
139.2	145.	28.	132.	18.	28.8	406.8	175.	180.
102.	248.	22.	111.59	48.	130.5	153.6	96.	63.8
110.	166.5	0.	88.	157.51	396.	60.	165.	0.
79.2	122.4	186.2	34.	72.83	350.	1516.83	279.8	110.
134.	44.	18.	68.	72.	34.	56.	350.	126.
36.	80.	136.32	64.8	32.	130.6	64.8	149.6	18.
91.79	240.	925.	88.	31.	106.	1032.5	603.	272.4
54.	210.	585.	14.	565.	84.	13.5	104.4	125.
70.	36.	125.	28.8	48.	507.5	34.	185.	43.2
89.1	28.	174.	18.	90.	145.	185.	66.6	18.
43.2	56.	87.2	95.2	68.	189.	36.	36.	32.
148.4	16.2	44.	328.5	319.	54.	34.	44.	421.2
58.	91.8	278.6	457.8	72.	103.5	124.	128.	40.49
240.	356.4	34.	240.	115.2	879.	88.	0.	64.
58.	78.	28.	98.8	42.	172.4	74.	175.	116.
332.8	104.4	264.59	157.5	38.	148.	207.6	145.	156.6
56.2	26.	36.	43.2	30.6	68.	125.	24.	28.88
124.8	18.	18.	99.	112.5	136.8	259.	18.	24.
77.2	34.	30.	72.2	110.2	58.	271.2	42.	24.
209.	84.	313.2	81.2	110.	34.6	18.	48.	55.1
48.	347.6	68.4	28.8	36.	174.	891.	34.2	38.1
39.6	204.	46.	164.	81.6	38.	68.	14.4	48.
528.	66.	145.	166.5	168.	734.4	150.	198.	34.
700.	34.	]						

```
In [10]: r2 = r2_score(y_test, prediction)
print("R² Score:", r2)
```

R² Score: 0.999998935727327

```
In [11]: plt.scatter(y_test, prediction, color='blue')
plt.plot([y_test.min(), y_test.max()], [y_test.min(), y_test.max()], 'r--')
plt.xlabel('Gross Sales')
plt.ylabel('Total Net Sales')
plt.title('Gross Sales vs Net Sales (Training Data)')
plt.show()
```

Gross Sales vs Net Sales (Training Data)

