

Exam marks prediction using Linear Regression-Multiple variable

Import Libraries

```
In [ ]: import pandas as pd
import numpy as np
from sklearn.linear_model import LinearRegression
import matplotlib.pyplot as plt
```

Load Dataset

```
In [ ]: df=pd.read_csv("exammark.csv")
```

```
In [ ]: df.head()
```

```
Out[ ]:
```

	hours	age	internet	marks
0	6.83	15	1	78.50
1	6.56	16	0	76.74
2	NaN	17	1	78.68
3	5.67	18	0	71.82
4	8.67	19	1	84.19

```
In [ ]: df.tail()
```

```
Out[ ]:
```

	hours	age	internet	marks
196	8.56	19	1	84.68
197	8.94	20	1	86.75
198	6.60	15	1	78.05
199	8.35	16	1	83.50
200	4.15	15	0	81.45

```
In [ ]: df.describe()
```

```
Out[ ]:
```

	hours	age	internet	marks
count	196.000000	201.000000	201.000000	201.000000
mean	6.981429	17.467662	0.552239	77.951244
std	1.266266	1.720523	0.498505	4.919626
min	4.150000	15.000000	0.000000	68.570000
25%	5.757500	16.000000	0.000000	73.400000
50%	7.110000	17.000000	1.000000	77.770000
75%	8.082500	19.000000	1.000000	82.300000
max	8.990000	20.000000	1.000000	86.990000

```
In [ ]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 201 entries, 0 to 200
Data columns (total 4 columns):
#   Column      Non-Null Count  Dtype
---  -
0   hours       196 non-null    float64
1   age         201 non-null    int64
2   internet    201 non-null    int64
3   marks       201 non-null    float64
dtypes: float64(2), int64(2)
memory usage: 6.4 KB
```

Load Summarize

```
In [ ]: print(df.shape)
print(df.head(5))
```

```
(201, 4)
   hours  age  internet  marks
0   6.83   15         1   78.50
1   6.56   16         0   76.74
2    NaN   17         1   78.68
3   5.67   18         0   71.82
4   8.67   19         1   84.19
```

Finding & Removing NA values from our features X

```
In [ ]: df.columns[df.isna().any()]
```

```
Out[ ]: Index(['hours'], dtype='object')
```

```
In [ ]: df.hours=df.hours.fillna(df.hours.mean())
```

Segregate Data into Input X & Output Y

```
In [ ]: X=df.iloc[:, :-1].values
print(X.shape)
X
```

```
(201, 3)
```

```
Out[ ]: array([[ 6.83      , 15.      , 1.      ],
 [ 6.56      , 16.      , 0.      ],
 [ 6.98142857, 17.      , 1.      ],
 [ 5.67      , 18.      , 0.      ],
 [ 8.67      , 19.      , 1.      ],
 [ 7.55      , 20.      , 0.      ],
 [ 6.67      , 15.      , 0.      ],
 [ 8.99      , 16.      , 0.      ],
 [ 5.19      , 17.      , 1.      ],
 [ 6.75      , 18.      , 0.      ],
 [ 6.59      , 19.      , 0.      ],
 [ 8.56      , 20.      , 1.      ],
 [ 7.75      , 15.      , 0.      ],
 [ 7.9       , 16.      , 1.      ],
 [ 8.19      , 17.      , 0.      ],
 [ 6.55      , 18.      , 1.      ],
 [ 6.36      , 19.      , 0.      ],
 [ 8.44      , 20.      , 1.      ],
 [ 8.41      , 15.      , 0.      ],
 [ 7.67      , 16.      , 1.      ],
 [ 7.42      , 17.      , 1.      ],
 [ 8.16      , 18.      , 1.      ],
 [ 5.05      , 19.      , 1.      ],
 [ 5.85      , 20.      , 1.      ],
 [ 5.45      , 15.      , 0.      ],
 [ 7.96      , 16.      , 0.      ],
 [ 6.51      , 17.      , 0.      ],
 [ 6.73      , 18.      , 0.      ],
 [ 5.94      , 19.      , 1.      ],
 [ 7.48      , 20.      , 0.      ],
 [ 8.13      , 15.      , 1.      ],
 [ 6.98142857, 16.      , 1.      ],
 [ 5.4       , 17.      , 1.      ],
 [ 8.78      , 18.      , 0.      ],
 [ 8.72      , 19.      , 1.      ],
 [ 7.1       , 20.      , 0.      ],
 [ 7.86      , 15.      , 1.      ],
 [ 7.19      , 16.      , 0.      ],
 [ 5.62      , 17.      , 1.      ],
 [ 7.88      , 18.      , 0.      ],
 [ 5.28      , 19.      , 1.      ],
 [ 8.92      , 20.      , 1.      ],
 [ 5.46      , 15.      , 0.      ],
 [ 8.3       , 16.      , 1.      ],
 [ 8.09      , 17.      , 0.      ],
 [ 6.18      , 18.      , 1.      ],
 [ 7.01      , 19.      , 1.      ],
 [ 5.01      , 20.      , 0.      ],
 [ 5.54      , 15.      , 1.      ],
 [ 5.09      , 16.      , 1.      ],
 [ 5.09      , 17.      , 0.      ],
 [ 7.31      , 18.      , 1.      ],
 [ 8.71      , 19.      , 0.      ],
 [ 5.52      , 20.      , 1.      ],
 [ 8.76      , 15.      , 0.      ],
 [ 8.69      , 16.      , 1.      ],
 [ 5.75      , 17.      , 1.      ],
 [ 8.93      , 18.      , 1.      ],
 [ 5.39      , 19.      , 1.      ],
 [ 5.65      , 20.      , 0.      ],
 [ 5.49      , 15.      , 1.      ],
 [ 7.26      , 16.      , 1.      ],
 [ 6.35      , 17.      , 0.      ]],
```

[7.72	, 18.	, 1.],
[8.88	, 19.	, 0.],
[5.45	, 20.	, 1.],
[7.86	, 15.	, 1.],
[8.26	, 16.	, 0.],
[5.07	, 17.	, 1.],
[8.25	, 18.	, 0.],
[5.37	, 19.	, 1.],
[5.11	, 20.	, 1.],
[6.35	, 15.	, 0.],
[7.41	, 16.	, 1.],
[7.31	, 17.	, 0.],
[6.04	, 18.	, 1.],
[5.11	, 19.	, 1.],
[6.56	, 20.	, 0.],
[5.09	, 15.	, 1.],
[5.88	, 16.	, 0.],
[8.34	, 17.	, 1.],
[7.94	, 18.	, 0.],
[6.66	, 19.	, 1.],
[6.01	, 20.	, 1.],
[6.88	, 15.	, 0.],
[5.63	, 16.	, 1.],
[5.88	, 17.	, 0.],
[8.05	, 18.	, 1.],
[5.33	, 19.	, 0.],
[8.79	, 20.	, 0.],
[7.52	, 15.	, 1.],
[8.2	, 16.	, 0.],
[5.44	, 17.	, 1.],
[7.9	, 18.	, 0.],
[7.69	, 19.	, 1.],
[6.09	, 20.	, 0.],
[6.98142857,	15.	, 1.],
[5.2	, 16.	, 1.],
[8.88	, 17.	, 0.],
[8.07	, 18.	, 1.],
[6.24	, 19.	, 1.],
[7.95	, 20.	, 0.],
[8.26	, 15.	, 0.],
[7.31	, 16.	, 1.],
[7.23	, 17.	, 1.],
[6.46	, 18.	, 1.],
[5.34	, 19.	, 1.],
[5.72	, 20.	, 1.],
[5.84	, 15.	, 0.],
[5.02	, 16.	, 1.],
[7.98	, 17.	, 0.],
[6.37	, 18.	, 1.],
[6.92	, 19.	, 0.],
[7.95	, 20.	, 1.],
[7.12	, 15.	, 0.],
[5.79	, 16.	, 1.],
[5.4	, 17.	, 0.],
[8.83	, 18.	, 1.],
[5.69	, 19.	, 0.],
[6.6	, 20.	, 1.],
[6.52	, 15.	, 0.],
[8.31	, 16.	, 0.],
[6.98142857,	17.	, 1.],
[7.62	, 18.	, 0.],
[8.69	, 19.	, 1.],
[8.75	, 20.	, 0.],

[6.46	, 15.	, 1.],
[7.14	, 16.	, 1.],
[6.38	, 17.	, 0.],
[6.33	, 18.	, 1.],
[5.64	, 19.	, 0.],
[5.26	, 20.	, 1.],
[6.83	, 15.	, 1.],
[5.76	, 16.	, 0.],
[6.51	, 17.	, 1.],
[8.33	, 18.	, 0.],
[8.16	, 19.	, 1.],
[5.14	, 20.	, 0.],
[8.71	, 15.	, 0.],
[8.6	, 16.	, 1.],
[8.6	, 17.	, 0.],
[7.43	, 18.	, 1.],
[7.81	, 19.	, 1.],
[6.51	, 20.	, 0.],
[8.11	, 15.	, 1.],
[8.95	, 16.	, 0.],
[7.99	, 17.	, 1.],
[5.92	, 18.	, 0.],
[8.3	, 19.	, 1.],
[8.97	, 20.	, 0.],
[5.39	, 15.	, 0.],
[6.77	, 16.	, 0.],
[8.08	, 17.	, 1.],
[5.24	, 18.	, 0.],
[6.93	, 19.	, 1.],
[5.14	, 20.	, 0.],
[8.39	, 15.	, 1.],
[6.18	, 16.	, 0.],
[7.53	, 17.	, 1.],
[7.86	, 18.	, 0.],
[7.7	, 19.	, 1.],
[7.3	, 20.	, 0.],
[7.79	, 15.	, 1.],
[6.75	, 16.	, 0.],
[7.87	, 17.	, 1.],
[5.38	, 18.	, 0.],
[7.8	, 19.	, 1.],
[5.07	, 20.	, 0.],
[7.95	, 15.	, 1.],
[8.35	, 16.	, 0.],
[5.19	, 17.	, 0.],
[7.19	, 18.	, 0.],
[7.35	, 19.	, 1.],
[5.22	, 20.	, 1.],
[5.39	, 15.	, 1.],
[5.39	, 16.	, 1.],
[8.93	, 17.	, 1.],
[5.79	, 18.	, 0.],
[8.42	, 19.	, 1.],
[7.26	, 20.	, 0.],
[6.97	, 15.	, 1.],
[5.55	, 16.	, 1.],
[8.66	, 17.	, 0.],
[8.61	, 18.	, 1.],
[5.22	, 19.	, 1.],
[8.05	, 20.	, 0.],
[8.87	, 15.	, 1.],
[5.54	, 16.	, 0.],
[6.98142857,	17.	, 0.],

```
[ 7.26      , 18.      , 1.      ],
[ 5.79      , 19.      , 0.      ],
[ 5.22      , 20.      , 0.      ],
[ 8.71      , 15.      , 1.      ],
[ 7.55      , 16.      , 1.      ],
[ 6.35      , 17.      , 1.      ],
[ 7.53      , 18.      , 0.      ],
[ 8.56      , 19.      , 1.      ],
[ 8.94      , 20.      , 1.      ],
[ 6.6       , 15.      , 1.      ],
[ 8.35      , 16.      , 1.      ],
[ 4.15      , 15.      , 0.      ]])
```

```
In [ ]: Y=df.iloc[:, -1].values
Y
```

```
Out[ ]: array([78.5 , 76.74, 78.68, 71.82, 84.19, 81.18, 76.99, 85.46, 70.66,
77.82, 75.37, 83.88, 79.5 , 80.76, 83.08, 76.03, 76.04, 85.11,
82.5 , 80.58, 82.18, 83.36, 70.67, 75.02, 70.96, 83.33, 74.75,
75.65, 74.15, 80.17, 82.27, 76.14, 71.1 , 84.35, 83.08, 76.76,
81.24, 78.21, 73.08, 83.23, 70.27, 86.41, 71.1 , 82.84, 82.38,
72.96, 77.46, 70.11, 72.38, 71.41, 72.22, 77.77, 84.44, 71.45,
82.21, 85.48, 75.03, 86.65, 70.9 , 71.7 , 73.61, 79.41, 76.19,
80.43, 85.78, 70.06, 81.25, 81.7 , 69.27, 82.79, 71.8 , 71.79,
74.97, 78.61, 77.59, 72.33, 72.08, 77.33, 70.05, 73.34, 84. ,
82.93, 76.63, 75.36, 77.29, 72.87, 73.4 , 81.74, 71.85, 84.6 ,
79.56, 82.1 , 72.08, 79.1 , 81.01, 76.48, 75.39, 68.57, 83.64,
82.3 , 75.18, 82.03, 82.99, 79.26, 77.55, 77.07, 72.1 , 73.25,
74.25, 70.58, 81.08, 75.04, 76.38, 80.86, 78.42, 74.44, 70.34,
85.04, 73.61, 75.55, 76.2 , 82.69, 76.83, 79.53, 83.57, 85.95,
76.02, 77.65, 77.01, 74.49, 73.19, 71.86, 75.8 , 72.46, 78.39,
83.48, 83.15, 71.22, 85.98, 83.91, 84.58, 80.31, 82.55, 75.52,
83.82, 85.15, 82.75, 74.34, 82.02, 86.12, 71.87, 76.7 , 81.7 ,
70.78, 78.45, 70.2 , 83.37, 75.52, 81.57, 80.72, 80.81, 79.49,
79.17, 77.07, 82.04, 71.94, 81.6 , 70.79, 82.68, 83.08, 71.18,
77.63, 77.78, 70.4 , 73.02, 71.11, 85.96, 73.64, 84.24, 78.17,
77.19, 71.83, 86.99, 83.87, 71.5 , 79.63, 85.1 , 72.01, 77.27,
79.87, 73.14, 70.51, 84.03, 79.64, 74.24, 81.67, 84.68, 86.75,
78.05, 83.5 , 81.45])
```

Training Dataset using Linear Regression

```
In [ ]: model=LinearRegression()
model.fit(X,Y)
```

```
Out[ ]: ▾ LinearRegression
LinearRegression()
```

Predicted Marks

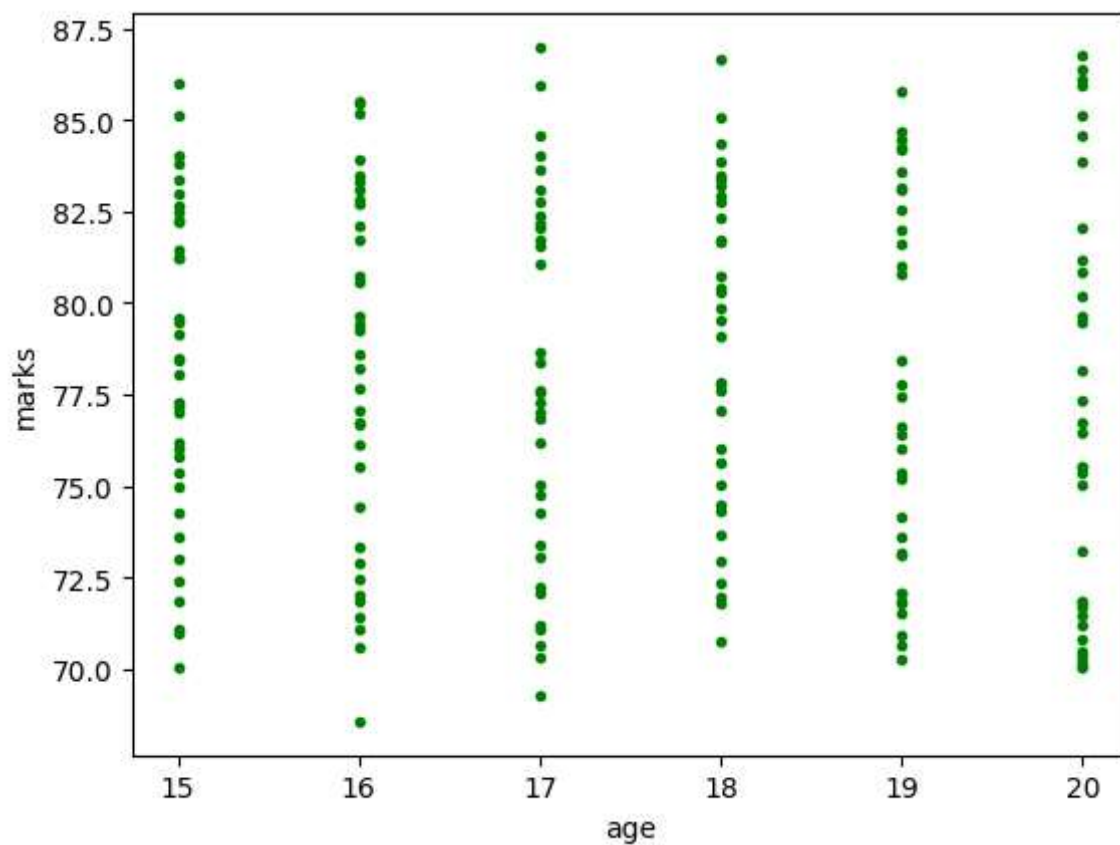
```
In [ ]: d = int(input("Enter the number of hours studied: "))
b = int(input("Enter the age: "))
c = int(input("Is Internet Available (0 or 1): "))

a=[[d,b,c]]
PredictedmodelResult=model.predict(a)
print(PredictedmodelResult)
```

[70.80240294]

```
In [ ]: plt.xlabel("age")
plt.ylabel("marks")
plt.scatter(df.age,df.marks,color="green",marker=".")
```

```
Out[ ]: <matplotlib.collections.PathCollection at 0x1ed26a78450>
```



```
In [ ]: plt.xlabel("internet")
plt.ylabel("marks")
plt.scatter(df.internet,df.marks,color="blue",marker=".")
```

```
Out[ ]: <matplotlib.collections.PathCollection at 0x1ed2b64fc90>
```

