

Figure 1. A schematic diagram of the experimental setup. The participant is seated in a chair, viewing a screen displaying a target (a red dot) and a starting point (a green dot). The participant's hand is positioned at the starting point, and the target is located at a distance of 10 cm from the starting point. The participant is instructed to move their hand from the starting point to the target. The screen is divided into two regions: a starting region (green) and a target region (red). The participant's hand is positioned at the starting point, and the target is located at a distance of 10 cm from the starting point. The participant is instructed to move their hand from the starting point to the target.

Figure 1. The effect of the number of trials on the number of correct responses. The number of correct responses (Y-axis) is plotted against the number of trials (X-axis). The data shows a positive correlation between the number of trials and the number of correct responses, with a slight increase in the number of correct responses as the number of trials increases.

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emp.ipynb X

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Generate+ Code+ Markdown|Run AllRestartClear All OutputsJupyter VariablesOutline...base (Python 3.12.4)

5Kim
Name: Name, dtype: object

emp['Name'] = emp['Name'].str.replace(r'\W','',regex=True)
emp['Name']

[28]Python

0Mike
1Teddy
2Umar
3Jane
4Uttam
5Kim
Name: Name, dtype: object

emp['Domain'] = emp['Domain'].str.replace(r'\W','',regex=True)
emp['Domain']

[29]Python

0Data science
1Testing
2Data analyst
3Analytics
4Statistics
5NLP
Name: Domain, dtype: object

emp['Age'] = emp['Age'].str.replace(r'\W','',regex=True)
emp['Age']

[30]Python

034years
145yr

Watchlist Ideas

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base (Python 3.12.4)

Name: Age, dtype: object

```
emp['Age'] = emp['Age'].str.extract(r'(\d+)')
```

[21]

...	0	34
	1	45
	2	NaN
	3	NaN
	4	67
	5	55

Name: Age, dtype: object

```
emp['Location'] = emp['Location'].str.replace(r'\W', '', regex=True)
```

[24]

...	0	Mumbai
	1	Bangalore
	2	NaN
	3	Hyderabad
	4	NaN
	5	Delhi

Name: Location, dtype: object

```
emp['Salary'] = emp['Salary'].str.replace(r'\W', '', regex=True)
```

[25]

...	0	5000
	1	10000
	2	15000

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base (Python 3.12.4)

```
emp['Exp'] = emp['Exp'].str.extract(r'(\d+)')
emp['Exp']
```

[27]

0 2
1 3
2 4
3 NaN
4 5
5 10
Name: Exp, dtype: object

emp

[28]

	Name	Domain	Age	Location	Salary	Exp
0	Mike	Datascience	34	Mumbai	5000	2
1	Teddy	Testing	45	Bangalore	10000	3
2	Umar	Dataanalyst	NaN	NaN	15000	4
3	Jane	Analytics	NaN	Hyderabad	20000	NaN
4	Uttam	Statistics	67	NaN	30000	5
5	Kim	NLP	55	Delhi	60000	10

we have raw data we use regex to clean the data and removed all noise characted from the dataset

We can also work in same things in sql query as well

```
clean_data = emp.copy()
```

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base (Python 3.12.4)

```
clean_data['Age']
```

[31]

Python

```
... 0 34
1 45
2 NaN
3 NaN
4 67
5 55
Name: Age, dtype: object
```

```
import numpy as np
```

[32]

Generate + Code + Markdown

```
clean_data['Age'] = clean_data['Age'].fillna(np.mean(pd.to_numeric(clean_data['Age'])))
```

[33]

Python

```
clean_data['Age']
```

[34]

Python

```
... 0 34
1 45
2 50.25
3 50.75
4 67
5 55
Name: Age, dtype: object
```

```
clean_data['Exp']
```

[35]

Python

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Name: Exp, dtype: object

base (Python 3.12.4)

```
clean_data['Location'].isnull().sum()
clean_data['Location']
```

[37]

```
0    Mumbai
1    Bangalore
2      NaN
3    Hyderabad
4      NaN
5      Delhi
Name: Location, dtype: object
```

Python

```
clean_data['Location'] = clean_data['Location'].fillna(clean_data['Location'].mode()[0])
```

[38]

Python

```
clean_data['Location']
```

[39]

```
0    Mumbai
1    Bangalore
2    Bangalore
3    Hyderabad
4    Bangalore
5      Delhi
Name: Location, dtype: object
```

Python

```
clean_data
```

[40]

Python

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EDA Technique

```
import matplotlib.pyplot as plt
import seaborn as sns
```

[41]Python

```
import warnings
warnings.filterwarnings('ignore')
```

[50]Python

Generate+ Code+ Markdown

```
clean_data['Salary']
```


[53]Python

...05000
110000
215000
320000
430000
560000
Name: Salary, dtype: int32

```
vis1 = sns.displot(clean_data['Salary'])
```

[54]Python

...

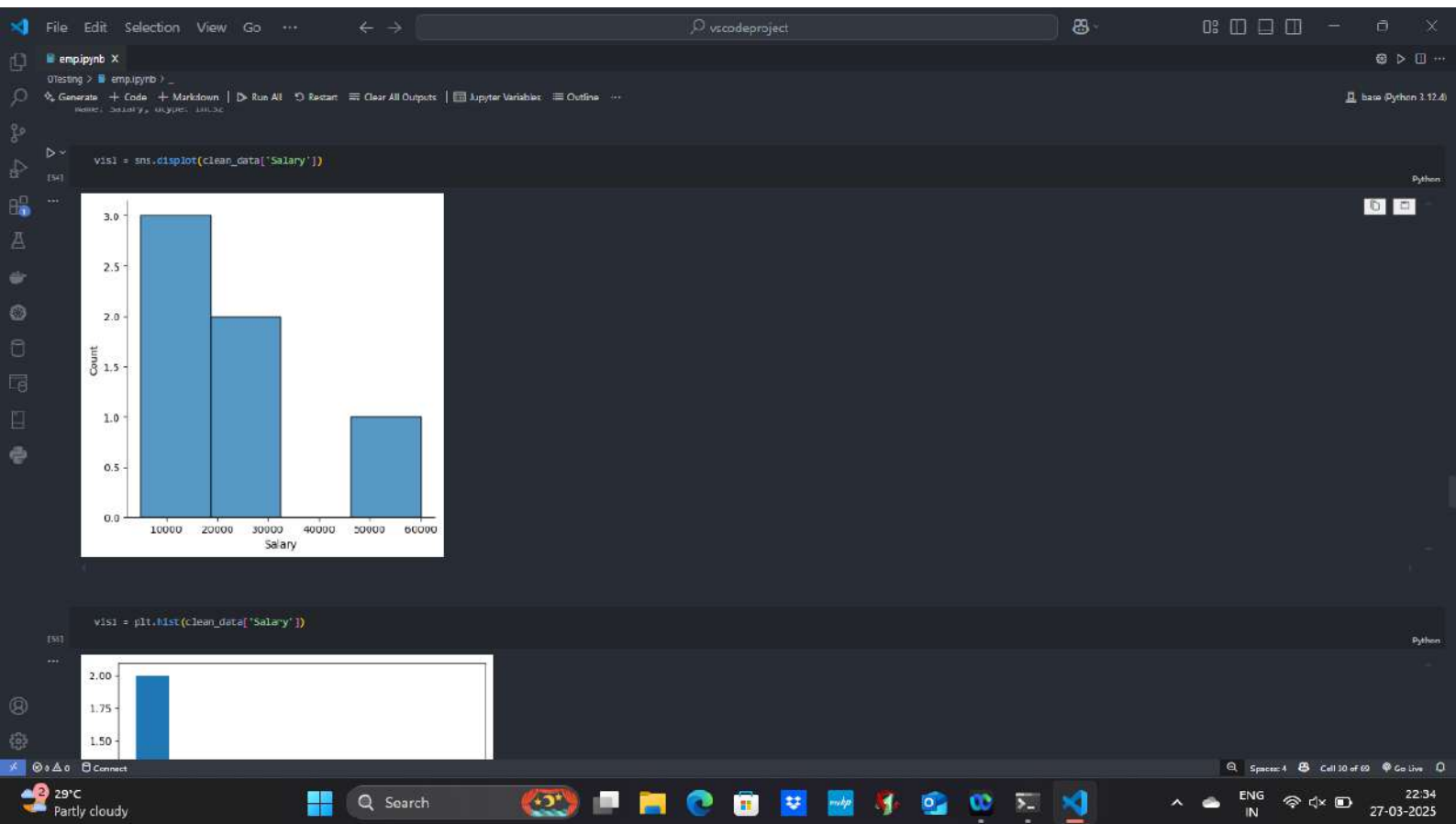


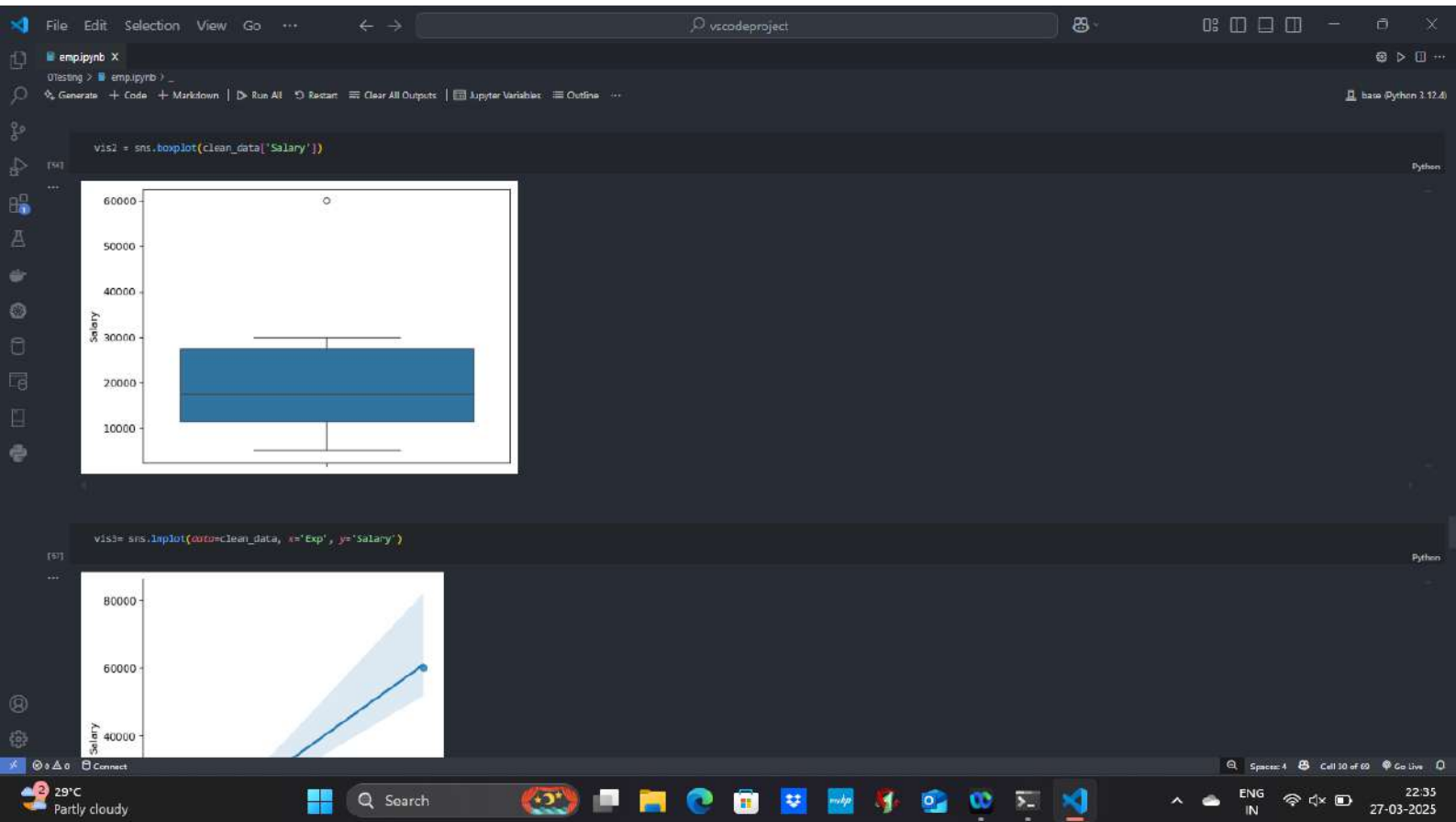
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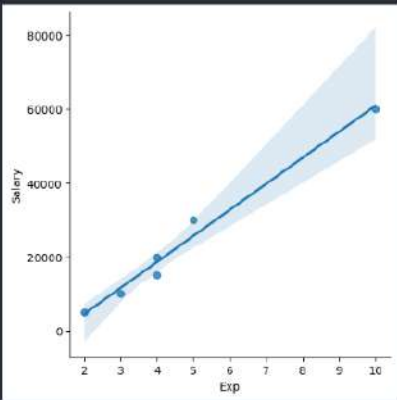
0Testing > empipynb > _

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base Python 3.12.0

```
vis3=sns.lmplot(data=clean_data, x='Exp', y='Salary')
```

[57]




Salary

Exp

```
vis4 = sns.lmplot(data=clean_data, x = 'Exp', y='Salary', fit_reg = False)
```

[58]



60000

50000

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empipynb X

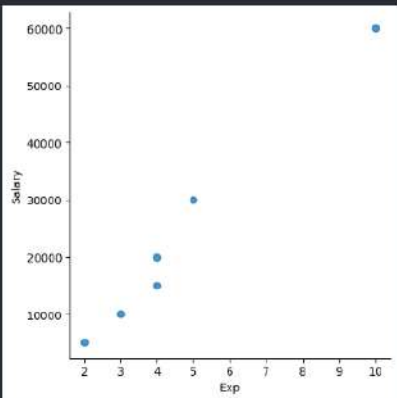
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base (Python 3.12.4)



```
vis4 = sns.lmplot(data=clean_data, x='Exp', y='Salary', fit_reg=False)
```



Python

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Add Code Cell

```
clean_data[:]
```

Name	Domain	Age	Location	Salary	Exp
Milica	Data Scientist	34	Mumbai	60000	2.5

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base Python 3.12.0

clean_data[:]

[53] Python

	Name	Domain	Age	Location	Salary	Exp
0	Mike	DataScience	34	Mumbai	5000	2
1	Teddy	Testing	45	Bangalore	10000	3
2	Umar	Dataanalyst	50	Bangalore	15000	4
3	Jane	Analytics	50	Hyderabad	20000	4
4	Uttam	Statistics	67	Bangalore	30000	5
5	Kim	NLP	55	Delhi	60000	10

clean_data[0:6:2]

[53] Python

	Name	Domain	Age	Location	Salary	Exp
0	Mike	DataScience	34	Mumbai	5000	2
2	Umar	Dataanalyst	50	Bangalore	15000	4
4	Uttam	Statistics	67	Bangalore	30000	5

clean_data[::-1]

[53] Python

	Name	Domain	Age	Location	Salary	Exp
5	Kim	NLP	55	Delhi	60000	10
4	Uttam	Statistics	67	Bangalore	30000	5
3	Jane	Analytics	50	Hyderabad	20000	4
2	Umar	Dataanalyst	50	Bangalore	15000	4
1	Teddy	Testing	45	Bangalore	10000	3
0	Mike	DataScience	34	Mumbai	5000	2

clean_data.columns

[53] Python

Index(['Name', 'Domain', 'Age', 'Location', 'Salary', 'Exp'], dtype='object')

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base Python 3.12.0

[51]

Index(['Name', 'Domain', 'Age', 'Location', 'Salary', 'Exp'], dtype=object)

Python

[52]

X_iv = clean_data[['Name', 'Domain', 'Age', 'Location', 'Exp']]

Python

[53]

X_iv

Python

[54]

	Name	Domain	Age	Location	Exp
0	Mike	DataScience	34	Mumbai	2
1	Teddy	Testing	45	Bangalore	3
2	Uniar	Dataanalyst	50	Bangalore	4
3	Jane	Analytics	50	Hyderbad	4
4	Uttam	Statistics	67	Bangalore	5
5	Kim	NLP	55	Delhi	10

Python

[55]

y_dv = clean_data['Salary']
y_dv

Python

[56]

0	5000
1	10000
2	15000
3	20000
4	30000
5	60000

Name: Salary, dtype: int32

Python

[57]

exp

Python

[58]

	Name	Domain	Age	Location	Salary	Exp
0	Mike	DataScience	34	Mumbai	5000	2
1	Teddy	Testing	45	Bangalore	10000	3

Python

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base Python 3.12.0

5 Kim NLP 55 Delhi 60000 10

X_iv

Python

	Name	Domain	Age	Location	Exp.
0	Mike	DataScience	34	Mumbai	2
1	Teddy	Testing	45	Bangalore	3
2	Umar	Dataanalyst	50	Bangalore	4
3	Jane	Analytics	50	Hyderabad	4
4	Uttam	Statistics	67	Bangalore	5
5	Kim	NLP	55	Delhi	10

y_dr

Python

```
6 5000
1 10000
2 15000
3 20000
4 30000
5 60000
Name: Salary, dtype: int32
```

clean_data

Python

	Name	Domain	Age	Location	Salary	Exp.
0	Mike	DataScience	34	Mumbai	5000	2
1	Teddy	Testing	45	Bangalore	10000	3
2	Umar	Dataanalyst	50	Bangalore	15000	4
3	Jane	Analytics	50	Hyderabad	20000	4
4	Uttam	Statistics	67	Bangalore	30000	5
5	Kim	NLP	55	Delhi	60000	10

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[72]inputation = pd.get_dummies(clean_data)Python

[73]inputationPython

AgeSalaryExpName_JaneName_KimName_MikeName_TeddyName_UmarName_UttamDomain_AnalyticsDomain_DataanalystDomain_DatascienceDomain_NLPDomain_StatisticsDomain_TestingLocation_BangaloreLocation_DelhiLocation_

0	34	5000	2	False	False	True	False	False	False	False	False	False	True	False	False	False	False	False
1	45	10000	3	False	False	False	True	False	False	False	False	False	False	False	True	True	False	False
2	50	15000	4	False	False	False	False	True	False	False	True	False	False	False	False	True	False	False
3	50	20000	4	True	False	False	False	False	True	False	False	False	False	False	False	False	False	False
4	67	30000	5	False	False	False	False	False	True	False	False	False	False	True	False	True	False	False
5	55	60000	10	False	True	False	False	False	False	False	False	False	True	False	False	False	True	True

raw data with lot of regex, missing, unclean data

regex, clean

fill missing numerical & categorical

clean_dataset (data cleaning)

outlier treatment, univariate, bivariate, co-relation

split the data into x_i.v & y_dv

impute categorical data to numerical

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