

Q1. —
→ if $i = 1$.
while $i < 11$:
 print (i)
 $i = i + 1$

Q2. —
→ $n = 4$
for i in range(4):
 for j in range (i+1):
 print (" ", end = " ")
 for j in range (j,n):
 print ("1", end = " ")
 print () .

Q3. —
→ $S = \text{int}(\text{input}())$
 $\text{sum} = 0$
for i in range (1, S+1):
 $\text{sum} = \text{sum} + i$.
 print (sum) .

Q4) → $\text{num} = \text{int}(\text{input}(\text{"please enter a number"}))$
for i in range (1, 11):
 Print (num, " * " * i, num) .
 print

Q5) — $l1 = [1, 2, 3, 4, 5]$.
for i in (l1):
 print (i) .

Q6).
04
 $\text{Num} = \text{int}(\text{input}())$.
 $\text{Total_dig} = \text{str}(\text{Num})$.
print (len (Total_dig)).

Q7] `n = int(input("Enter NO of Rows"))
 for i in range(n):
 print(" " * (n-1-i) + "*" * (2*i+1))
 for i in range(n-2, -1, -1):
 print(" " * (n-i-1) + "*" * (2*i+1))`

Q8] -
 - `li = [1, 2, 3, 4, 5]`
`li.reverse()`
`li`
 using loop -
`l = [1, 2, 3, 4, 5, 6, 7]`
`l = []`
 for i in l:
 i.insert(0, i)
 print(l)

Q9] - `n = -10`
 for n in range(-10, 0):
 print(n)

Q10] -
 → for i in range(10):
 print(i)
 else print("done")

Q11] `L = 10`
`u = 100`
 print("prime no. are")
 for num in range(1, u+1):
 if num > 1:
 for i in range(2, num):
 if (num % i) == 0:
 break
 else:
 print(num)

Q.12] -

```

num = int(input("Enter No"))
n1, n2 = 0, 1
count = 0
if num <= 0:
    print("please enter a positive number")
else:
    print("fibonacci")
    while count < num:
        print(n1)
        nth = n1 + n2
        n1 = n2
        n2 = nth
        count = 1

```

Q.13] -

```

→ num = int(input("Enter Number"))
fact = 1
if num < 0:
    print("Enter the number")
elif num == 0:
    print("The factorial of 0 is 1")
else:
    for i in range(1, num + 1):
        fact = i
    print("The factorial of", num, "is", fact)

```


Q14]

```
Num = int(input("Enter the Number"))  
reversed Num = 0
```

```
while Num > 0:
```

```
    digit = Num % 10
```

```
    reversed_Num = reversed_Num * 10 + digit  
    Num = Num // 10
```

```
print("Reversed Number:", reversed_Num)
```

Q15] -

→

```
my_list = [1, 2, 3, 4, 5, 6, 7]
```

```
odd element. my_list[1::2]
```

```
print(odd - elements)
```

Q16] -

```
Num = int(input("Enter the number"))  
for i in range(1, Num+1):
```

```
    cube = i * i * i
```

```
    print(f"the cube of {i} is {cube}")
```

17] -

```
n = int(input("Enter the number of terms"))
```

```
sum = 0
```

```
j = 1
```

```
for i in range(1, n+1):
```

```
    sum = sum + j
```

```
    j = (j * 10) + 1
```

818) -

```
n = int(input('Enter number of rows')).  
for i in range (1, n+1):  
    for j in range (1, i+1):  
        print (j, end, " ").  
    print ()
```

Q1) calculate the z-score for the below data set assume $\sigma = 1.5$ How do you perform normalisation (only formulae) Calcul.

2

3

1

3

2

4

→ we have, $z = \frac{x - \mu}{\sigma}$

where, z is the z-score.
 x is the individual data point.
 μ is the mean.
 σ is the standard deviation.

Here $\sigma = 1.5$;
 and, the dataset is: 2 3 1 3 2 4
 The mean (μ) is calculated as.

$$\mu = \frac{2 + 3 + 1 + 3 + 2 + 4}{6} = \frac{15}{6} = 2.5.$$

Now, calculate z-score for each data point.

i) for $x = 2$:

$$z = \frac{2 - 2.5}{1.5} = \frac{-0.5}{1.5} = -0.333.$$

ii) for $x = 3$:

$$z = \frac{3 - 2.5}{1.5} = \frac{0.5}{1.5} = 0.333.$$

iii) for $x = 1$:

$$z = \frac{1 - 2.5}{1.5} = \frac{-1.5}{1.5} = -1.$$

iv) for $x = 3$:

$$z = \frac{3 - 2.5}{1.5} = \frac{0.5}{1.5} = 0.333.$$

xi) for $x = 4$;

$$z = \frac{4 - 2.5}{1.5} = \frac{1.5}{1.5} = 1.$$

So, the z score for the data set are approximately $-0.333, 0.333, -1, 0.333, -0.333, 1$.

2] \rightarrow One-hot encoding is a technique used to convert categorical variables into a binary matrix, where each category is represented by a binary column. In pandas, the `get_dummies` function is commonly used for one-hot encoding.

3] list all the transformers (function and power). The question seems incomplete or ambiguous, if you're referring to mathematical transformations, it could include functions like square root, logarithm, exponential, etc. with different powers.

4] Linear regression assumes that the relationship between the independent variables is linear, and the residuals (the difference between actual and predicted values) are normally distributed with constant variance.

5] Gradient descent is an optimization algorithm used to minimize the cost function in machine learning models. The diagram typically, of steepest decrease of the cost function towards the minimum point of loss function.

6) Pandas Profiling is a library for generating exploratory data analysis reports for a pandas Dataframe. To use it -

python

copy code -

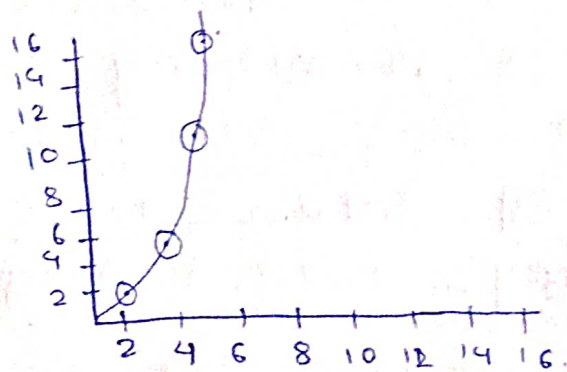
import pandas-profiling.

profile =

pandas-profiling.ProfileReport(df).
profile.to_file("output.html").

→ Draw the line for the following equation -
 $y = x^2$.

$$\begin{aligned} &= 2 \\ &2y = x \\ &2 : \\ &\text{since} \\ &= \\ &2 \\ &y = x. \\ &2. \end{aligned}$$



represents a quadratic function, the graph is a parabola. it opens upwards if the coefficient of

2

2

it positive. The specific shape and location of the parabola depend on the coefficient values.

8]. Python.

Copy code.

```
# import necessary libraries.
```

```
import seaborn as sns.
```

```
from sklearn. - selection. import
```

```
train - test - split
```

```
from sklearn. linear. model import
```

```
Linear Regression.
```

```
from sklearn. metrics import
```

```
mean - squared - error.
```

```
# Load the dataset.
```

```
mpg - data = sns.load_dataset('mpg')
```

```
# check for missing values.
```

```
print (mpg - data.isnull().sum(),
```

```
x = mpg - data[['horse power']].
```

```
y = mpg - data[['mpg']].
```

```
x = train ; x = test , y = train , y = test =
```

```
train = test - split (x, y ; test - size = 0.2).
```

```
random - state = 42).
```

```
# create and fit the model.
```

```
model = Linear Regression.
```

```
# Evaluate the model.
```

```
mse = mean - squared - error (y - test, y - pre
```

```
print(f' mean squared Error : {mse}').
```

This code imports the necessary libraries, loads the 'mpg' dataset, checks for the missing values, splits data into training and testing datasets, builds a linear regression model, makes predictions, and evaluates the model using mean squared error.