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Assignment 2 Group A

Data Wrangling II

Create an “Academic performance” dataset of students and perform the following operations using Python.

1. Scan all variables for missing values and inconsistencies. If there are missing values and/or inconsistencies, use any of the suitable techniques to deal with them.
2. Scan all numeric variables for outliers. If there are outliers, use any of the suitable techniques to deal with them.
3. Apply data transformations on at least one of the variables. The purpose of this transformation should be one of the following reasons:
 - a. To change the scale for better understanding of the variable, to convert a non-linear relation into a linear one,
 - b. To decrease the skewness and convert the distribution into a normal distribution.

Reason and document your approach properly.

```
import pandas as pd
import numpy as np
import seaborn as sns
```

```
df = pd.read_excel('./random_academic_data.xlsx')
df.head(3)
```

	RollNo	Name	Sem1	Sem2	Sem3	Sem4	Sem5	Sem6
0	TCOA01	Arjun Verma	9.71	4.14	4.39	6.89	7.56	7.52
1	TCOA02	Isha Shah	6.11	4.62	6.56	5.03	7.72	5.42
2	TCOA03	Riya Nair	4.25	8.48	4.60	5.43	6.88	7.90

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 309 entries, 0 to 308
Data columns (total 8 columns):
#   Column      Non-Null Count  Dtype
---  ---
0   RollNo      309 non-null   object
1   Name        309 non-null   object
2   Sem1        309 non-null   float64
```

```

3   Sem2    309 non-null    float64
4   Sem3    309 non-null    float64
5   Sem4    309 non-null    float64
6   Sem5    309 non-null    float64
7   Sem6    309 non-null    float64

```

```
dtypes: float64(6), object(2)
```

```
memory usage: 19.4+ KB
```

```
df.describe()
```

	Sem1	Sem2	Sem3	Sem4	Sem5
Sem6					
count	309.000000	309.000000	309.000000	309.000000	309.000000
mean	7.077476	6.966084	7.108544	6.982233	7.137087
std	3.004871	1.835419	2.758916	2.877877	3.817599
min	0.960000	4.000000	4.020000	4.000000	4.010000
25%	5.660000	5.250000	5.480000	5.390000	5.520000
50%	6.770000	7.000000	6.860000	6.750000	6.910000
75%	8.300000	8.530000	8.450000	8.410000	8.390000
max	50.600000	9.980000	45.300000	47.500000	67.200000

```

for column in df.columns:
    print(column)

```

```
RollNo
```

```
Name
```

```
Sem1
```

```
Sem2
```

```
Sem3
```

```
Sem4
```

```
Sem5
```

```
Sem6
```

Missing Values

```
df.isna().sum()
```

```

RollNo    0
Name      0
Sem1      0
Sem2      0
Sem3      0

```

```
Sem4      0
Sem5      0
Sem6      0
dtype: int64
```

Data has no missing values

RollNo is combination of Year, Dept, Div and number. We can extract Div from it and Number

```
df['Year'] = df['RollNo'].str[0]+'E'
df['Dept'] = df['RollNo'].str[1:3]+'MP'
df['Div'] = df['RollNo'].str[3]
df.head(5)
```

	RollNo	Name	Sem1	Sem2	Sem3	Sem4	Sem5	Sem6	Year
0	TCOA01	Arjun Verma	9.71	4.14	4.39	6.89	7.56	7.52	TE
1	TCOA02	Isha Shah	6.11	4.62	6.56	5.03	7.72	5.42	TE
2	TCOA03	Riya Nair	4.25	8.48	4.60	5.43	6.88	7.90	TE
3	TCOA04	Sonia Patel	9.09	6.17	9.11	9.25	9.99	7.70	TE
4	TCOA05	Anjali Mittal	5.40	4.29	9.06	5.14	6.70	5.86	TE

Finally replace rollNo with int

```
df['RollNo'] = df['RollNo'].str[4:].astype(int)
df.head(5)
```

	RollNo	Name	Sem1	Sem2	Sem3	Sem4	Sem5	Sem6	Year
0	1	Arjun Verma	9.71	4.14	4.39	6.89	7.56	7.52	TE
1	2	Isha Shah	6.11	4.62	6.56	5.03	7.72	5.42	TE
2	3	Riya Nair	4.25	8.48	4.60	5.43	6.88	7.90	TE
3	4	Sonia Patel	9.09	6.17	9.11	9.25	9.99	7.70	TE
4	5	Anjali Mittal	5.40	4.29	9.06	5.14	6.70	5.86	TE

```
df.dtypes
```

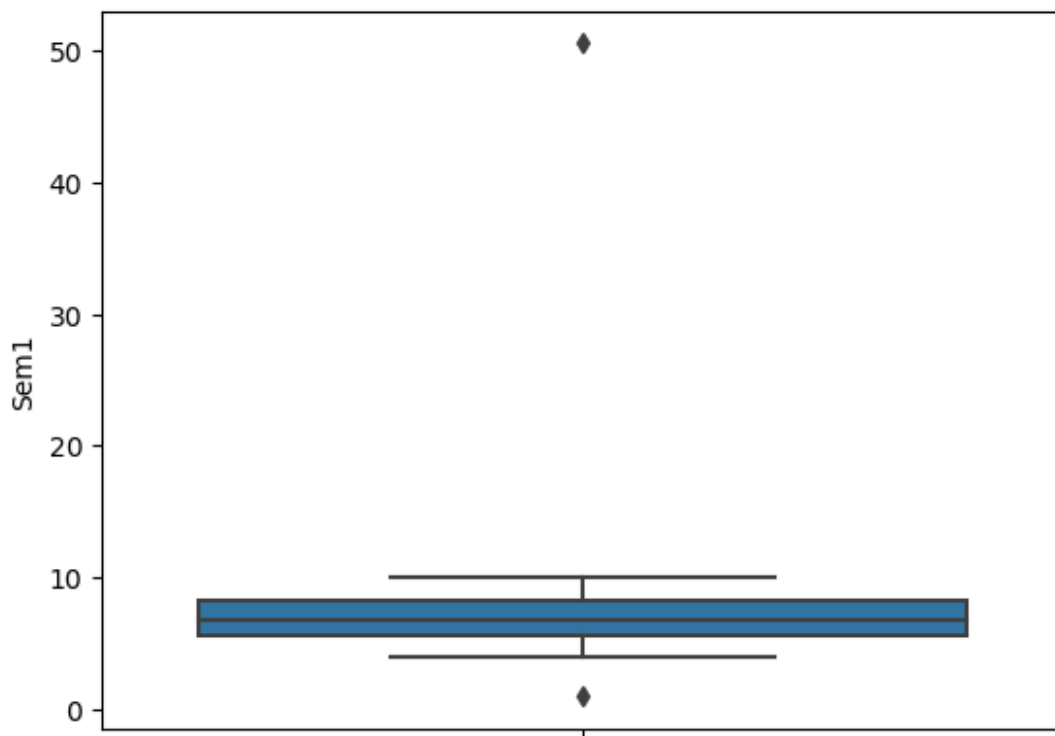
```
RollNo      int32
Name         object
Sem1         float64
```

```
Sem2      float64
Sem3      float64
Sem4      float64
Sem5      float64
Sem6      float64
Year      object
Dept      object
Div       object
dtype: object
```

Outliners

```
sns.boxplot(y=df['Sem1'])
```

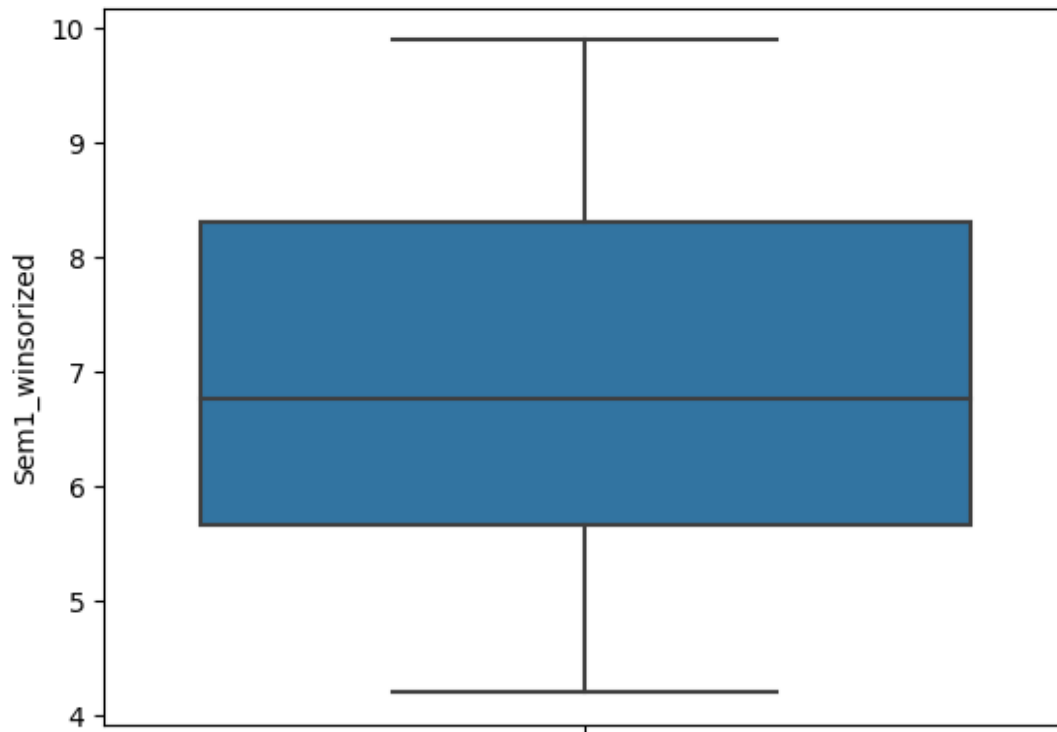
```
<Axes: ylabel='Sem1'>
```



```
def winsorize(data, lower_percentile, upper_percentile):
    lower = np.percentile(data, lower_percentile)
    upper = np.percentile(data, upper_percentile)
    return np.where(data < lower, lower, np.where(data > upper, upper,
data))
```

```
df['Sem1_winsorized'] = winsorize(df['Sem1'], 2, 98)
sns.boxplot(y=df['Sem1_winsorized'])
```

```
<Axes: ylabel='Sem1_winsorized'>
```



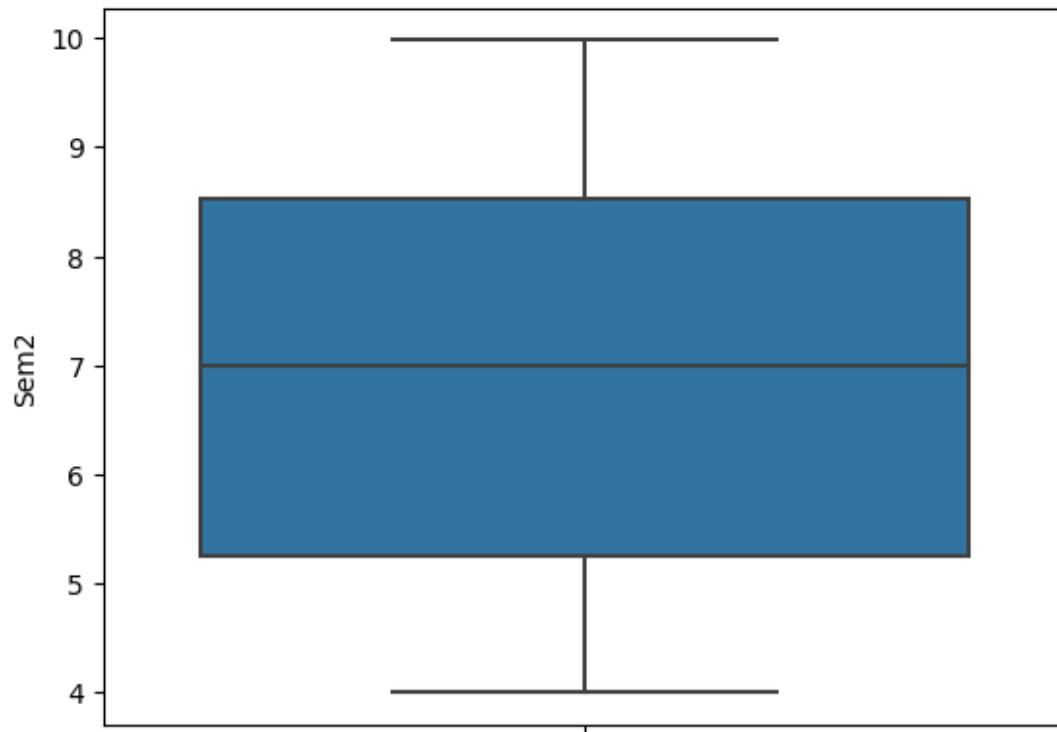
```
new_df = df.copy()
new_df['diff1'] = new_df['Sem1'] - new_df['Sem1_winsorized']
new_df[new_df['diff1']>10]
```

	RollNo	Name	Sem1	Sem2	Sem3	Sem4	Sem5	Sem6	Year	Dept
Div \										
31	32	Yash Kumar	50.6	5.82	8.99	9.03	9.16	4.59	TE	COMP
A										

	Sem1_winsorized	diff1
31	9.8884	40.7116

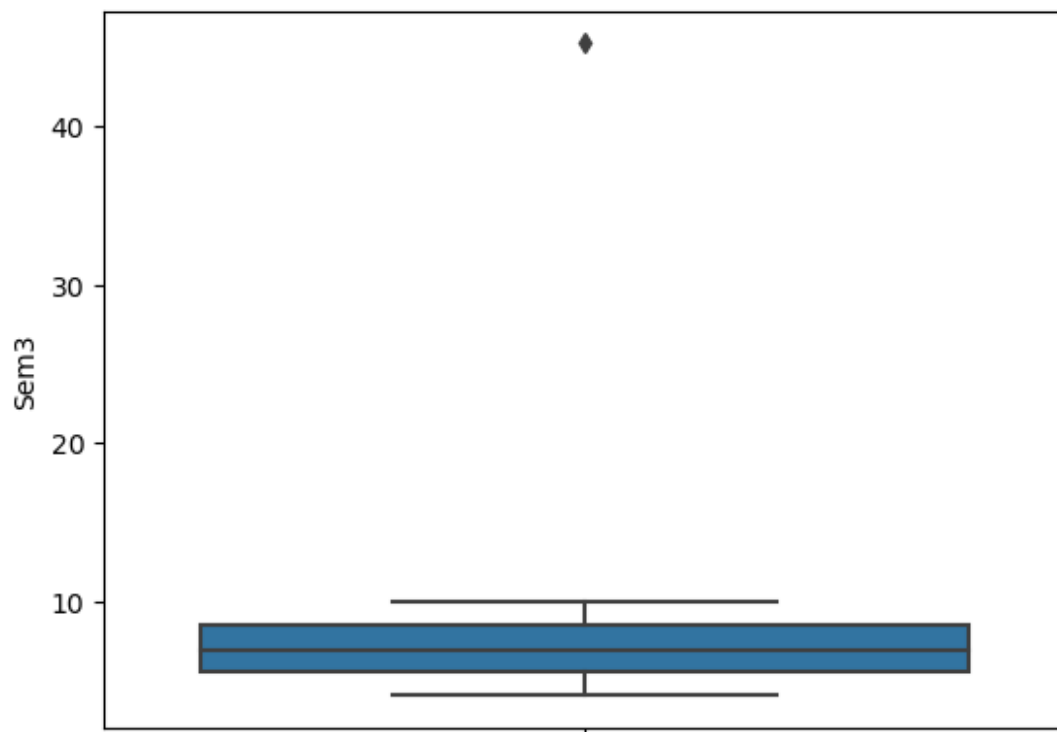
```
sns.boxplot(y=df['Sem2'])
```

```
<Axes: ylabel='Sem2'>
```

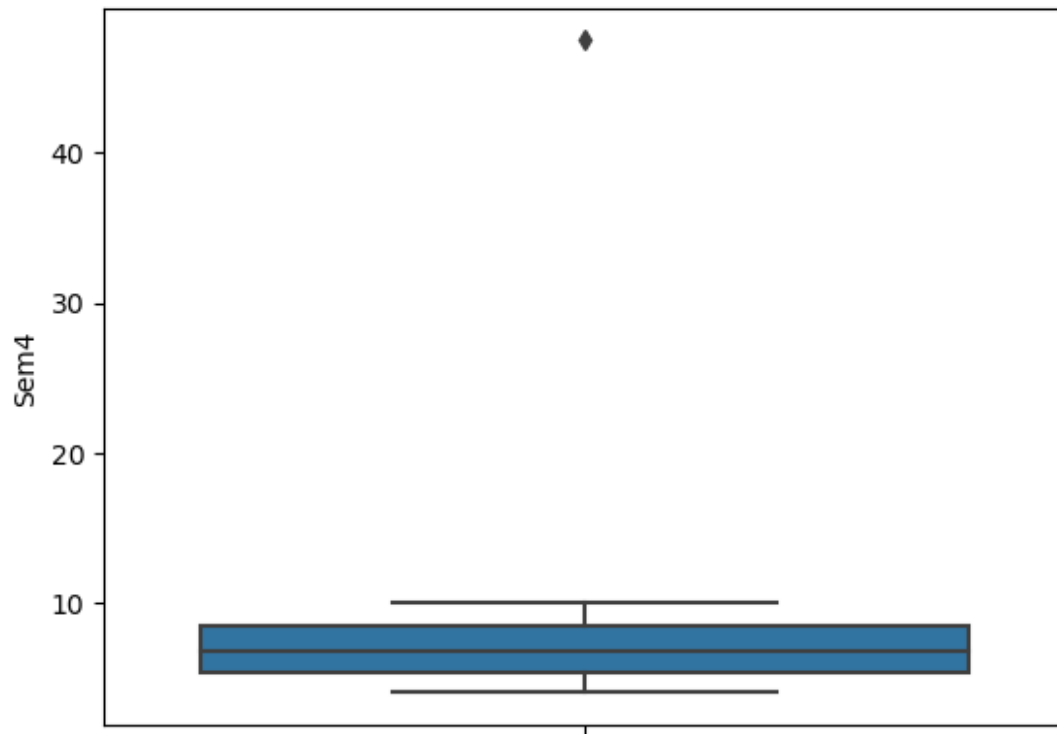


```
sns.boxplot(y=df['Sem3'])
```

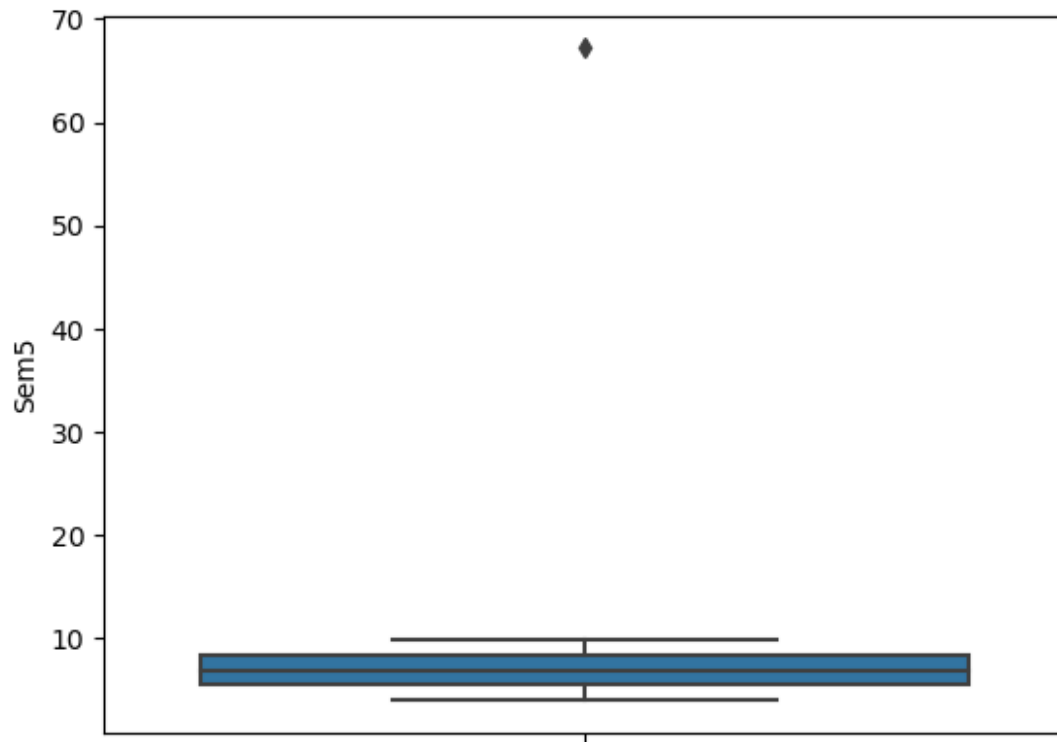
```
<Axes: ylabel='Sem3'>
```



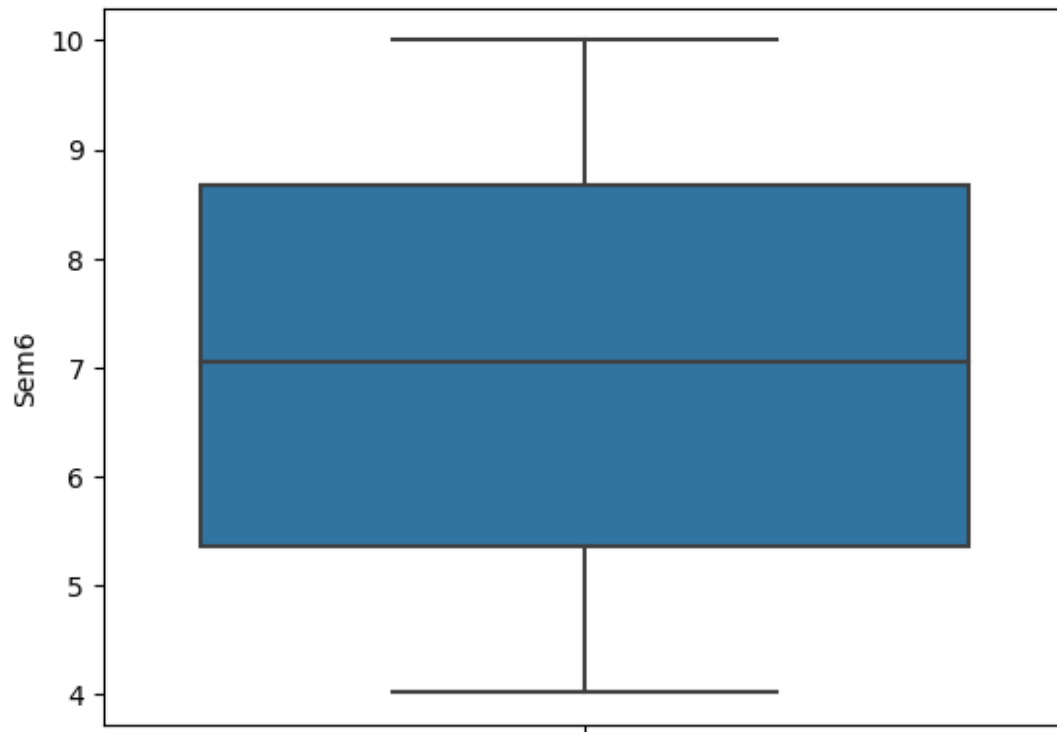
```
df['Sem3'] = winsorize(df['Sem3'], 2, 98)
sns.boxplot(y=df['Sem4'])
<Axes: ylabel='Sem4'>
```



```
df['Sem4'] = winsorize(df['Sem4'], 2, 98)
sns.boxplot(y=df['Sem5'])
<Axes: ylabel='Sem5'>
```

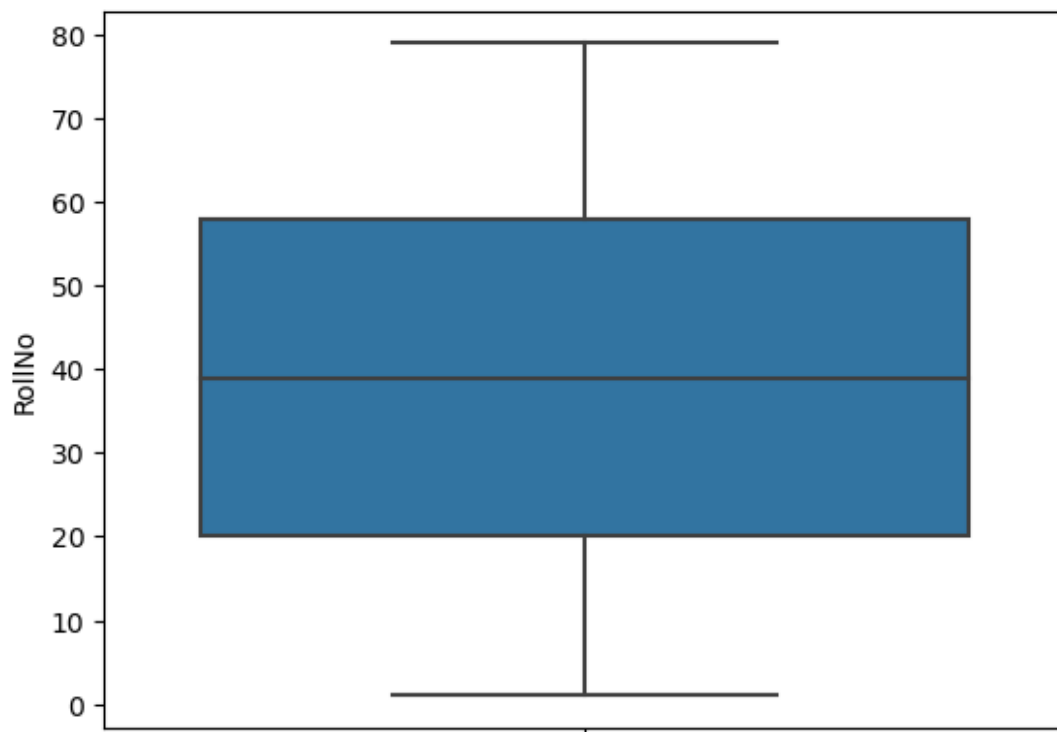


```
df['Sem5'] = winsorize(df['Sem5'], 2, 98)
sns.boxplot(y=df['Sem6'])
<Axes: ylabel='Sem6'>
```

```
sns.boxplot(y=df['RollNo'])
```

<Axes: ylabel='RollNo'>



Transformation and Scaling

```
from sklearn.preprocessing import LabelEncoder
from sklearn.preprocessing import StandardScaler

le = LabelEncoder()
df['Div'] = le.fit_transform(df['Div'])
df['Div'].value_counts()

Div
2    79
3    79
1    77
0    74
Name: count, dtype: int64
```

Thus we performed all the operation as specified in problem statement.