

## **Experiment 10**

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### **Aim:**

Outlier detection using R programming.

### **Theory and Output:**

#### **Outliers**

In R, the IQR() function is used to compute the interquartile range of a given object of numerical values. The interquartile range of these values is a range where 25% on either side is cut off. Statistically, the interquartile range is the difference between the upper quartile and the lower quartile.

Then, we calculate the interquartile range (IQR) using the IQR() function in R.

The quantile function divides the data into equal halves, in which the median acts as middle and over that the remaining lower part is lower quartile and upper part is upper quartile.

Next, we use the Tukey method to calculate the lower and upper bounds. Any data points that fall below the lower bound or above the upper bound are considered outliers. Finally, we print the lower and upper bounds as well as any identified outliers.

Note that this is just one method for outlier detection and there are many other methods that can be used depending on the specific needs of your analysis.

#### **# Outlier Program**

```
data <- c(1, 2, 3, 4, 5, 6, 7, 8, 9, 10, 50)
# Calculate interquartile range
IQR<- IQR(data)
# Calculate lower and upper bounds using Tukey method
lower bound <- quantile(data, 0.25) - 1.5 * IQR
upper bound <- quantile(data, 0.75) + 1.5 * IQR
# Identify outliers
outliers <- data[data < lower bound | data > upper bound]
#Print results
```

```
cat("Lower bound:", lower bound, "\n")
cat("Upper bound:", upper bound, "\n")
cat("Outliers:", outliers, "\n")

# Create a vector of random data
data <- rnorm(100, mean = 50, sd = 10)

# Calculate the quartiles and interquartile range
q1 <- quantile(data, 0.25)
q3 <- quantile(data,
0.75) iqr <- q3 - q1

# Calculate the lower and upper bounds for outliers
lower <- q1 - 1.5 * iqr
upper <- q3 + 1.5 * iqr

# Create a box plot of the data
boxplot(data, main = "Outlier Detection", ylim = c(0, 100), ylab = "Data")
abline(h = lower, col = "red")
abline(h = upper, col = "red")

# Identify the outliers and plot them as points
outliers <- data[data < lower | data > upper]
points(rep(1, length(outliers)), outliers, col = "red", pch = 19)
```

## Output:

Environment	History	Connections	Tutorial
<div> <div> <div>Import Dataset</div> <div>103 MiB</div> </div> <div> <div>Global Environment</div> <div>Search</div> </div> </div>			
Data			
glm_model	List of 30		
lm_model	List of 12		
Values			
data	num [1:11] 1 2 3 4 5 6 7 8 9 10 ...		
IQR	5		
lower_bound	Named num -4		
outliers	50		
p	num [1:10] 0.246 0.387 0.958 0.535 0.564 ...		
upper_bound	Named num 16		
x	num [1:10] -0.5605 -0.2302 1.5587 0.0705 0.1293 ...		
y	int [1:10] 1 1 1 0 0 1 1 0 0 0		

## Outlier Detection

