



## ANOMALY DETECTION IN R

# What is an anomaly?

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# Defining the term anomaly

***Anomaly:*** a data point or collection of data points that do not follow the same pattern or have the same structure as the rest of the data



# Point anomaly

- A single data point
- Unusual when compared to the rest of the data

**Example:** A single 30C daily high temperature among a set of ordinary spring days

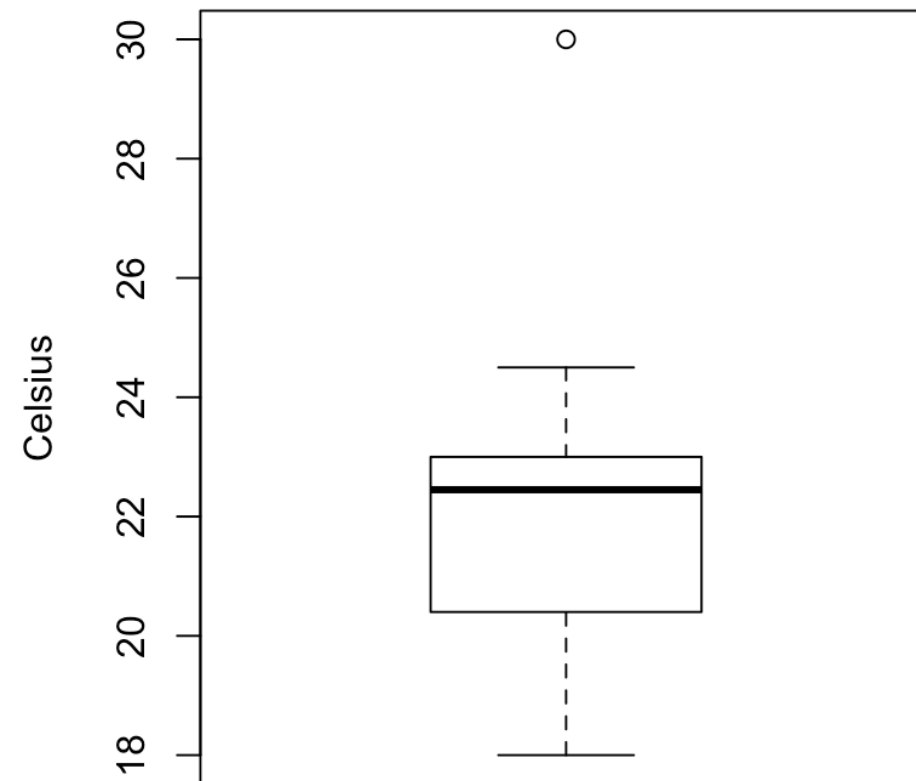
```
summary(temperature)
```

Min.	1st Qu.	Median	Mean	3rd Qu.	Max.
18.00	20.45	22.45	22.30	22.98	30.00



# Visualizing point anomalies with a boxplot

```
boxplot(temperature, ylab = "Celsius")
```

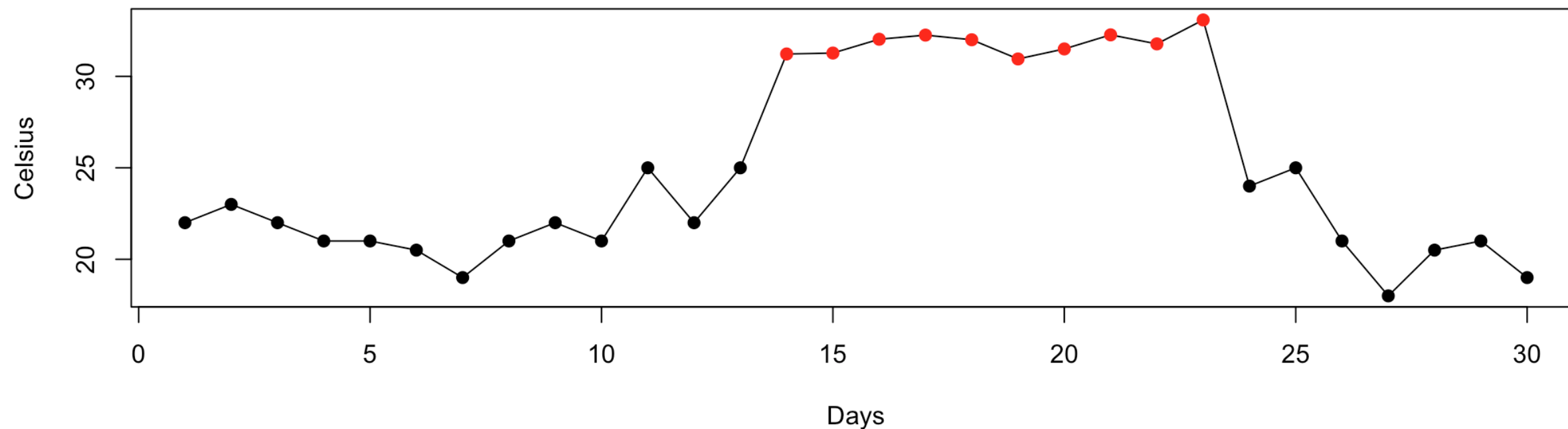




# Collective anomaly

- An anomalous collection of data instances
- Unusual when considered together

**Example:** 10 consecutive high daily temperatures





## ANOMALY DETECTION IN R

**Let's practice!**



ANOMALY DETECTION IN R

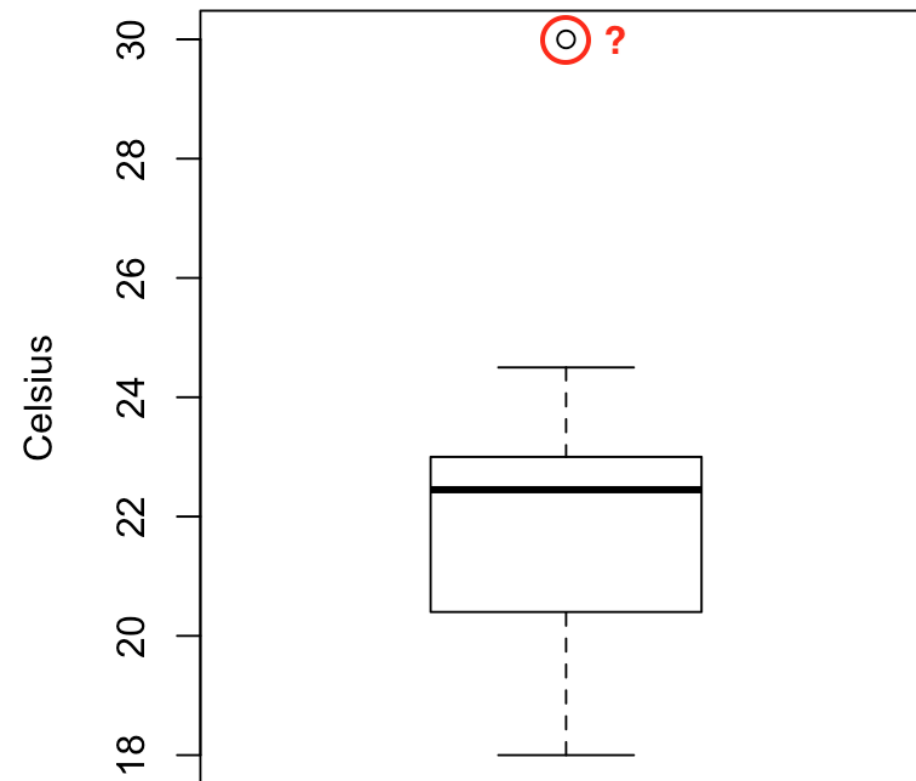
# Testing the extremes with Grubbs' test

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# Visual assessment is not always reliable!

```
boxplot(temperature, ylab = "Celsius")
```







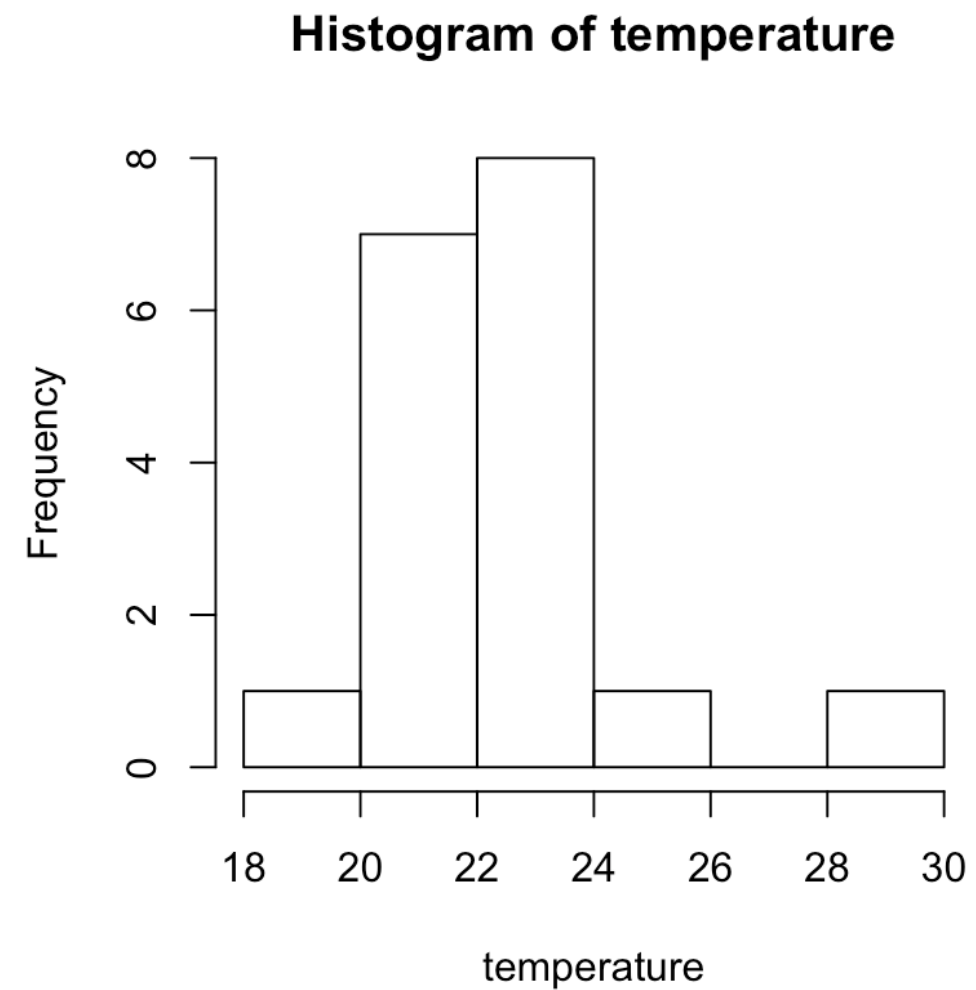
# Grubbs' test

- Statistical test to decide if a point is outlying
- Assumes the data are normally distributed
- Requires checking the normality assumption first



# Checking normality with a histogram

```
hist(temperature, breaks = 6)
```



Symmetrical & bell shaped?



# Running Grubbs' test

Use the `grubbs.test()` function:

```
grubbs.test(temperature)
```

```
Grubbs test for one outlier  
data: temp  
G = 3.07610, U = 0.41065, p-value = 0.001796  
alternative hypothesis: highest value 30 is an outlier
```



# Interpreting the p-value

```
grubbs.test(temperature)
```

```
Grubbs test for one outlier
```

```
data: temperature  
G = 3.07610, U = 0.41065, p-value = 0.001796  
alternative hypothesis: highest value 30 is an outlier
```

p-value

- Near 0 - *stronger* evidence of an outlier
- Near 1 - *weaker* evidence of an outlier



# Get the row index of an outlier

## Location of the **maximum**

```
which.max(weights)
```

```
[1] 5
```

## Location of the **minimum**

```
which.min(temperature)
```

```
[1] 12
```



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# Detecting multiple anomalies in seasonal time series

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# Monthly revenue data

```
head(msales)
```

	sales	month
1	6.068	1
2	5.966	2
3	6.133	3
4	6.230	4
5	6.407	5
6	6.433	6

Grubbs' test not appropriate here

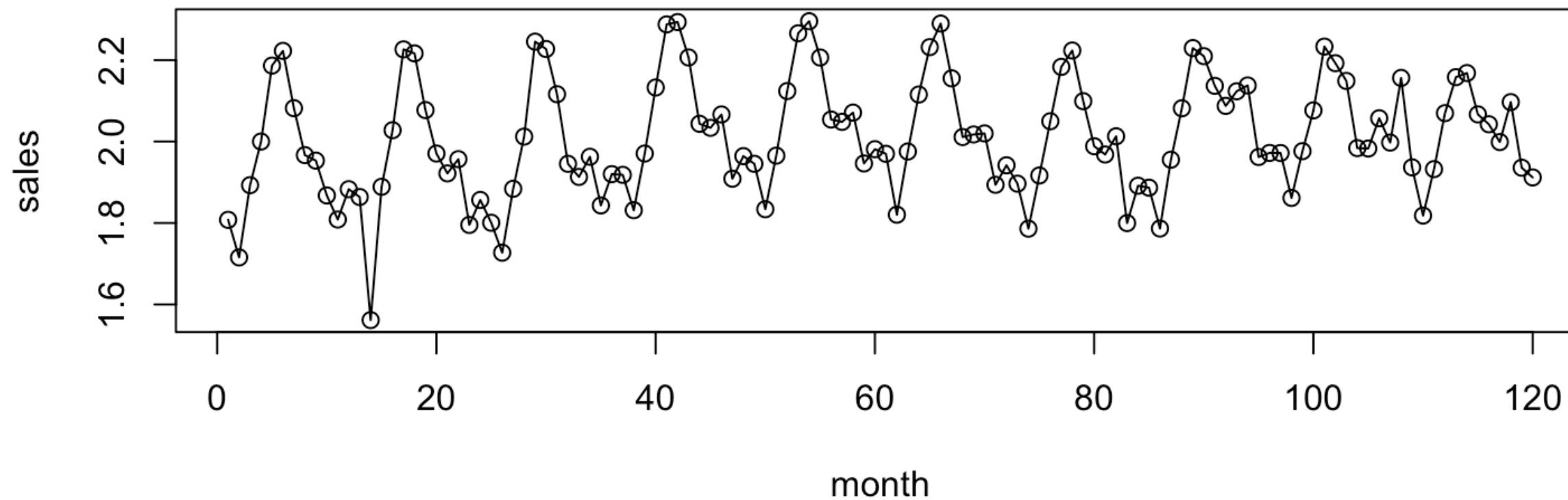
- Seasonality may be present
- May be multiple anomalies





# Visualizing monthly revenue

```
plot(sales ~ month, data = msales, type = 'o')
```





# Seasonal-Hybrid ESD algorithm usage

```
library(AnomalyDetection)

sales_ad <- AnomalyDetectionVec(x = msales$sales, period = 12,
                              direction = 'both')
```

## Arguments

- `x`: vector of values
- `period`: period of repeating pattern
- `direction`: find anomalies that are small, large, or both



# Seasonal-Hybrid ESD algorithm output

```
sales_ad <- AnomalyDetectionVec(x = msales$sales, period = 12,  
                               direction = 'both')
```

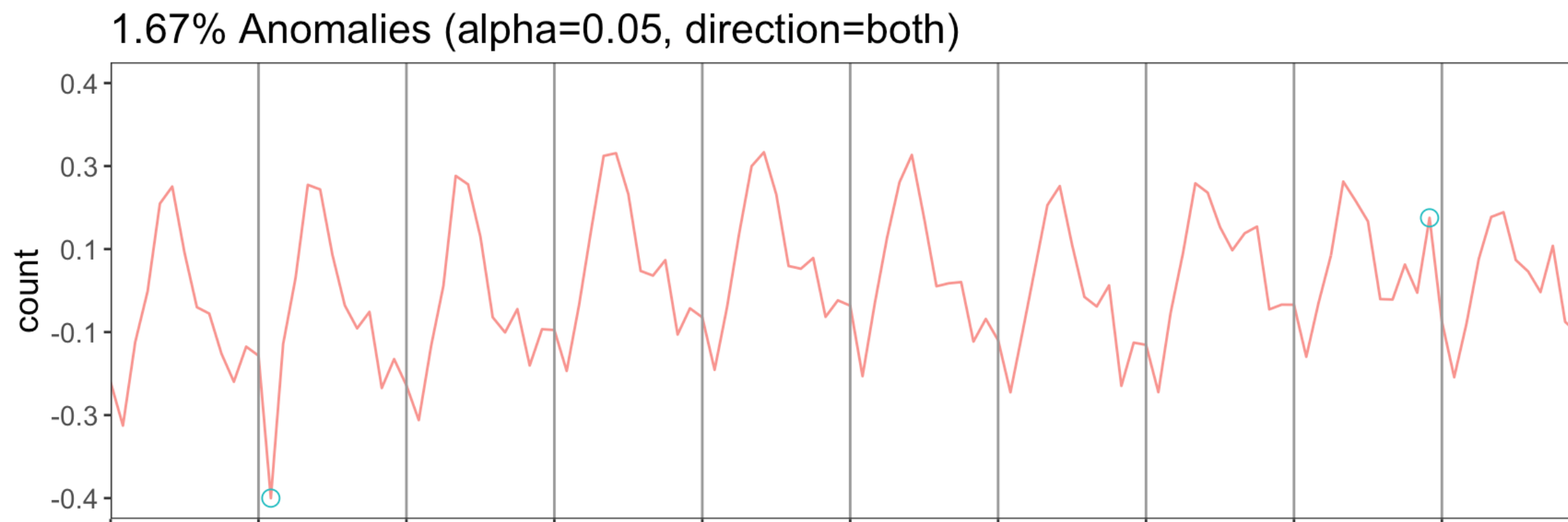
```
sales_ad$anoms
```

	index	anoms
1	14	1.561
2	108	2.156



# Seasonal-Hybrid ESD algorithm plot

```
AnomalyDetectionVec(x = msales$sales, period = 12,  
                    direction = 'both', plot = T)
```





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**Let's practice!**