

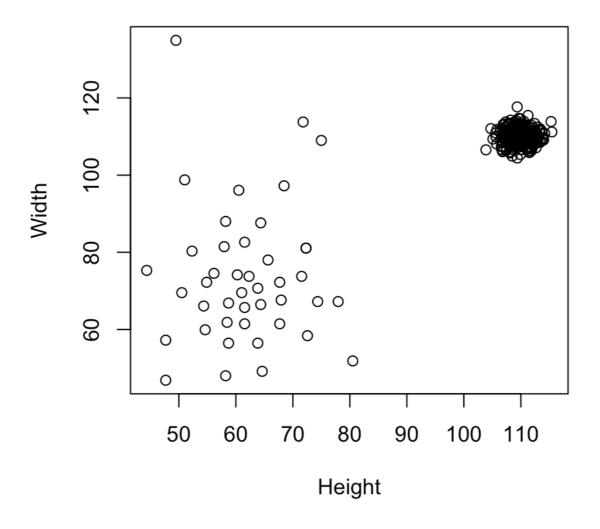


# k-nearest neighbors distance score

Alastair Rushworth
Data Scientist

## Furniture dimensions

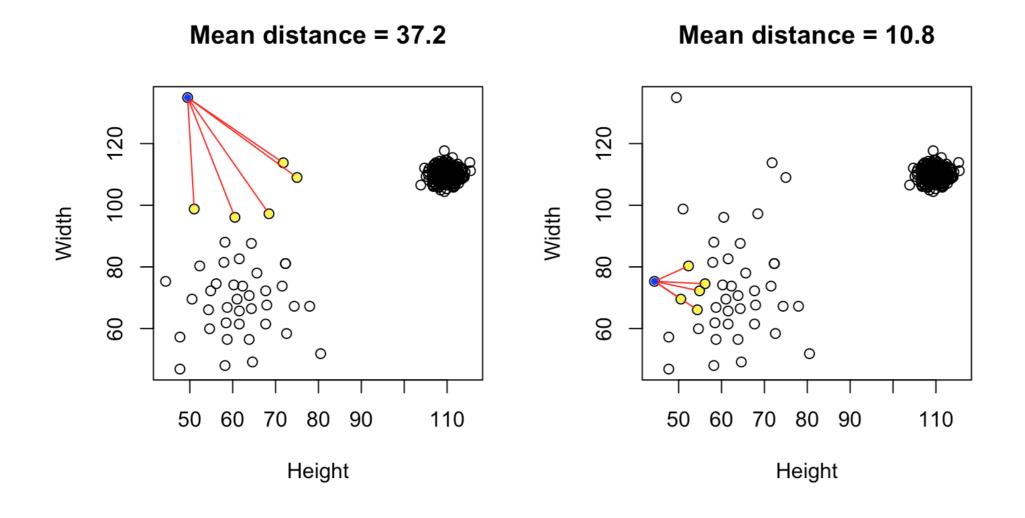
plot(Width ~ Height, data = furniture)





## k-nearest neighbors (kNN) distance

#### **Anomalies usually lie far from their neighbors**



## Inputs for distance matrix calculation

```
library(FNN)
furniture_knn <- get.knn(data = furniture, k = 5)</pre>
```

#### **Arguments**

- data: matrix of data
- k: the number of neighbors

## Distance matrix output

get.knn() returns two matrices

```
names(furniture_knn)
[1] "nn.index" "nn.dist"
```

#### **Distance matrix**

```
head(furniture_knn$nn.dist, 3)
```

```
[,1] [,2] [,3] [,4] [,5]
[1,] 5.128300 5.367791 5.390801 5.740713 8.477025
[2,] 4.300093 5.367791 6.159139 7.091966 7.428176
[3,] 3.047502 3.545978 4.426266 5.006570 5.654202
```



### kNN distance score

Average distance to nearest neighbors

```
furniture_score <- rowMeans(furniture_knn$nn.dist)</pre>
```

#### Largest score?

```
which.max(furniture_score)
[1] 29
```





## Let's practice!



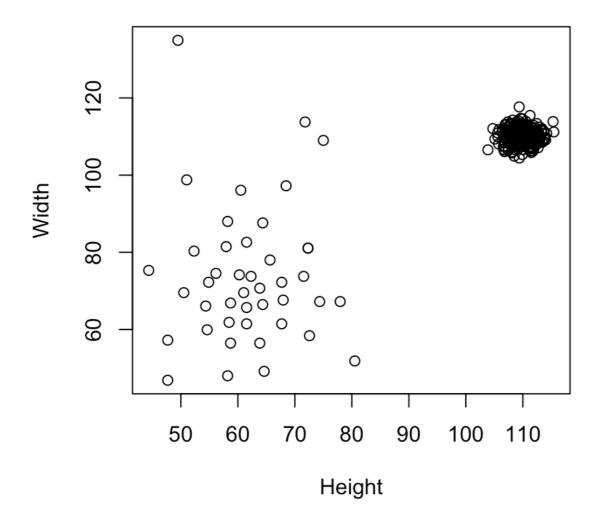


# Visualizing kNN distance score

Alastair Rushworth
Data Scientist

## Standardizing feature scales

plot(Width ~ Height, data = furniture)

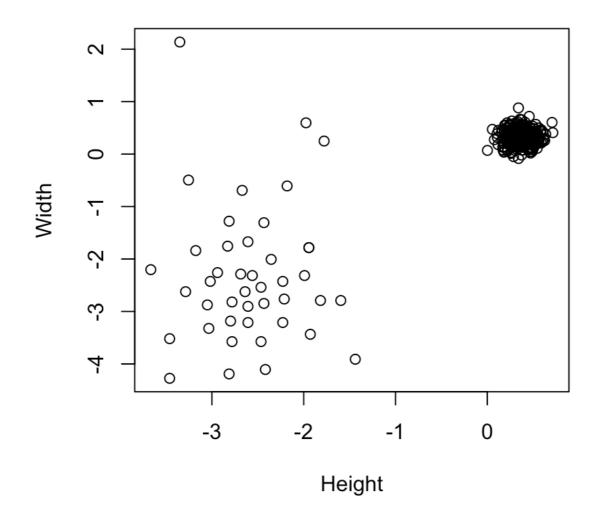




## Standardizing features

```
furniture_scaled <- scale(furniture)</pre>
```

plot(Width ~ Height, data = furniture\_scaled)





### Create and append distance score

#### **Distance matrix**

```
furniture_scaled <- scale(furniture)
furniture_knn <- get.knn(furniture_scaled, 5)</pre>
```

#### Calculate and append score

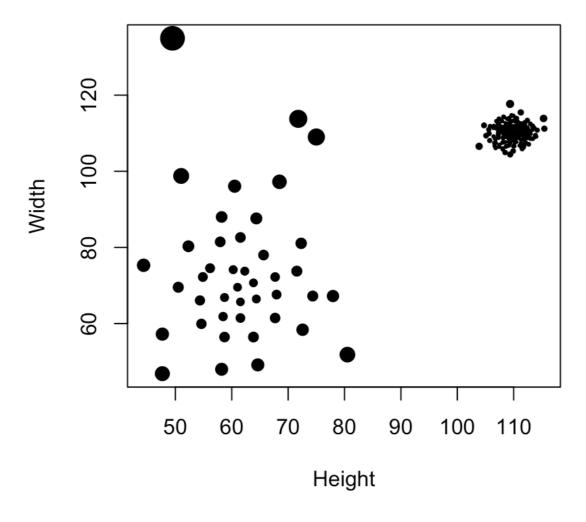
```
furniture$score <- rowMeans(furniture_knn$nn.dist)

head(furniture, 4)

Height Width score
1 58.7179 56.4663 0.4170000
2 54.6154 59.9279 0.3981695
3 58.7179 66.8510 0.2845042
4 63.8462 56.4663 0.4376807
```

## Visualizing distance score

```
plot(Width ~ Height, cex = sqrt(score), data = furniture, pch = 20)
```







## Let's practice!



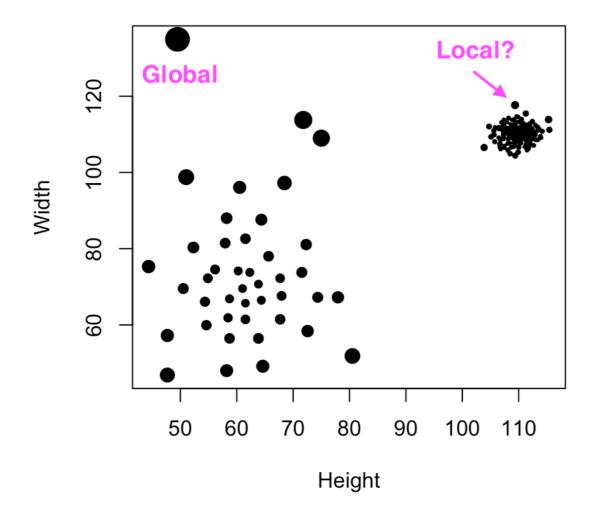


# The local outlier factor (LOF)

Alastair Rushworth
Data Scientist

### Postmortem of kNN distance

#### Global versus local anomalies





## Calculating LOF

#### Obtain LOF for furniture data

```
library(dbscan)
furniture_lof <- lof(scale(furniture), k = 5)</pre>
```

#### View the scores

```
furniture_lof[1:10]

[1] 1.0649669 1.1071205 0.9980290 1.0392385 0.9725305

[6] 1.1933199 1.3210459 1.1409659 1.0613144 1.0805445
```



## Interpreting LOF

#### LOF is a ratio of densities

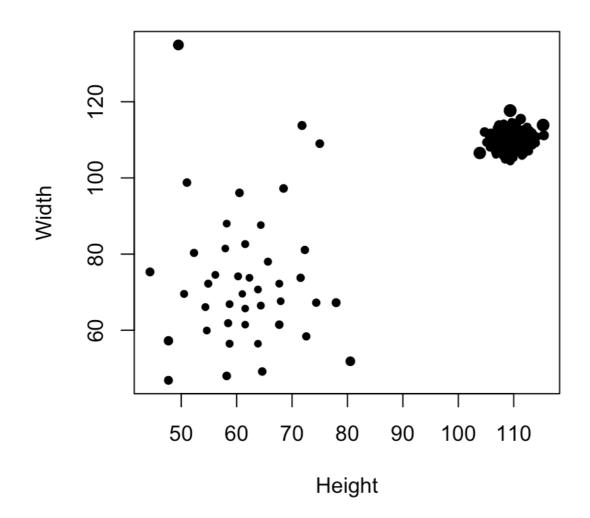
- ullet LOF > 1 more likely to be anomalous
- LOF  $\leq 1$  less likely to be anomalous

#### Large LOF values indicate more isolated points

## Visualizing LOF

```
furniture$score_lof <- furniture_lof</pre>
```

```
plot(Width ~ Height, data = furniture, cex = score_lof, pch = 20)
```







## Let's practice!