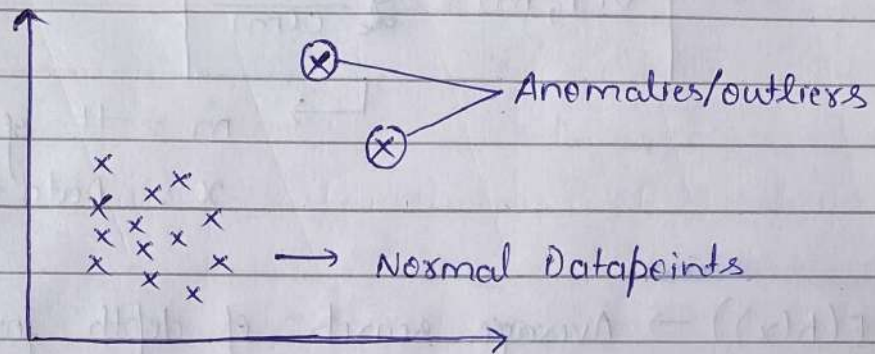


## Anomaly Detection

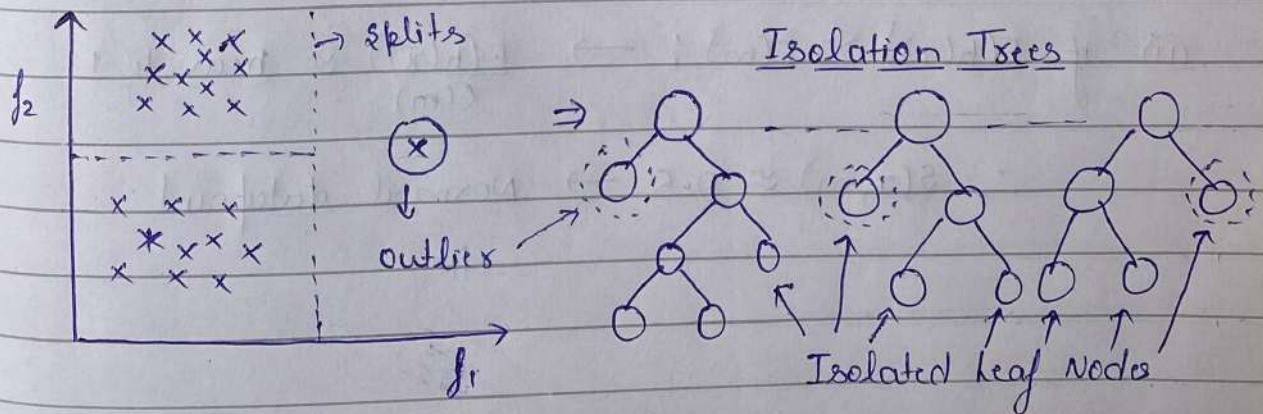
→ It is a technique in machine learning to identify the datapoints that are different from other datapoints.



Note: Anomalies in a dataset can be identified from the following:

- (i) Isolation Forest
- (ii) DBScan
- (iii) Local Outlier Factor

\* Isolation Forest: → Decision Trees



Note: For datapoints we are getting quickly isolated leaf nodes or when depth is small then the probability of those points to be anomalies are high.

→ Anomaly Score : → use to find whether an isolated point is an anomaly or not.

$$S(x, m) = \frac{-E(h(x))}{c(m)}$$

→  $m = \#$  of data points  
 $x =$  Data point

$E(h(x))$  → Average search of depth for  $x$  from the isolated tree.

$c(m)$  → Average depth of all the data points.

• Case :

(i) if  $E(h(x)) \ll c(m)$  : →  $\frac{E(h(x))}{c(m)} \approx 0$

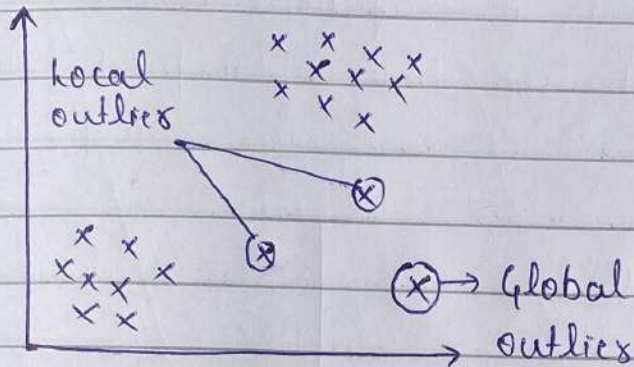
$\therefore S(x, m) \approx 1$  → Anomaly Score → outliers

(ii) if  $E(h(x)) \gg c(m)$  : →  $\frac{E(h(x))}{c(m)} \approx \text{near to } 1$

$\therefore S(x, m) \approx 0.5$  → Normal datapoint

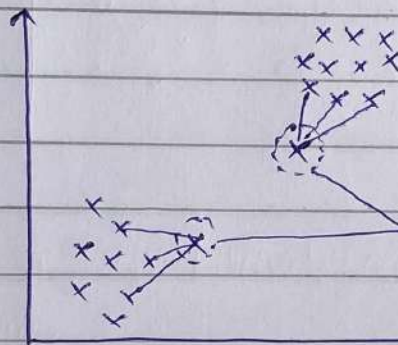


## \* Local Outlier Factor :



LOF

- (i) local outliers
- (ii) global outliers



let  $k = 3$

local density of these points  
is smaller than neighbouring  
3 datapoints.

Note: It considers a datapoint as an outlier/anomaly by comparing the local density of that datapoint to the local densities of its neighbors.

- If sample's/datapoint's local density is smaller then it is an outlier.