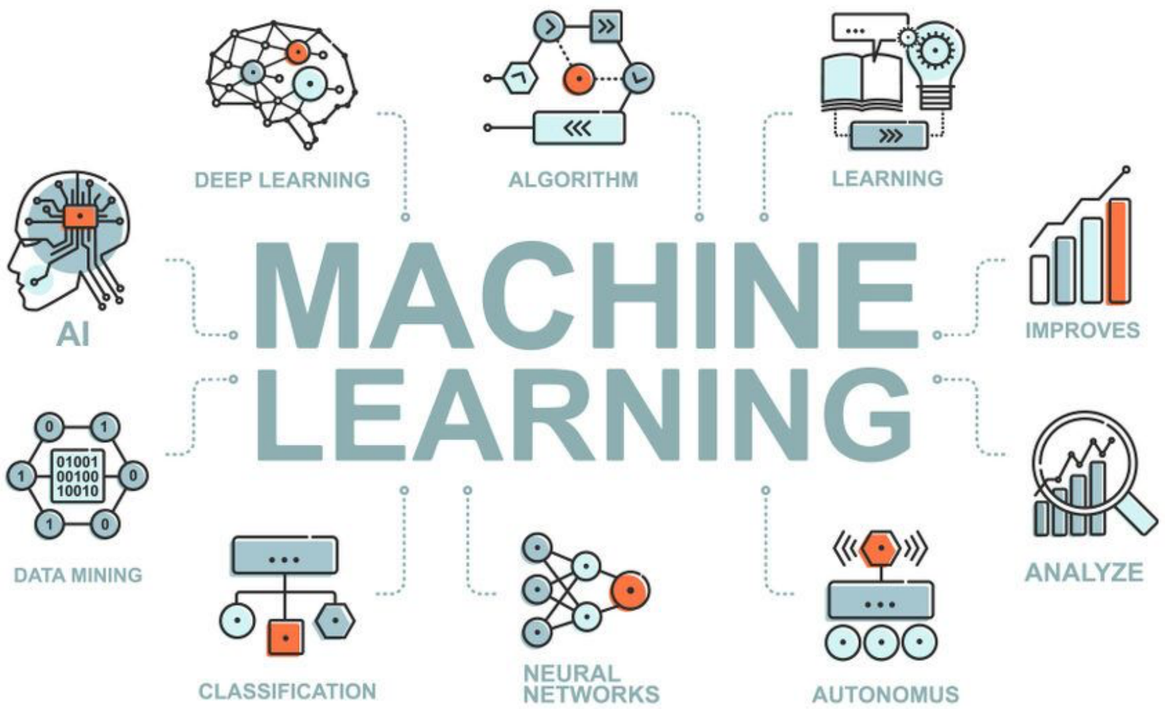
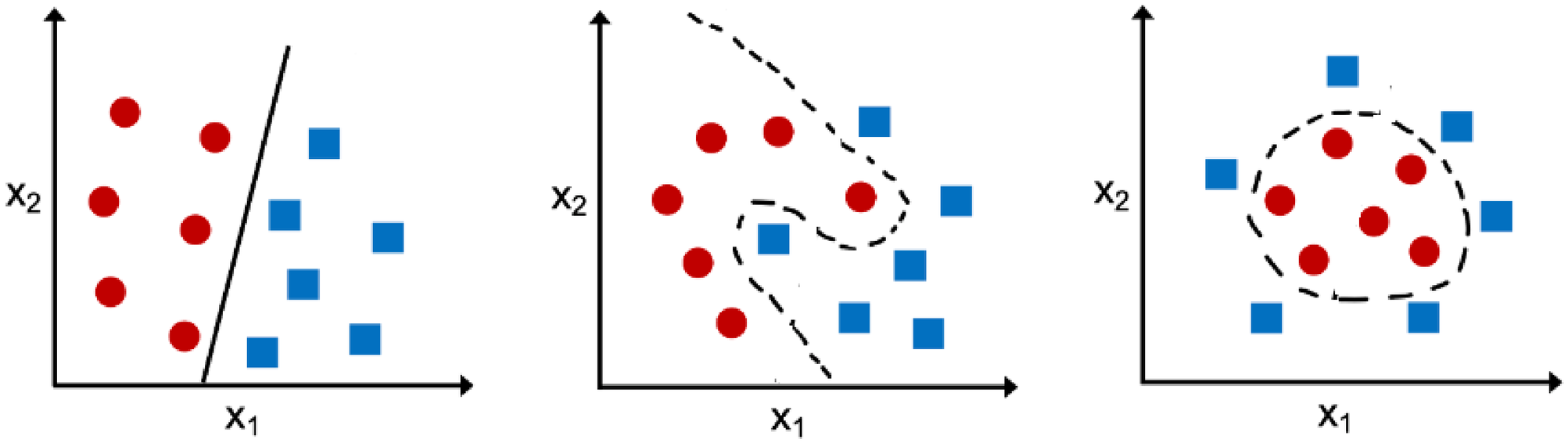


ALAN BADILLO SALAS

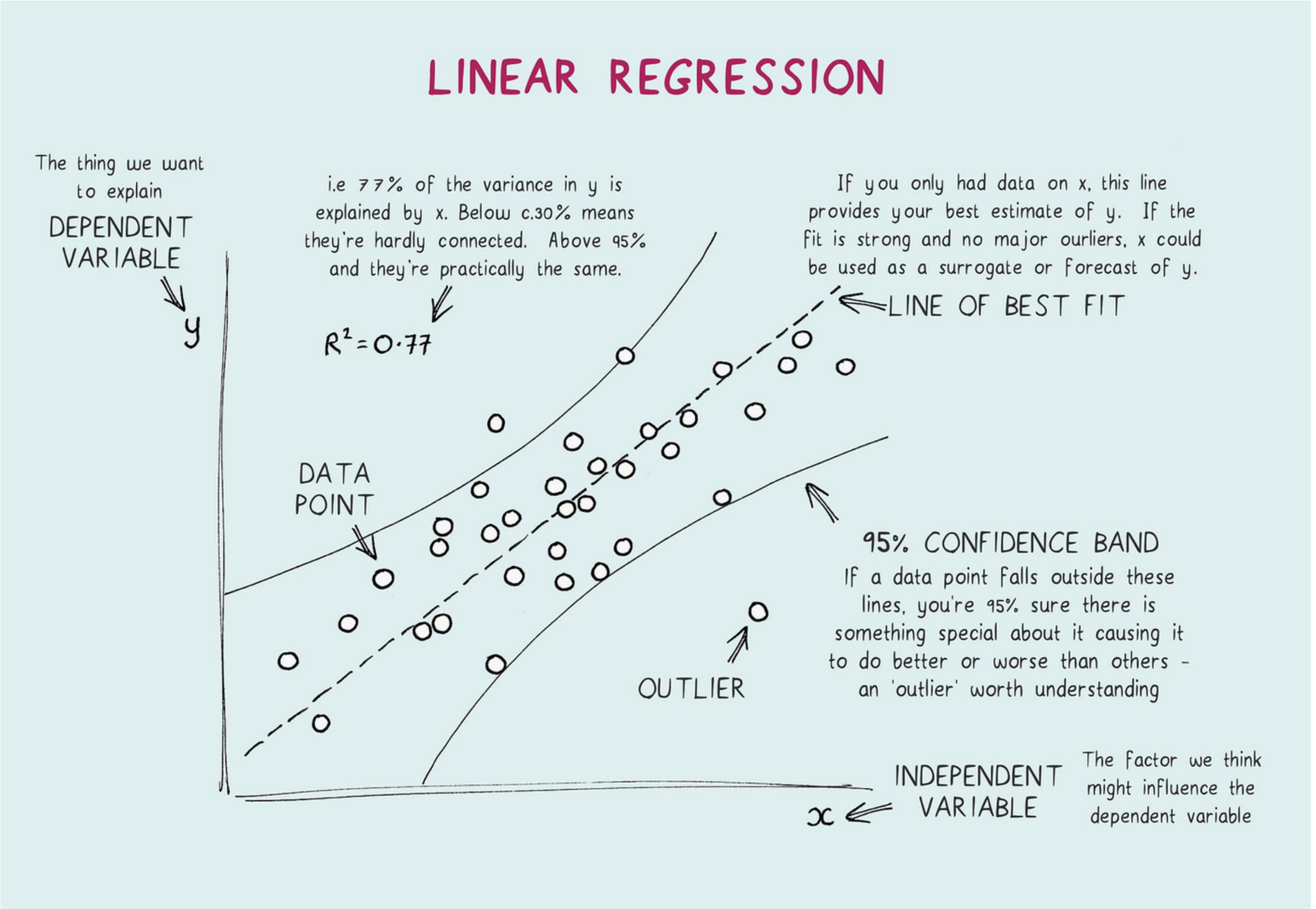
MÓDULO II

OCTUBRE 2023

CLASIFICACIÓN



REGRESIÓN


LINEAR REGRESSION
The thing we want to explain
DEPENDENT VARIABLE
y
i.e 77% of the variance in y is explained by x. Below c.30% means they're hardly connected. Above 95% and they're practically the same.
$R^2 = 0.77$
If you only had data on x, this line provides your best estimate of y. If the fit is strong and no major ourliers, x could be used as a surrogate or forecast of y.
LINE OF BEST FIT
DATA POINT
95% CONFIDENCE BAND
If a data point falls outside these lines, you're 95% sure there is something special about it causing it to do better or worse than others - an 'outlier' worth understanding
OUTLIER
INDEPENDENT VARIABLE
x
The factor we think might influence the dependent variable

REGRESIÓN

Formula

$$R^2 = 1 - \frac{RSS}{TSS}$$

R^2 = coefficient of determination
RSS = sum of squares of residuals
TSS = total sum of squares

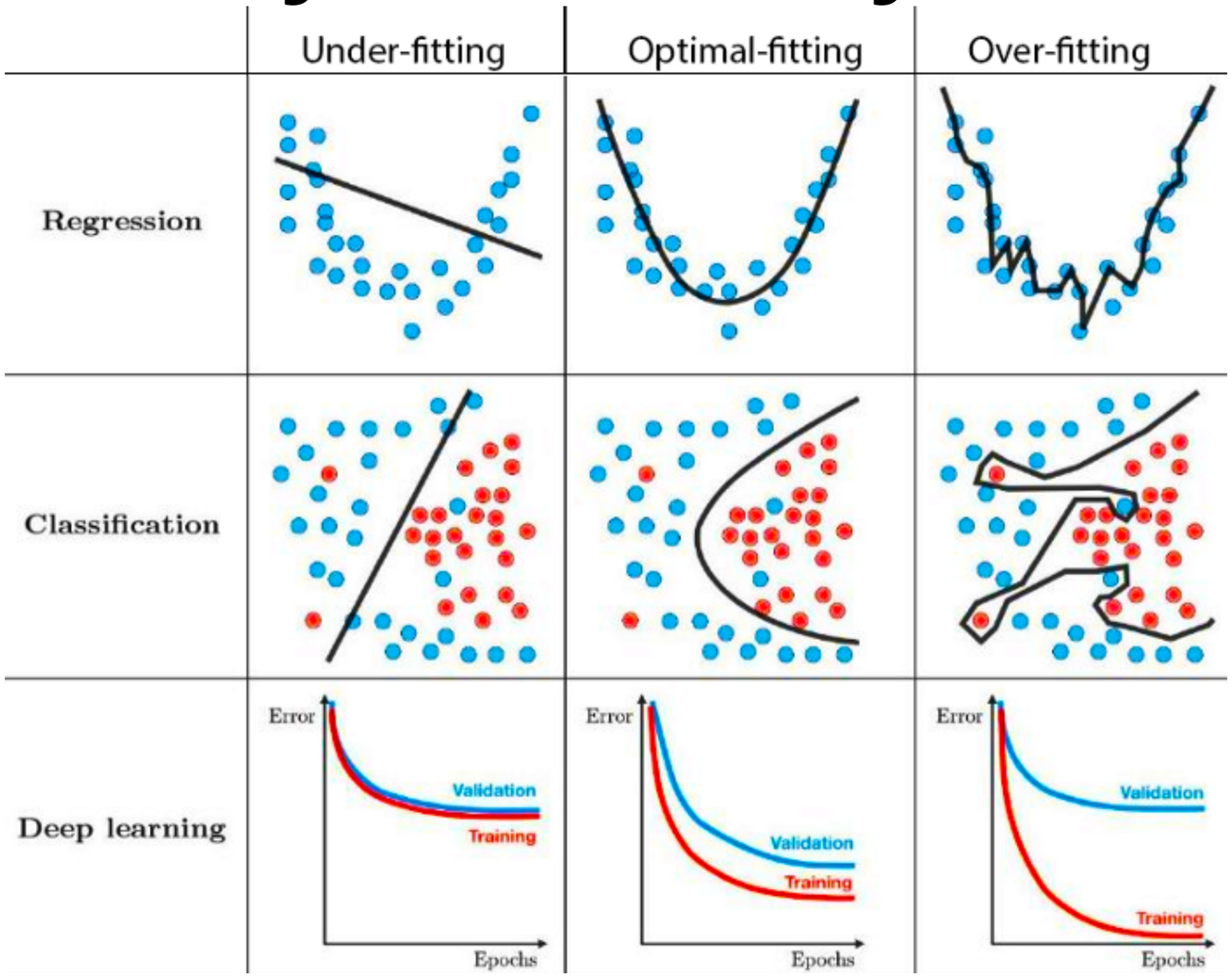
$$\text{RSS} = \Sigma\left(y_i - \widehat{y_i}\right)^2$$

Where: y_i is the actual value and, $\widehat{y_i}$ is the predicted value.

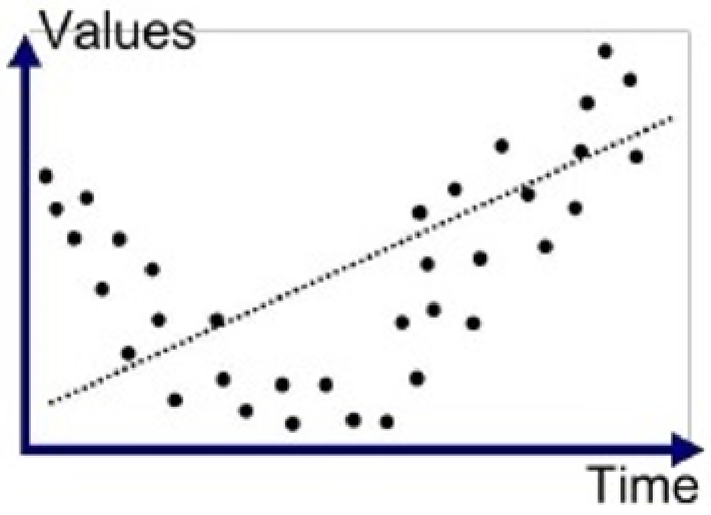
$$\text{TSS} = \Sigma\left(y_i - \overline{y}\right)^2$$

Where: y_i is the actual value and $\overline{y}$ is the mean value of the variable/feature

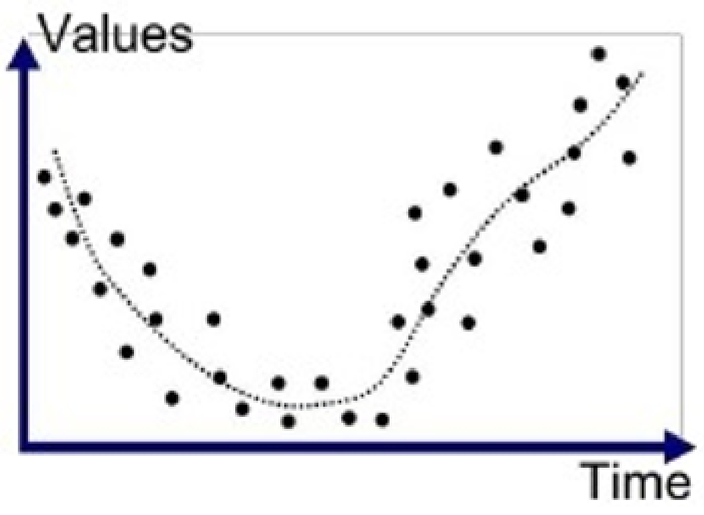
GENERALIZACIÓN, SOBREAJUSTE Y SUBAJUSTE



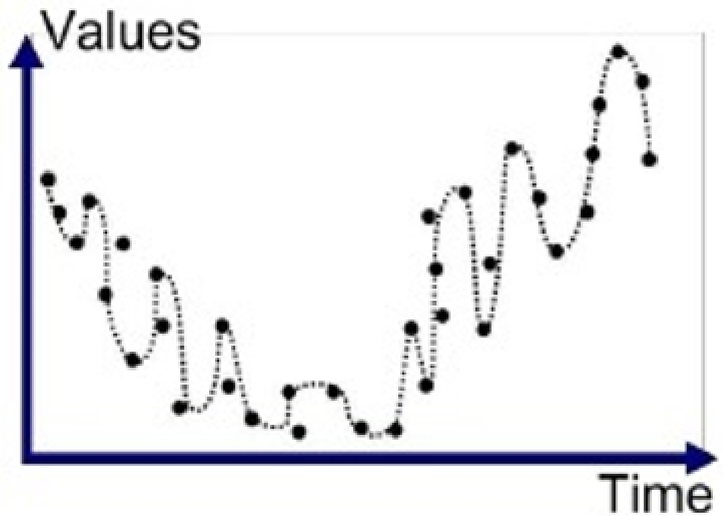
GENERALIZACIÓN, SOBREAJUSTE Y SUBAJUSTE



Underfitted

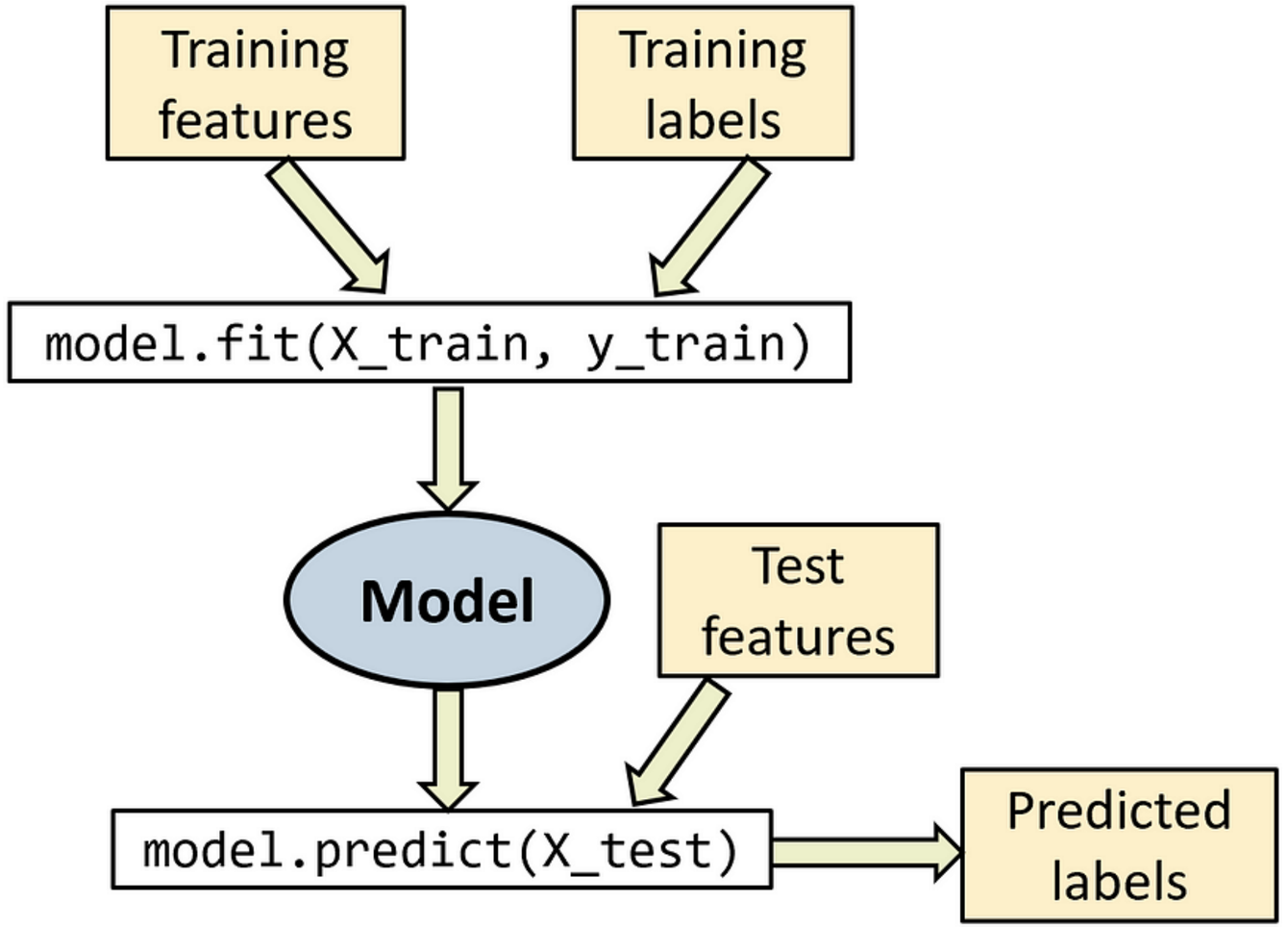


Good Fit/Robust



Overfitted

Algoritmo de Machine Learning Supervisado



Algoritmo de Machine Learning Supervisado

```
from sklearn import linear_model as lm

X = iris[["petal_length"]]
y = iris["petal_width"]

# Fit the linear model
model = lm.LinearRegression()
results = model.fit(X, y)

# Print the coefficients
print model.intercept_, model.coef_
```

```
-0.363075521319 [ 0.41575542]
```

Estimaciones de Incertidumbre