Primeros pasos en Machine Learning y regresión lineal

En breve comenzamos...



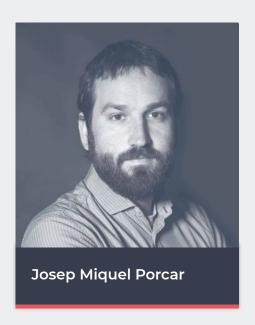
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PONENTE



Data Scientist, matemático y estadístico con varios años de experiencia en proyectos de investigación de Data Science.

Actualmente es **Head Teacher del Data Science Bootcamp de Barcelona en NEOLAND**.

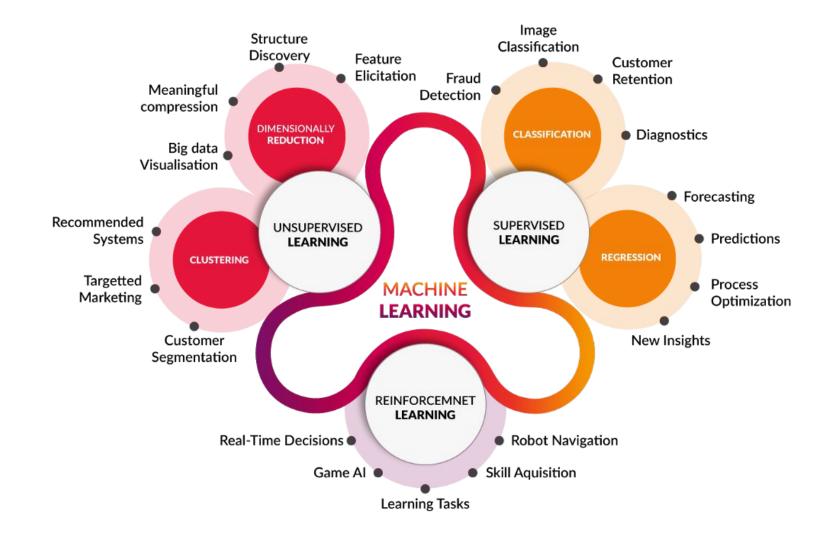
DATA SCIENCE

¿Qué veremos?

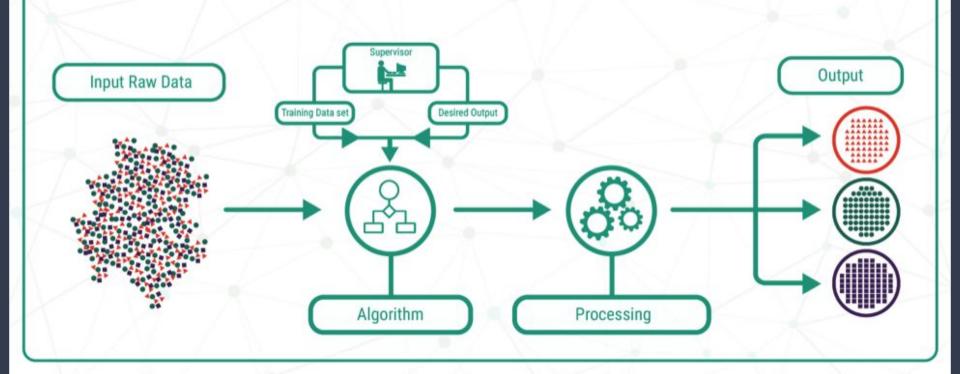
- 1. Statistical learning
- 2. Types of algorithms
- 3. Supervised learning
- 4. Linear Regression

- 5. Features types
- 6. Metrics
- 7. Polynomial regression
- 8. Validation

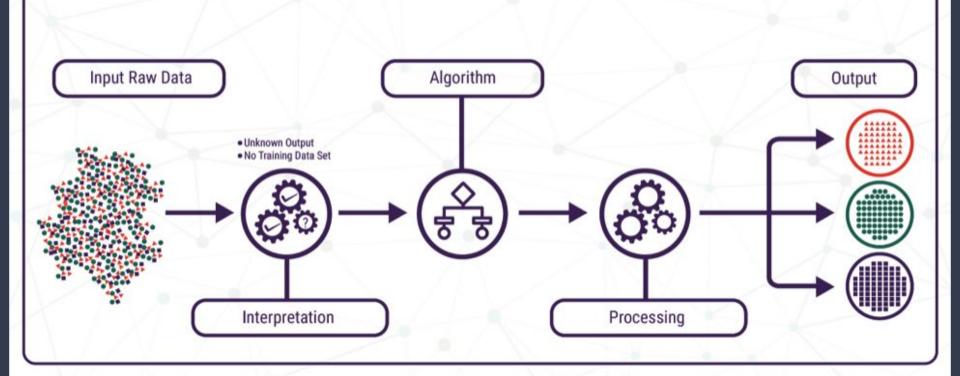
BUSINESS UNDERSTANDING 02 **DATA MINING DATA SCIENCE LIFECYCLE** 06 03 sudeep.co **DATA CLEANING PREDICTIVE** Fix the inconsistencies within the data and handle the missing values. **DATA EXPLORATION FEATURE ENGINEERING** Form hypotheses about your defined problem by visually analyzing the data.



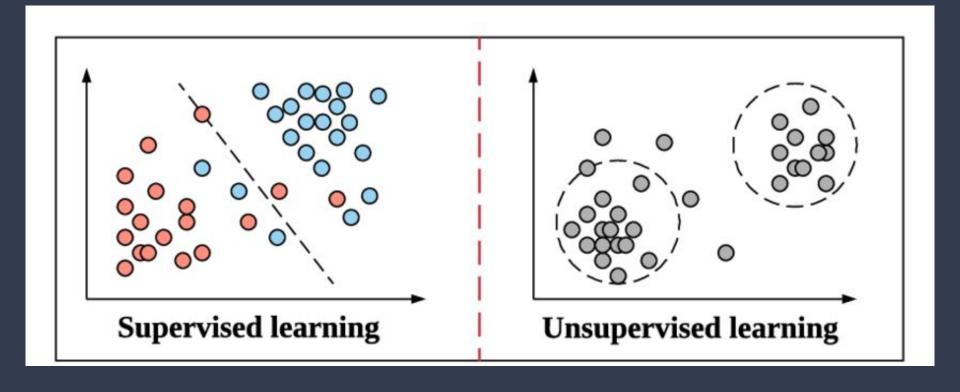
SUPERVISED LEARNING

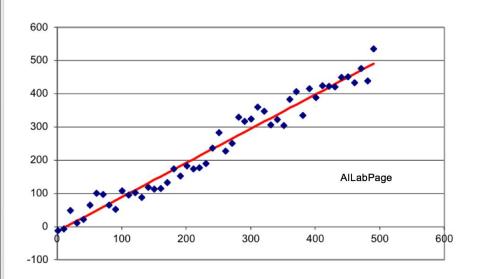


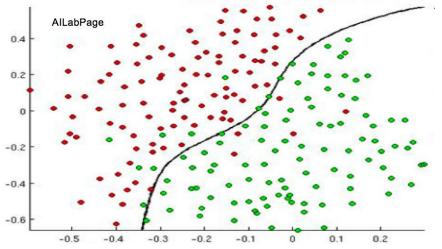
UNSUPERVISED LEARNING



Types of algorithms









Regression

The system attempts to predict a value for an input based on past data.

Example – 1. Temperature for tomorrow



Classification

In classification, predictions are made by classifying them into different categories. Example – 1. Type of cancer 2. Cancer Y/N

Supervised learning

Supervised learning means learning from data:

- We have a quantitative outcome (regression) or categorical outcome (classification)
- We want to predict the *outcome* based on a set of *features* (supervised)
- We have a training set
- We build a prediction model for new unseen objects. The objective is to predict accurately

Vocabulary

Outcome, target, response: Usually denoted by Y

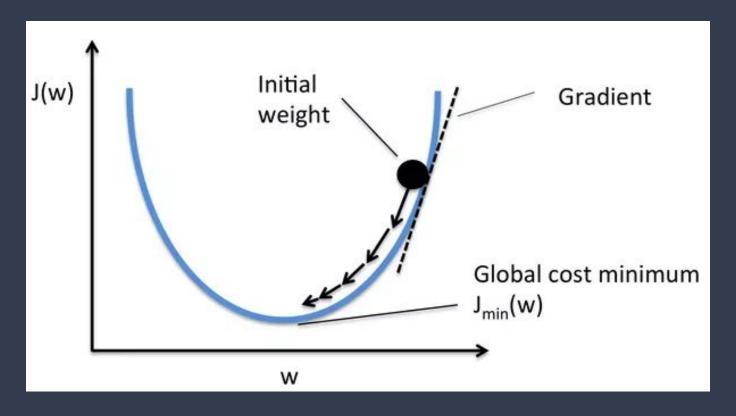
Features, columns, variables: Usually denoted by X (X is a vector of k features)

Training set: (x1, y1), ..., (xn, yn)

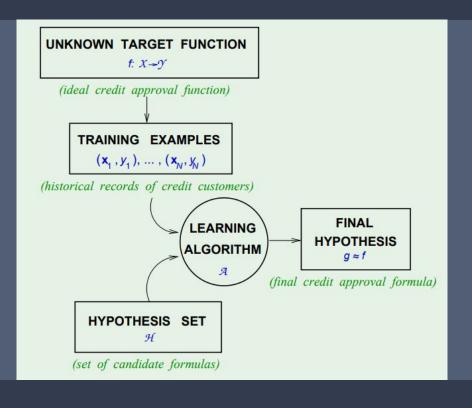
Objective: get a good prediction of Y called $\hat{Y} = f(X)$.

LOSS FUNCTION for penalizing errors (cost function) Squared loss error (Y - f(X))^2

Loss function



Learning process



Linear Regression

A linear regression model assumes that Y is linear in the inputs X:

$$f(X) = \beta_0 + \sum_{j=1}^p X_j \beta_j$$
 $Y = f(X) + \varepsilon$

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Basic assumptions on errors: $\varepsilon \sim N(0, \sigma^2)$

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- Independent
- Mean zero
- Constant variance

Predicting new data

Given a new set of features (X_nuevo), we can predict the outcome as:

$$\hat{Y}_{nuevo} = \hat{\beta} X_{nuevo} = \hat{\beta}_0 + \hat{\beta}_1 X_{1,nuevo} + \dots + \hat{\beta}_p X_{p,nuevo}$$

Features types in Linear regression

Quantitative - Continuous variables:

- Transformations: log, square root...
- Expansions: ^2, ^3, ...
- Interactions: X3 = X1*X2

Qualitative - Categorical variables:

- Dummy coding of the levels. 1 variable with K categories -> K dummy variables

Metrics

$$R^2 = 1 - rac{\Sigma (y - \hat{y})^2}{\Sigma \left(y - ar{y}
ight)^2}$$

RMSE:

$$ext{RMSE} = \sqrt{rac{1}{n}\sum_{i=1}^n (y_i - \hat{y}_i)^2}$$

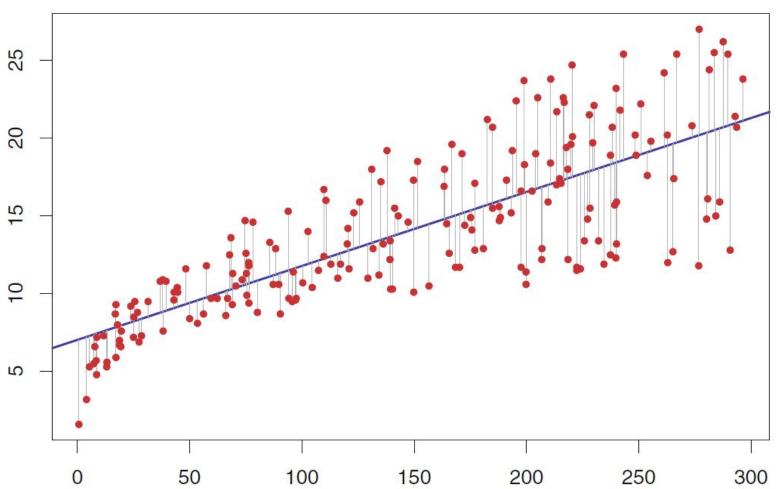
AIC:

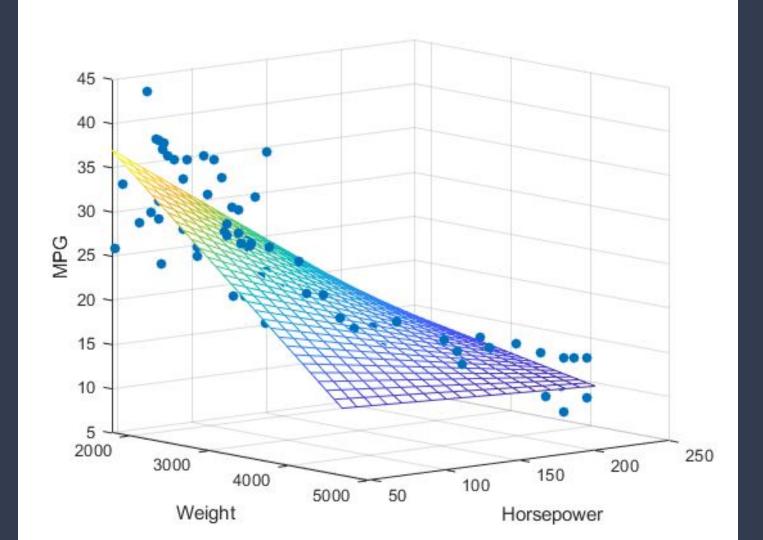
$$AIC = -2logL + 2q$$

BIC:

$$BIC = 2log(L) + qlog(N)$$

3. Linear Regression





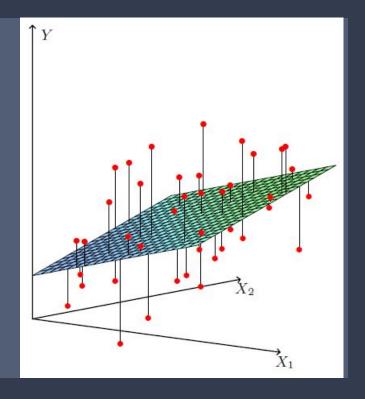
Estimation: Least Squares in LR

$$RSS(\beta) = \sum_{i=1}^{N} (y_i - x_i^T \beta)^2.$$

$$RSS(\beta) = (y - X\beta)^T (y - X\beta)$$

$$\frac{\partial RSS}{\partial \beta} = -2X^T (y - X\beta)$$

$$\hat{\beta} = (X^T X)^{-1} X^T y,$$



Interpreting estimators

$$\hat{Y} \,=\, \hat{\beta}_0 + \hat{\beta}_1 X_1 + \hat{\beta}_2 D_1 + \hat{\beta}_3 D_2$$

X1 is a continuous variable:

- sign
- size
- marginal effect

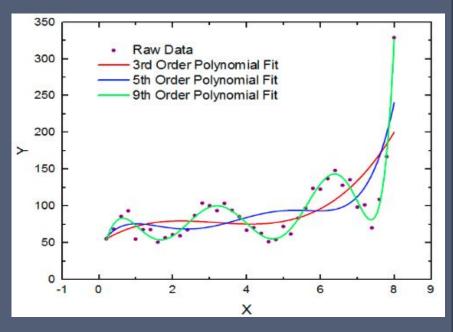
D1, D2 are the dummies of a categorical variable with 3 levels:

- reference category D3

Polynomial regression

f(x) is a polynomial of order k:

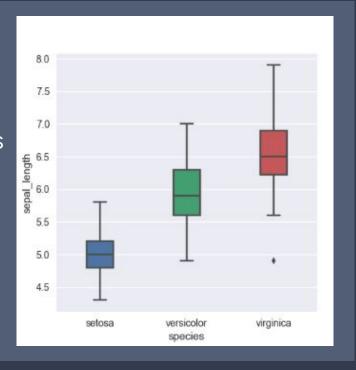
$$f(X) = \sum_{j=0}^{k} \beta_j X^j$$



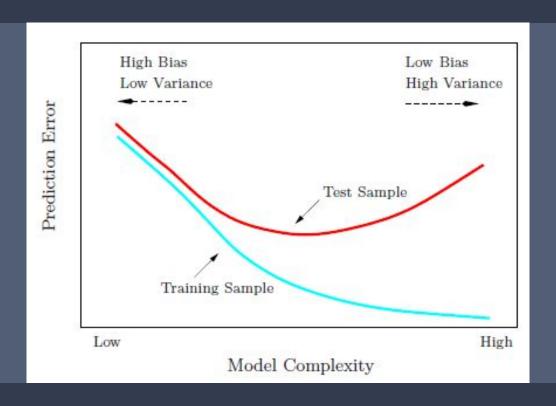
Outliers

In words of Hawkins, 1980:

"An outlier is an observation which deviates so much from the other observations as to arouse suspicions that it was generated by a different mechanism"



Validation





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Más...

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