

Intro Machine Learning

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¿Que es AI?

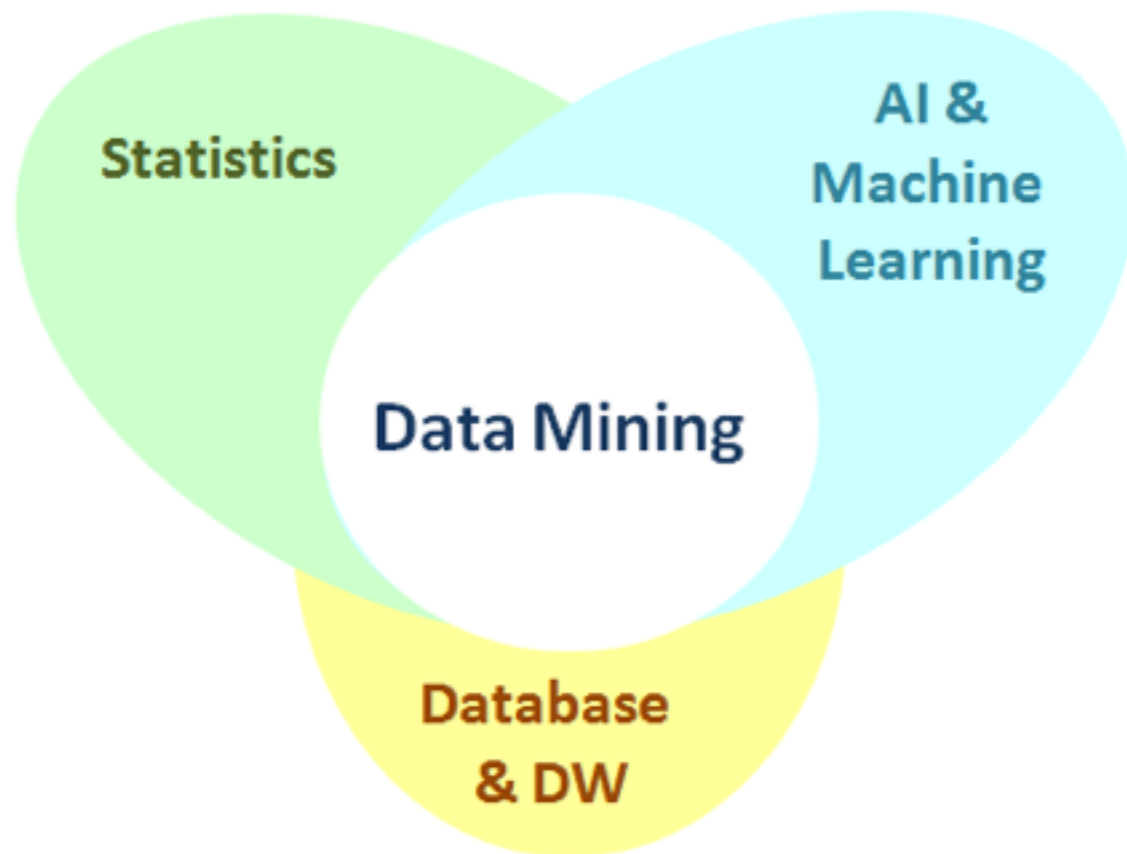
"AI is the study of making computers do things, at which, at the moment, people are better"

-- Elaine Rich (author of Artificial Intelligence book)

Data Mining

- ☹ No es fácil definirlo
- ☹ Machine Learning es una parte
- ☹ Conviene conocer todo el entorno
- ☹ Data Mining intenta explicar el pasado y predecir el futuro.

Data Mining



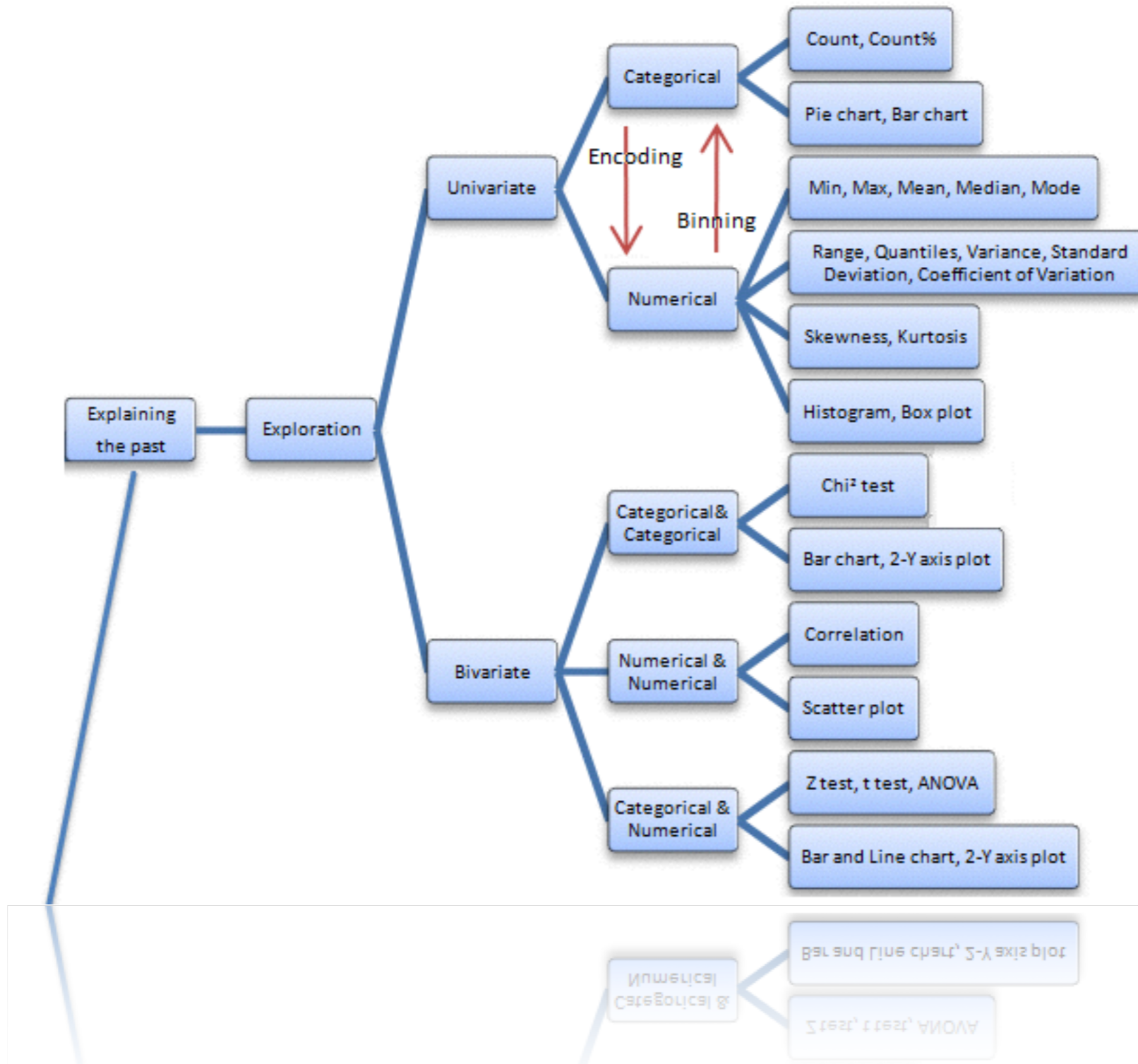
- ☺ Estadísticas
- ☺ AI & Machine Learning
- ☺ Database & Datawarehouse

Data Mining

- ☺ Podemos dividirlo en dos subcategorias:
 - ☺ Explicando el Pasado
 - ☺ Prediciendo el Futuro
 - ☺ Incluso una tercera Prescribiendolo ;-)

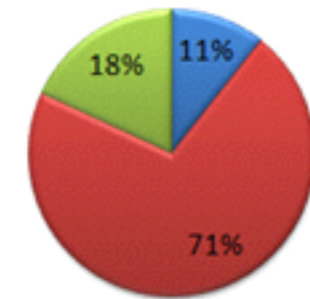
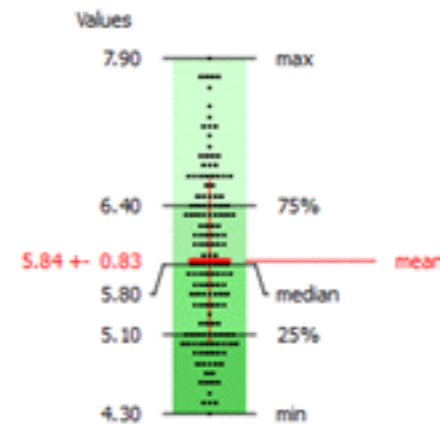
Data Mining

Explicando el Pasado

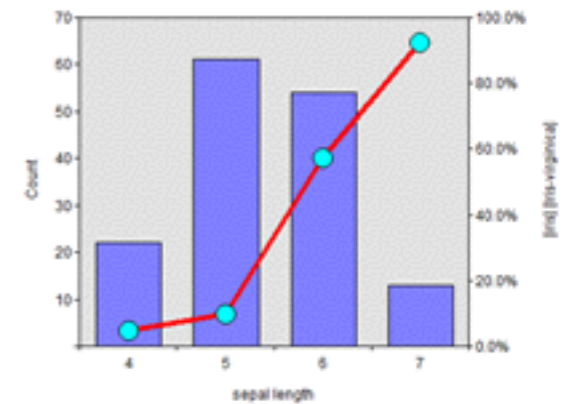
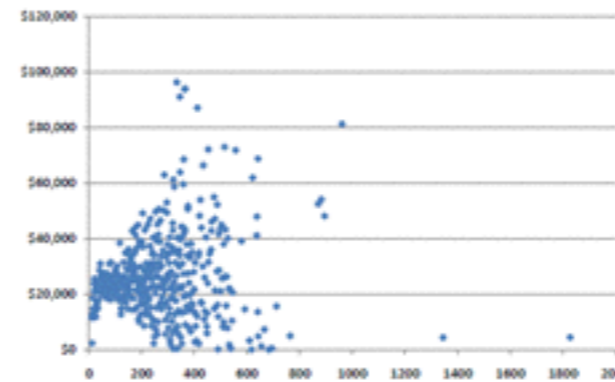


Explorando los datos

☹️ 1 Variable



☹️ 2 Variables



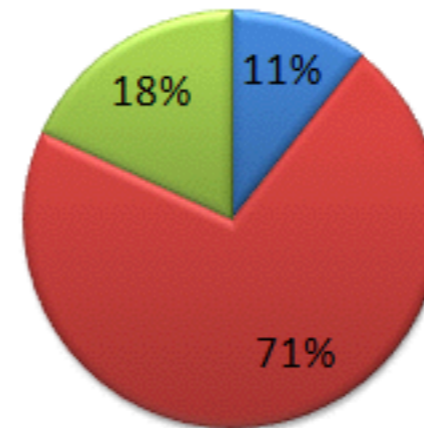
1 Variable

- ☺ Exploramos las variables (o atributos) de una en una.
 - ☺ Categorías (nominales, ordenadas)
 - ☺ Numéricas

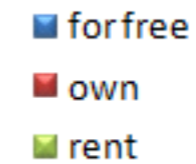
Categorias

Frequency Table

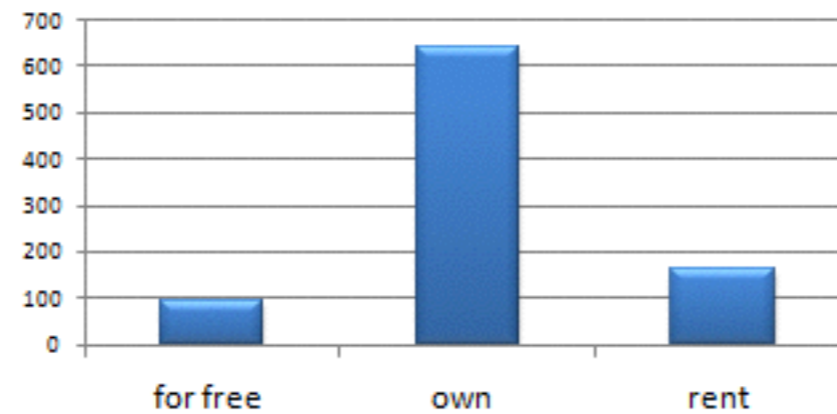
Housing	Count	Count%
for free	96	10.67%
own	641	71.22%
rent	163	18.11%



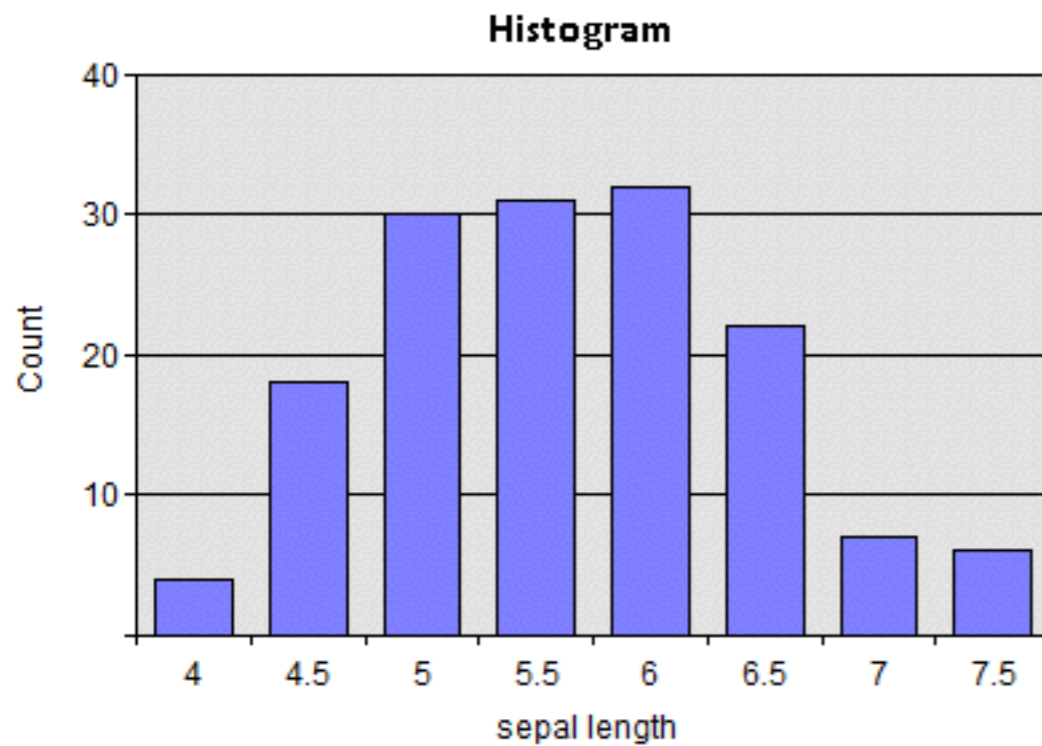
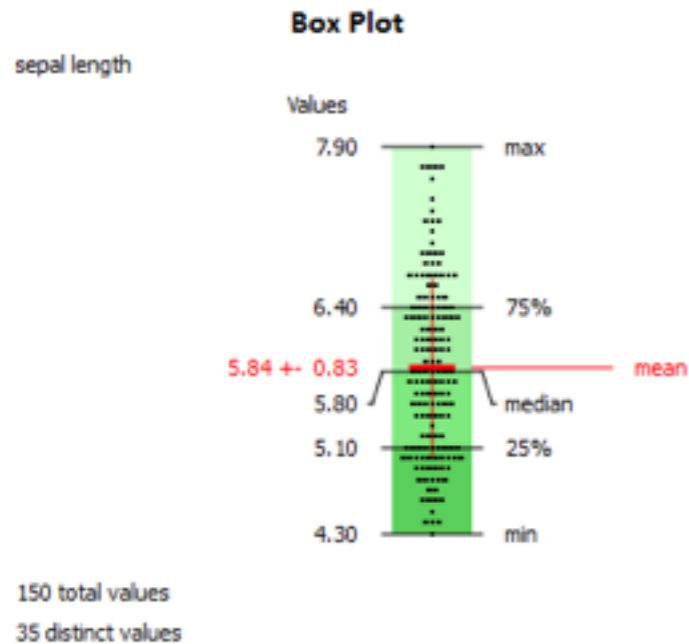
Housing



Housing



Numéricas

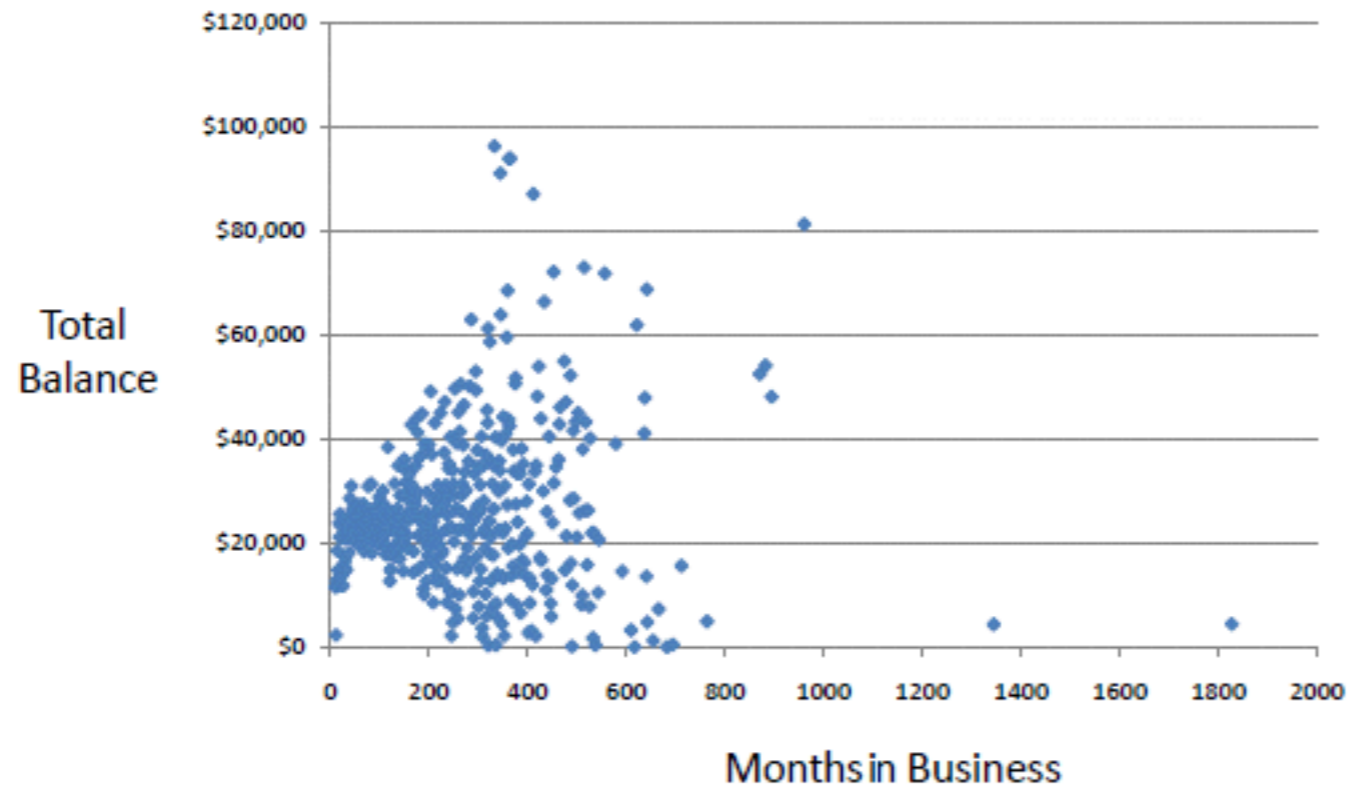


sepal length	
Count	150
Minimum	4.3
Maximum	7.9
Mean	5.84
Median	5.8
Mode	5
Quartile 1	5.1
Range	3.6
Variance	0.69
Standard Deviation	0.83
Coefficient of Variation	14.2%
Skewness	0.31
Kurtosis	-0.55

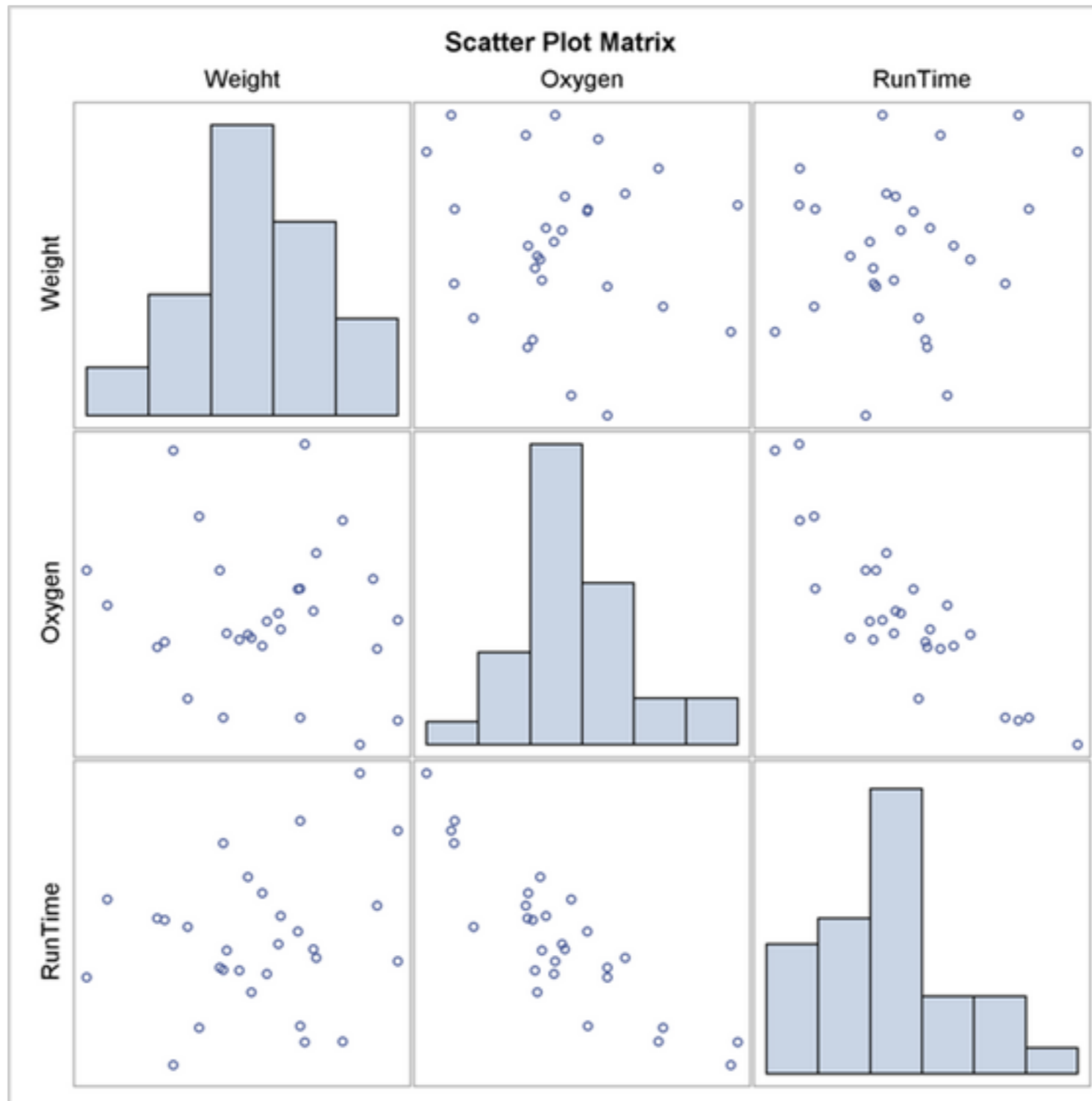
2 Variables

- ☺ Números vs Números
- ☺ Categorias vs Categorias
- ☺ Números vs Categorias

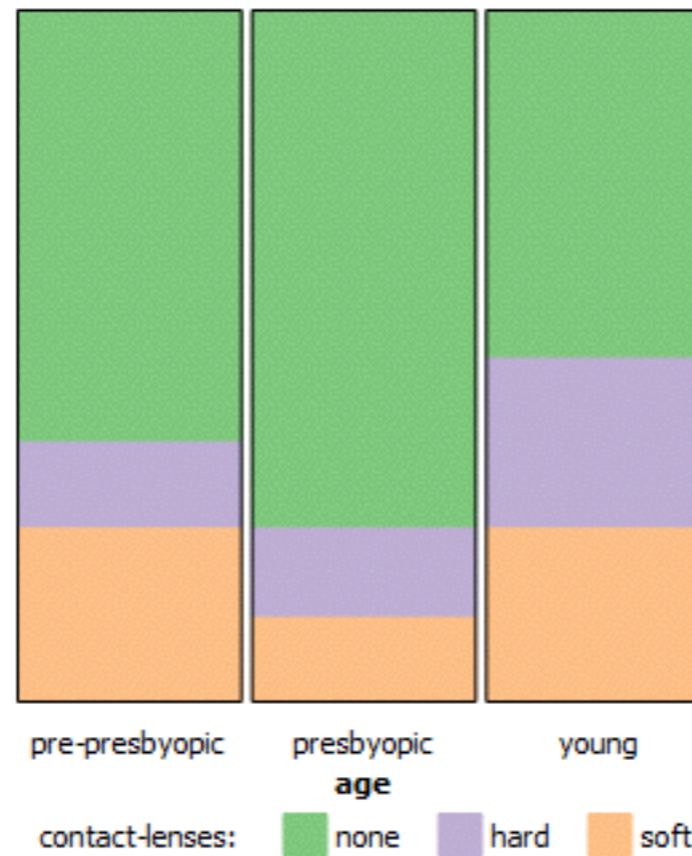
Númericas vs Númericas



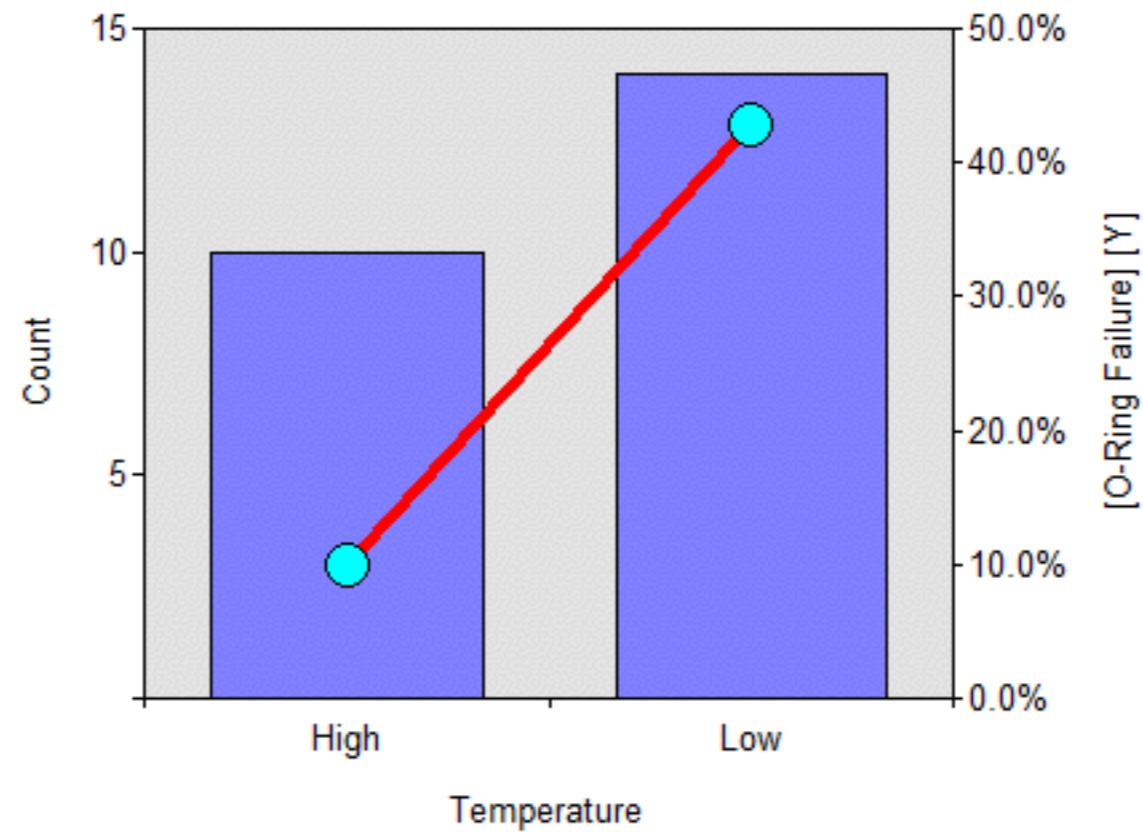
Númericas vs Númericas



C vs C: Stacked Plot



C vs C: Combination Chart



C vs C: Chi Square

$$\chi^2 = \sum_{i=1}^r \sum_{j=1}^c \frac{(n_{ij} - e_{ij})^2}{e_{ij}}$$

Tchouproff Contingency Coefficient

$$\rho_c = \sqrt{\frac{\chi^2}{n \sqrt{(c-1)(r-1)}}$$

$$e_{ij} = \frac{n_i \cdot n_j}{n}$$

$$df = (r-1)(c-1)$$

Hair

		Light	Dark	
<i>Eye</i>	Black	32 (24.1)	12 (19.9)	44
	Green/Blue	14 (19.7)	22 (16.3)	36
	Others	6 (8.2)	9 (6.8)	15
		52	43	95

e=(44 * 52)/95=24.1

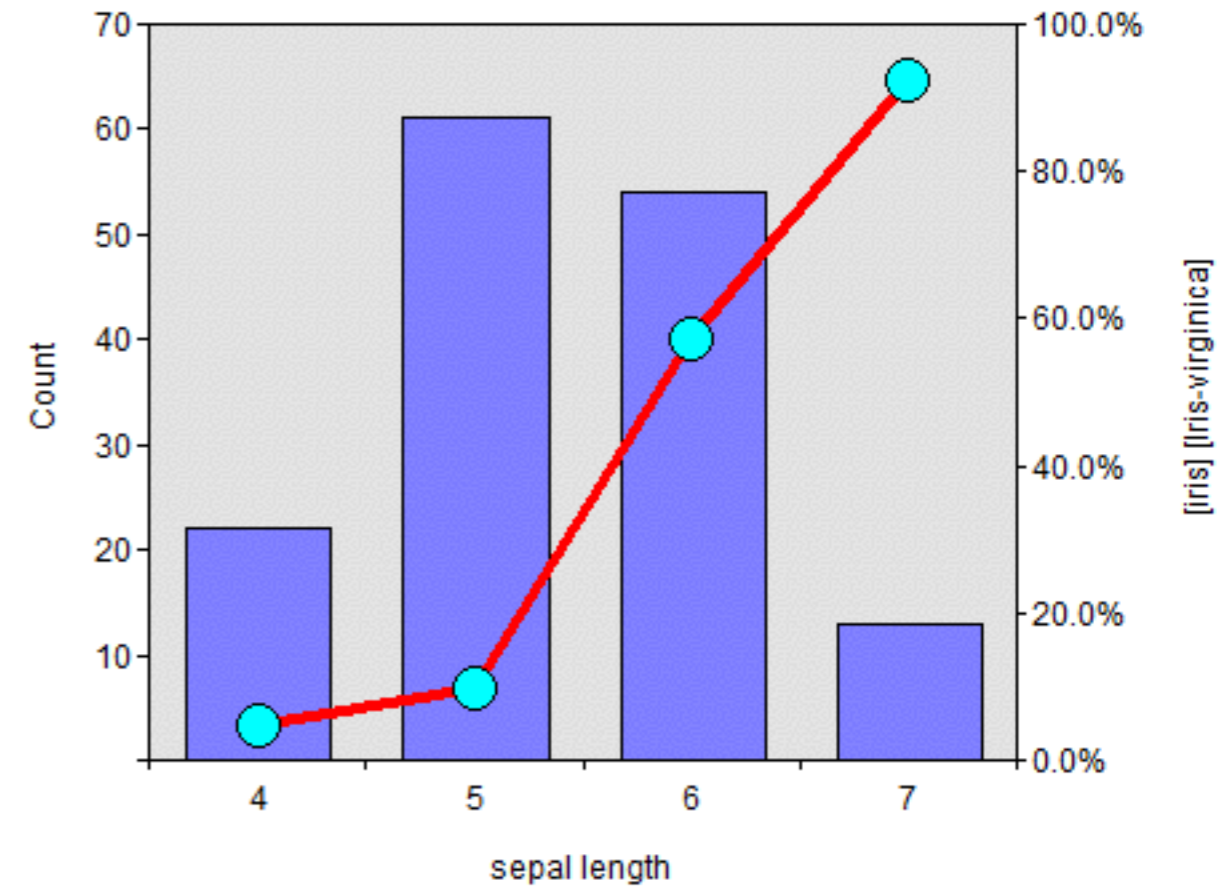
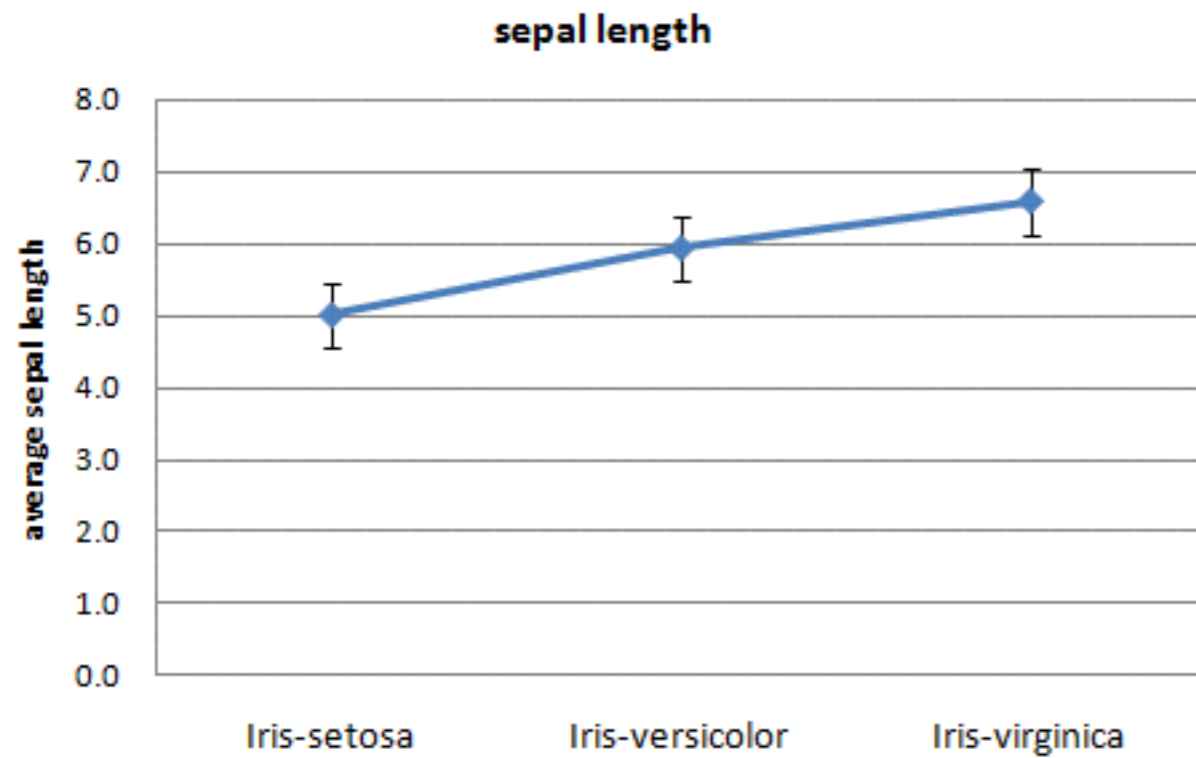
$$\chi^2 = 10.67$$

$$df = (r-1)(c-1) = (3-1)(2-1) = 2$$

$$p = 0.005$$

$$\rho_c = \sqrt{\frac{10.67}{95 \sqrt{(3-1)(2-1)}}} = 0.28$$

Categorías vs Numéricas



Categorías vs Numéricas

- ☹ Y además no podían faltar:
- ☹ El Z-test y t-test
- ☹ Análisis de Varianza (ANOVA)

Prediciendo el Futuro



Prediciendo el Futuro

- ☺ Generamos Modelos
 - ☺ Clasificación: resultado categoría
 - ☺ Regresión: resultado numérico
 - ☺ Clustering
 - ☺ Reglas Asociativas

Prediciendo el Futuro

- ☺ La Clasificaciones las podemos dividir en grupos, basados:
 - ☺ Tabla de Frecuencias (ZeroR, OneR, Naive Bayesian, Decision Tree)
 - ☺ Matriz de Covariancia (A.Linear Discriminante, Regresión Logística)

Prediciendo el Futuro

- ☺ La Clasificaciones las podemos dividir en grupos, basados en:
 - ☺ Funciones de similitud: KNN (K vecinos más cercanos)
 - ☺ Otros: ANN (Redes Neuronales Artificiales), SVM (Support Vector Machine)

Zero R

- ☹ La más simple.
- ☹ De hecho ignora las variables

Predictors				Target
Outlook	Temp.	Humidity	Windy	Play Golf
Rainy	Hot	High	False	No
Rainy	Hot	High	True	No
Overcast	Hot	High	False	Yes
Sunny	Mild	High	False	Yes
Sunny	Cool	Normal	False	Yes
Sunny	Cool	Normal	True	No
Overcast	Cool	Normal	True	Yes
Rainy	Mild	High	False	No
Rainy	Cool	Normal	False	Yes
Sunny	Mild	Normal	False	Yes
Rainy	Mild	Normal	True	Yes
Overcast	Mild	High	True	Yes
Overcast	Hot	Normal	False	Yes
Sunny	Mild	High	True	No

OneR

- ☹ Se generan los modelos para cada variable
- ☹ Se escoge el que menos error tiene

Which one is the best predictor ?

Outlook	Temp	Humidity	Windy	Play Golf
Rainy	Hot	High	False	No
Rainy	Hot	High	True	No
Overcast	Hot	High	False	Yes
Sunny	Mild	High	False	Yes
Sunny	Cool	Normal	False	Yes
Sunny	Cool	Normal	True	No
Overcast	Cool	Normal	True	Yes
Rainy	Mild	High	False	No
Rainy	Cool	Normal	False	Yes
Sunny	Mild	Normal	False	Yes
Rainy	Mild	Normal	True	Yes
Overcast	Mild	High	True	Yes
Overcast	Hot	Normal	False	Yes
Sunny	Mild	High	True	No

Frequency Tables

★ Outlook		Play Golf	
		Yes	No
Sunny		3	2
Overcast		4	0
Rainy		2	3

Temp.		Play Golf	
		Yes	No
Hot		2	2
Mild		4	2
Cool		3	1

Humidity		Play Golf	
		Yes	No
High		3	4
Normal		6	1

Windy		Play Golf	
		Yes	No
False		6	2
True		3	3

Naive Bayesian

☺ Basado en el teorema de Bayes

☺ Mucho más sencillo de construir

$$P(c|x) = \frac{P(x|c)P(c)}{P(x)}$$

Likelihood: $P(x|c)$
 Class Prior Probability: $P(c)$
 Posterior Probability: $P(c|x)$
 Predictor Prior Probability: $P(x)$

$$P(c|X) = P(x_1|c) \times P(x_2|c) \times \dots \times P(x_n|c) \times P(c)$$

Frequency Table		Play Golf	
		Yes	No
Outlook	Sunny	3	2
	Overcast	4	0
	Rainy	2	3



Likelihood Table		Play Golf		
		Yes	No	
Outlook	Sunny	3/9	2/5	5/14
	Overcast	4/9	0/5	4/14
	Rainy	2/9	3/5	5/14
		9/14	5/14	

$$P(x|c) = P(\text{Sunny} | \text{Yes}) = 3/9 = 0.33$$

$$P(x) = P(\text{Sunny}) = 5/14 = 0.36$$

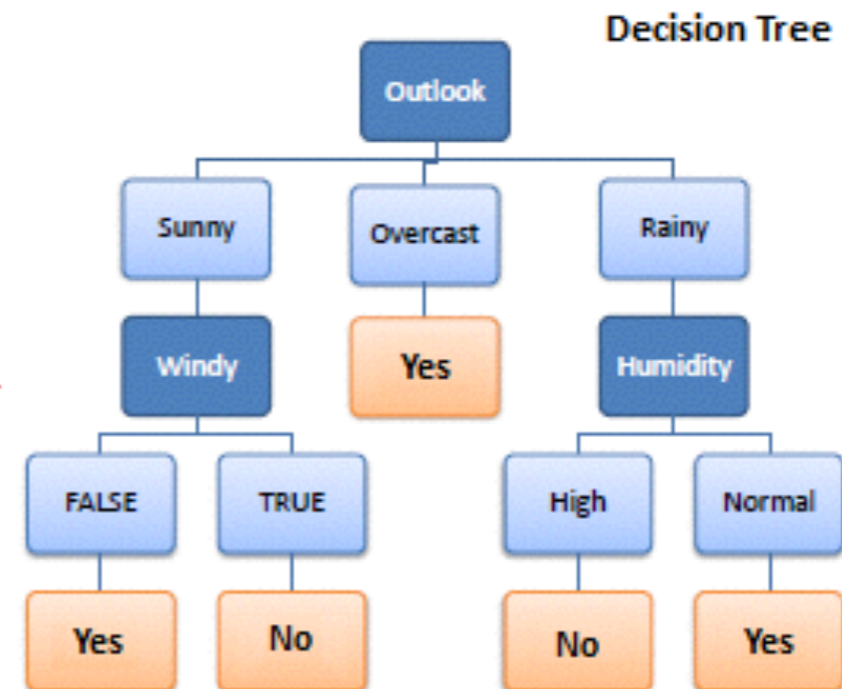
$$P(c) = P(\text{Yes}) = 9/14 = 0.64$$

Posterior Probability: $P(c|x) = P(\text{Yes} | \text{Sunny}) = 0.33 \times 0.64 \div 0.36 = 0.60$

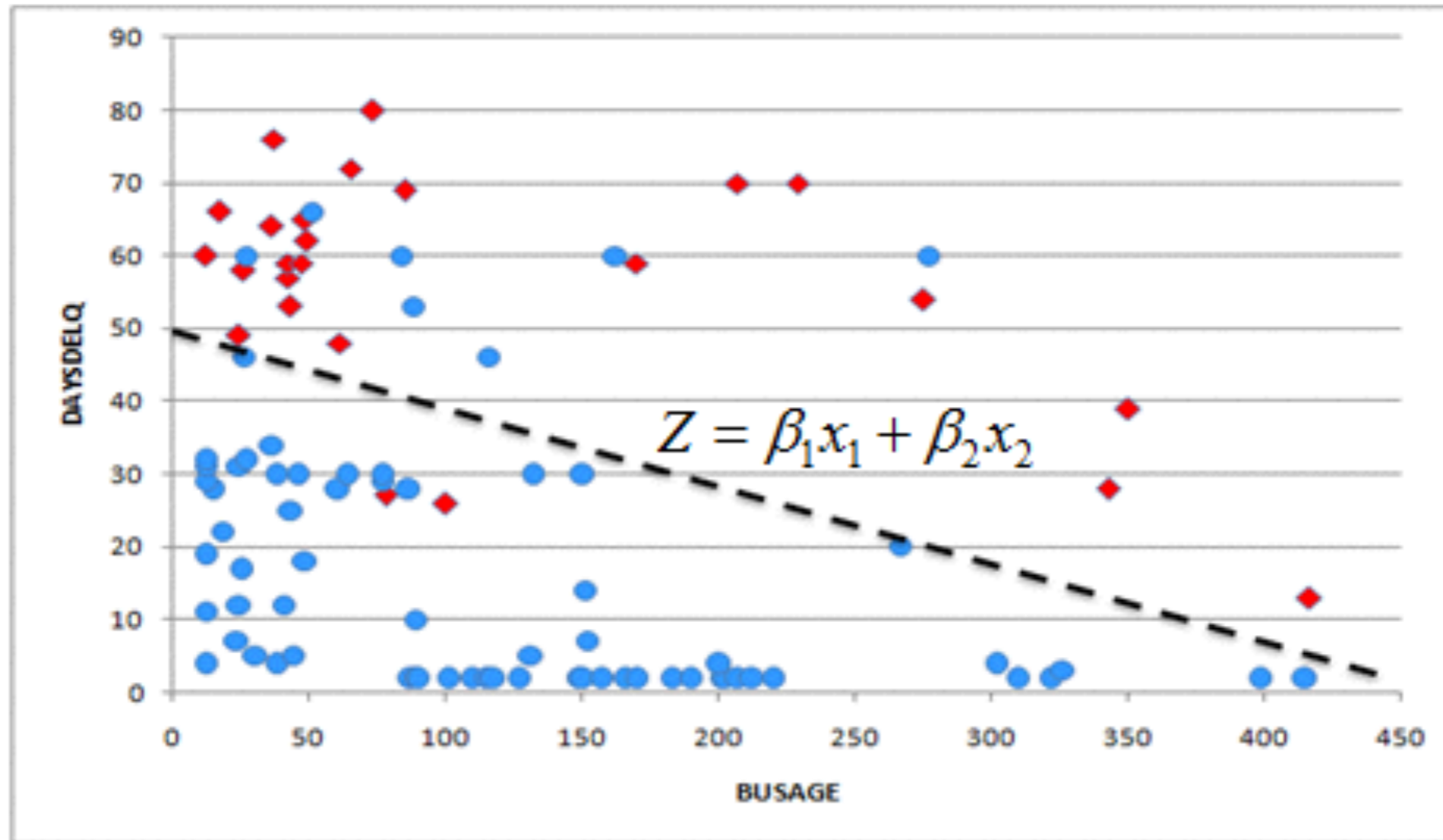
Decision Tree

☺ El árbol de decisiones se monta basado en la entropía

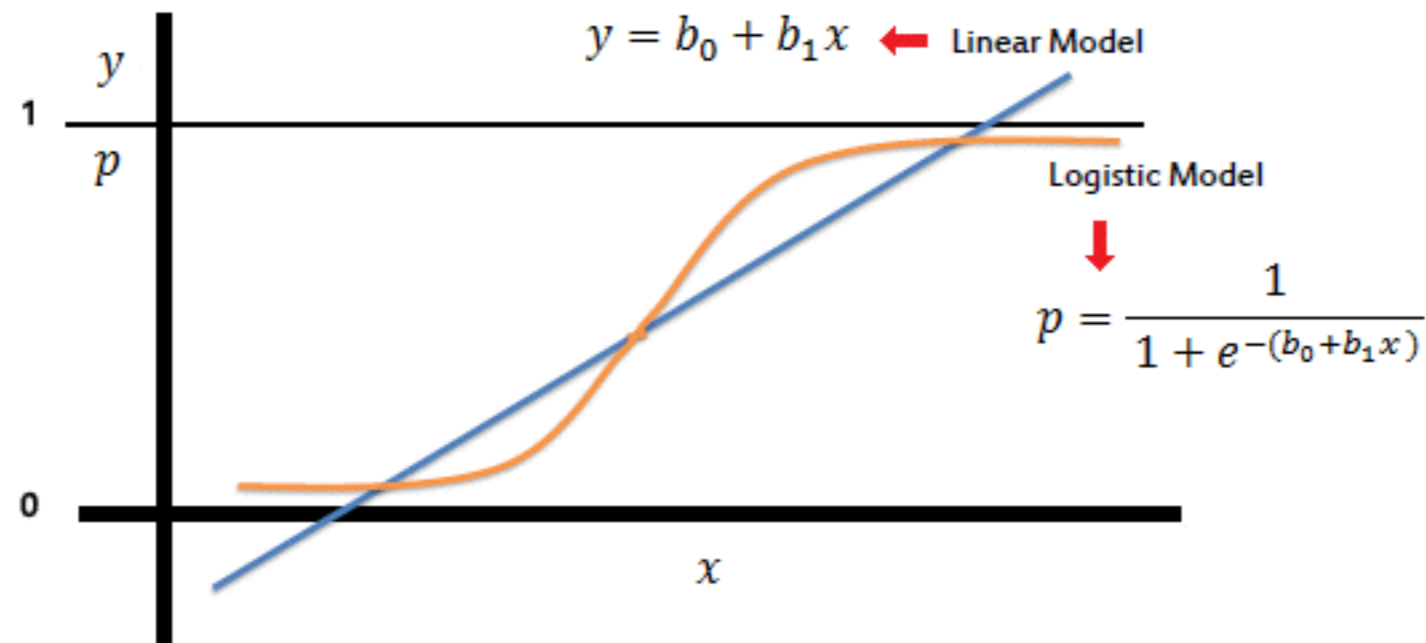
Predictors				Target
Outlook	Temp.	Humidity	Windy	Play Golf
Rainy	Hot	High	False	No
Rainy	Hot	High	True	No
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Sunny	Mild	High	False	Yes
Sunny	Cool	Normal	False	Yes
Sunny	Cool	Normal	True	No
Overcast	Cool	Normal	True	Yes
Rainy	Mild	High	False	No
Rainy	Cool	Normal	False	Yes
Sunny	Mild	Normal	False	Yes
Rainy	Mild	Normal	True	Yes
Overcast	Mild	High	True	Yes
Overcast	Hot	Normal	False	Yes
Sunny	Mild	High	True	No



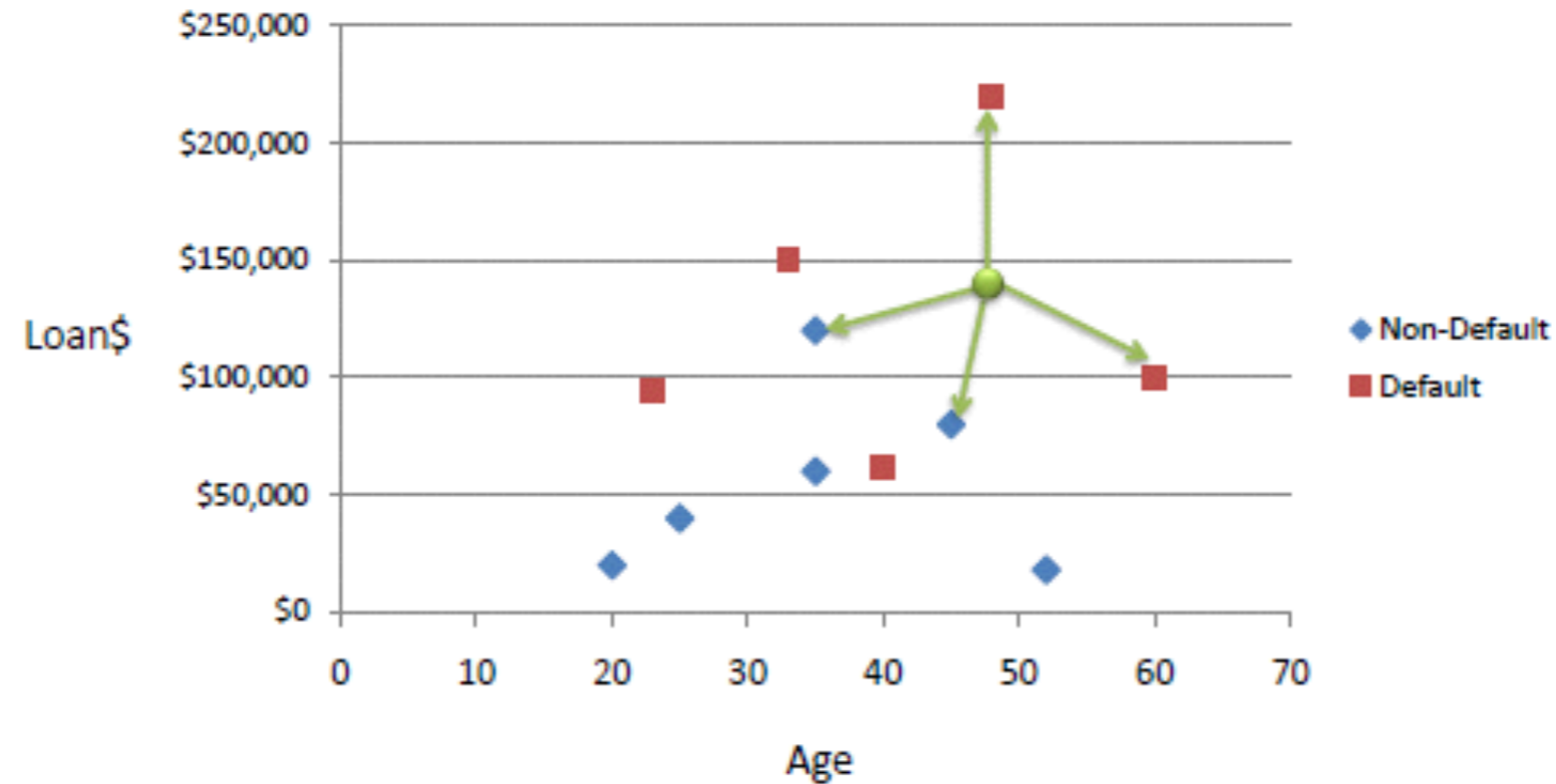
Analysis Discrim. Linear



Regresión Logística

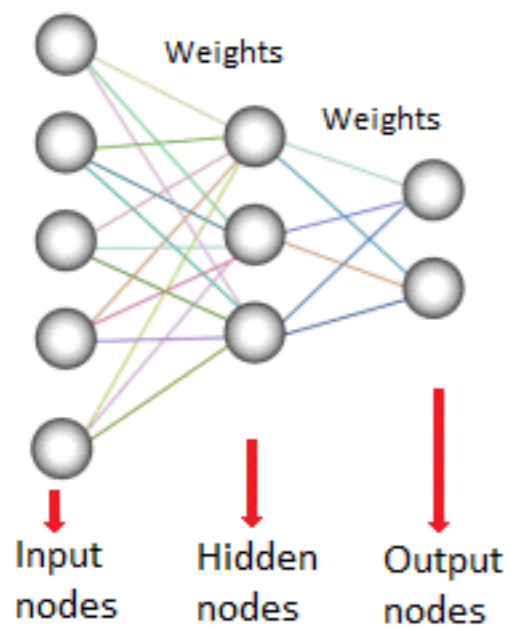


KNN K Nearest Neighbors

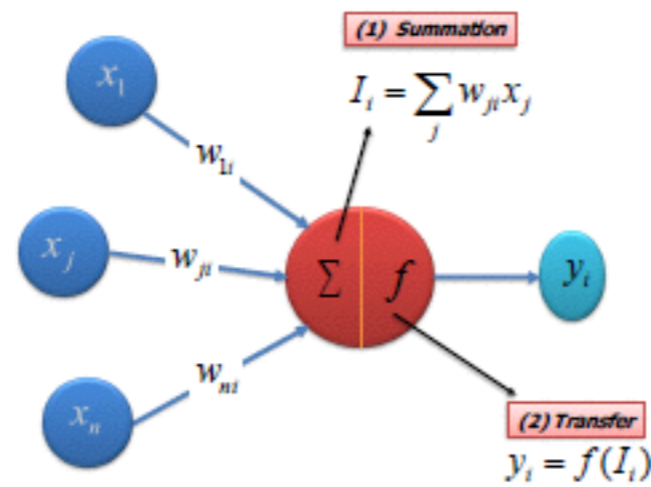


Artificial Neural Networks

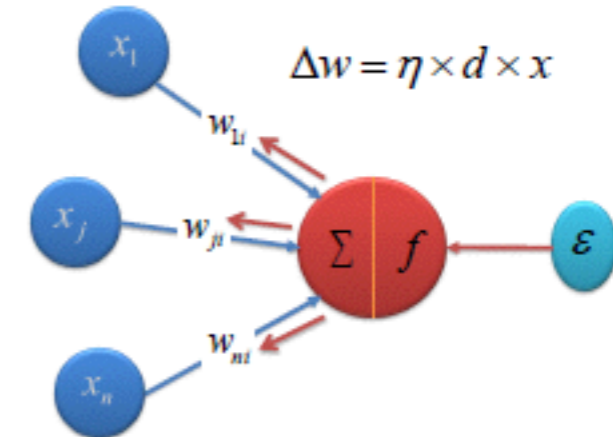
Artificial Neural Network



Feedforward Input Data



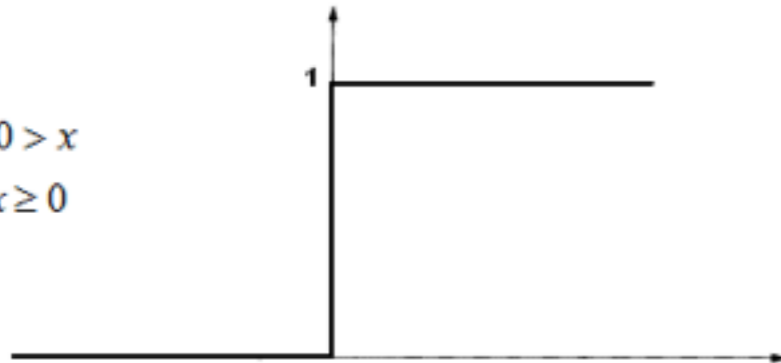
Backward Error Propagation



Artificial Neural Networks

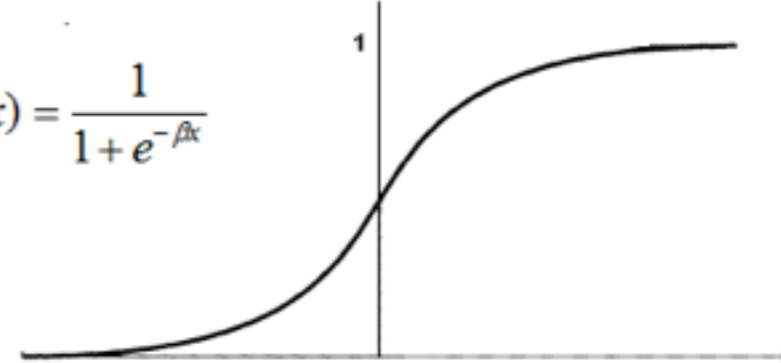
Unit step (threshold)

$$f(x) = \begin{cases} 0 & \text{if } x < 0 \\ 1 & \text{if } x \geq 0 \end{cases}$$



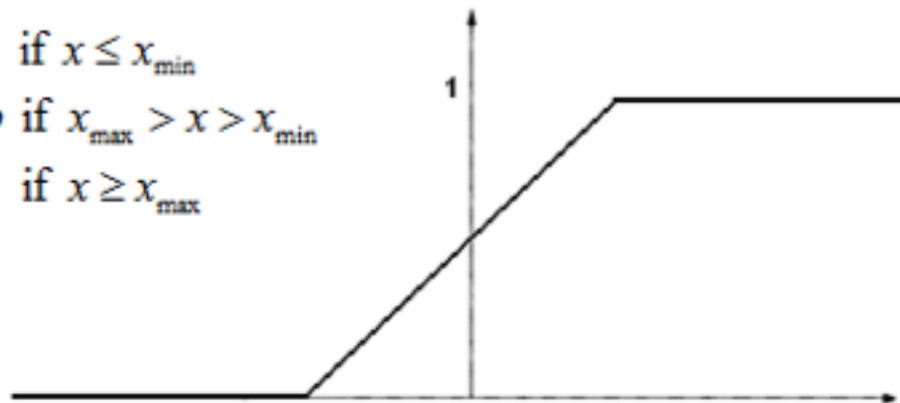
Sigmoid

$$f(x) = \frac{1}{1 + e^{-\beta x}}$$



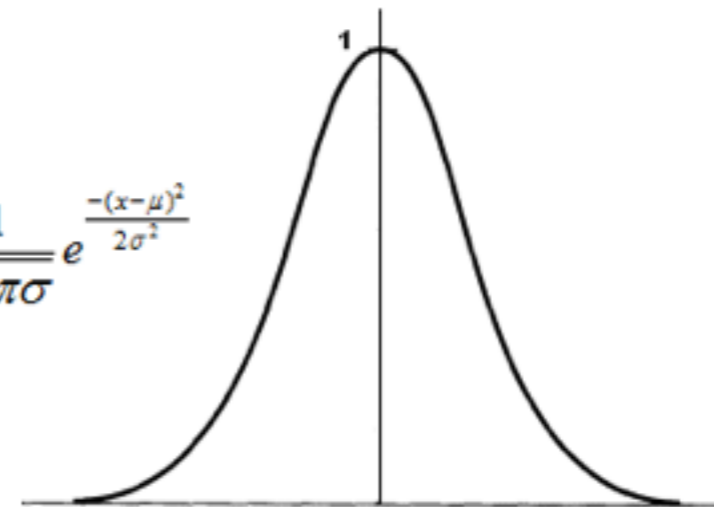
Piecewise Linear

$$f(x) = \begin{cases} 0 & \text{if } x \leq x_{\min} \\ mx + b & \text{if } x_{\max} > x > x_{\min} \\ 1 & \text{if } x \geq x_{\max} \end{cases}$$

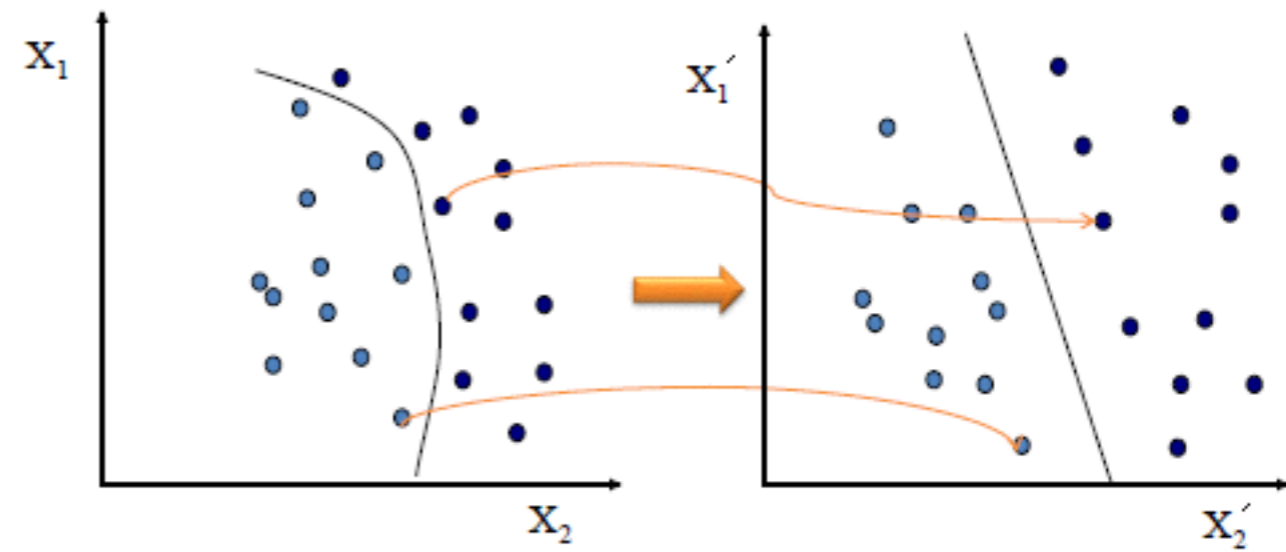
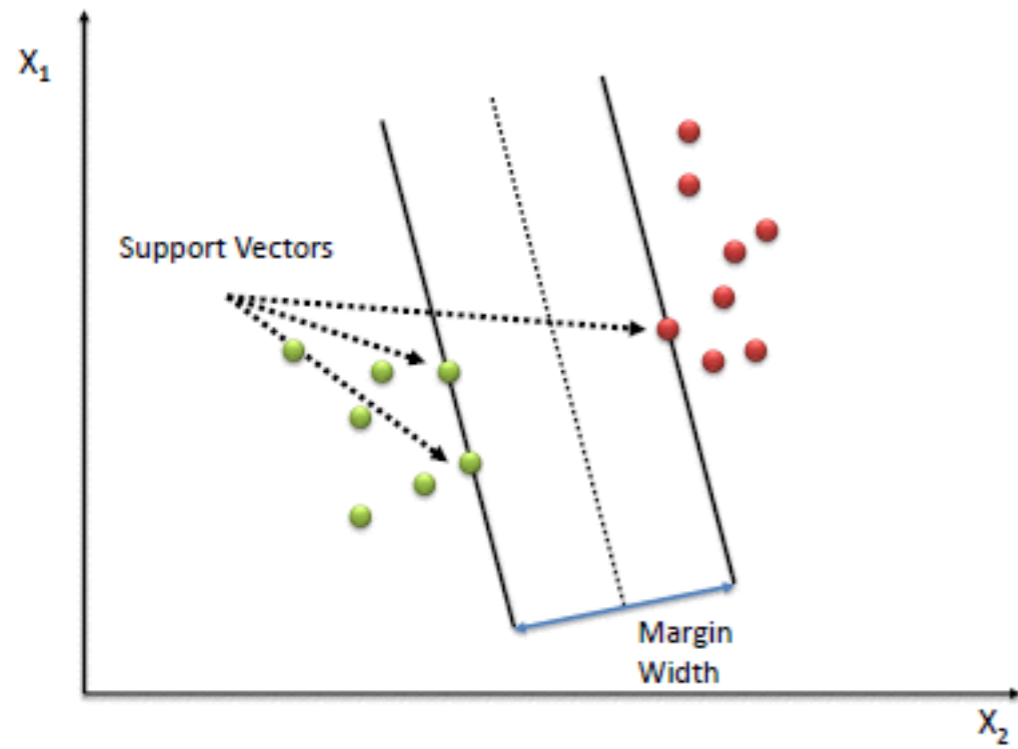


Gaussian

$$f(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}$$

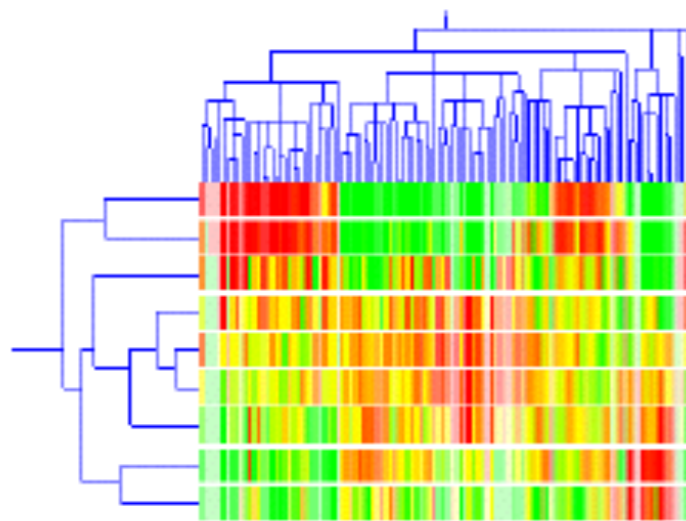


Support Vector Machine

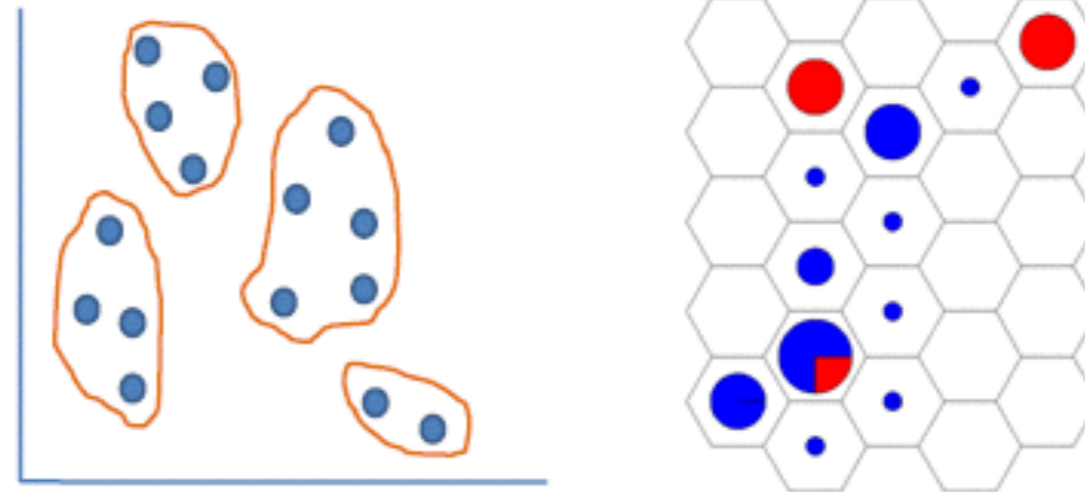


Clustering

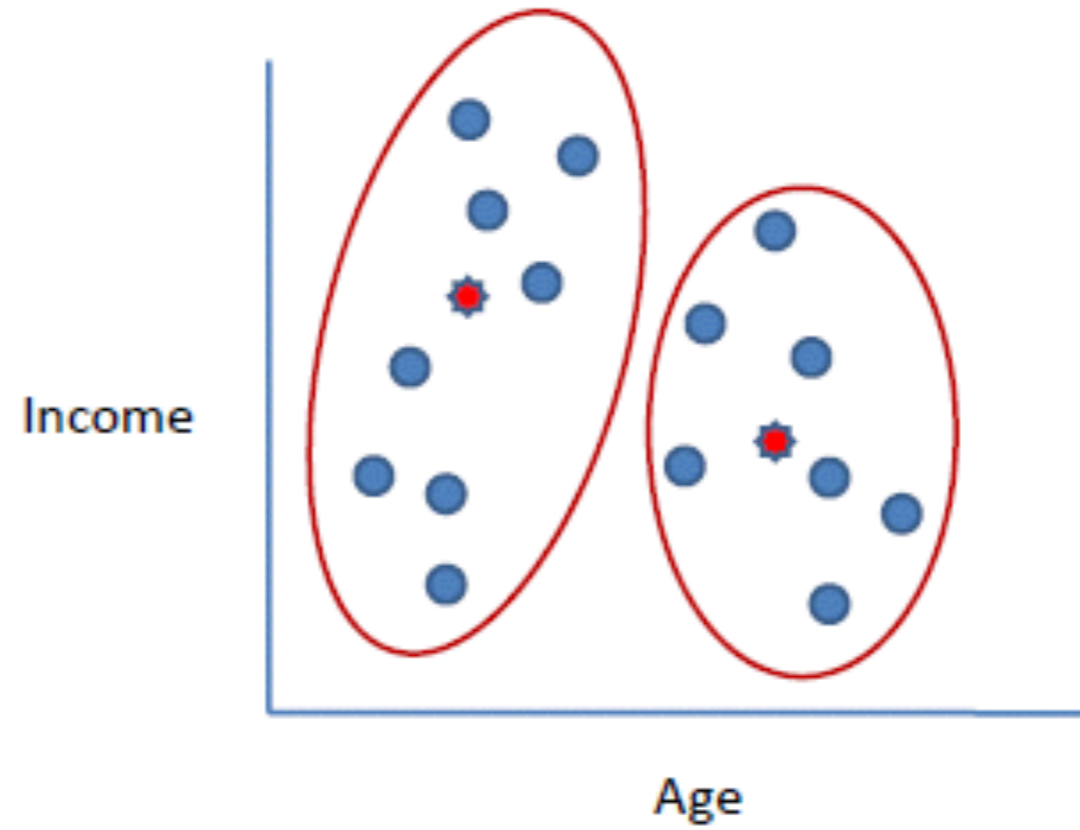
Hierarchical Clustering



Partitive Clustering

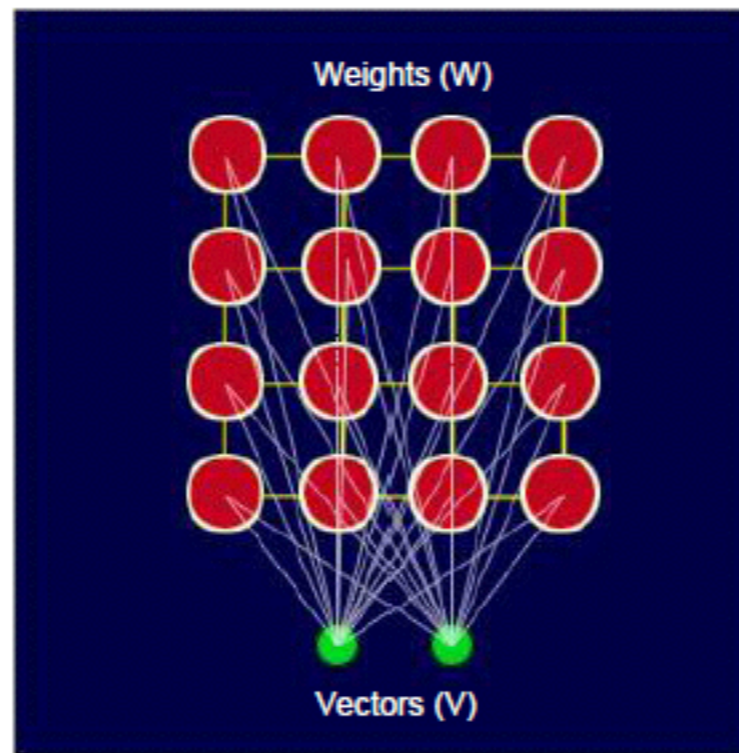


K-Means



SOM Mapas Autoorganiza.

2-D Self Organizing Map



Every node is connected to the input the same way, and no nodes are connected to each other.

Algunos Casos de Éxito

- ☺ Hay muchos ;-)
- ☺ Un par de mis Favoritos
 - ☺ TARGET
 - ☺ Envios

n

- ☺ Curso Machine Learning de Andrew Ng en coursera
- ☺ Smart Data: Foreman, John
- ☺ [kaggle.com](https://www.kaggle.com)
- ☺ Apuntarse al meetup del machine learning spain
- ☺ La siguiente charla

Muchas Gracias

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