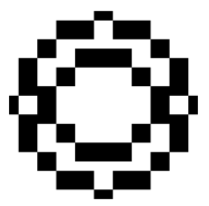


# Machine Learning Tribes

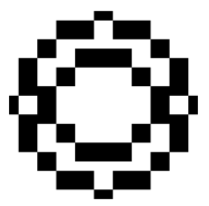
Master in Data Science and Advanced  
Analytics  
BA and DS

Roberto Henriques



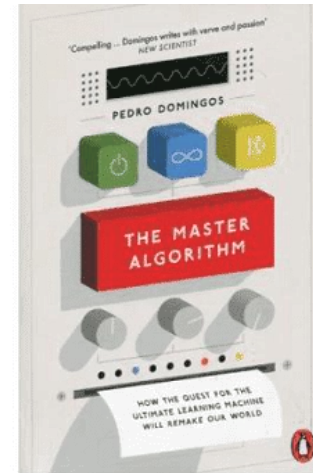
# How do computers get new knowledge?

- Fill in gaps in existing knowledge
- Emulate the brain
- Simulate evolution
- Reduce uncertainty
- Check for similarities between old and new



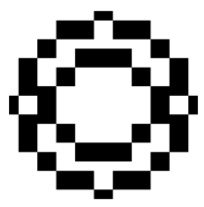
# Tribes of Machine Learning

- Associated to each of these ways, we have Machine Learning areas
  - Symbolism
  - Connectionism
  - Evolutionary Computation
  - Statistical Learning
  - Analogy Modelling



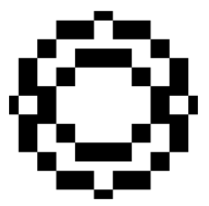
The Master Algorithm: How the Quest for the Ultimate Learning Machine Will Remake Our World

- Pedro Domingos, 2015
- ISBN:978-0465065707



# Tribes of Machine Learning

Tribe	Origins	Master Algorithm
Symbolists	Logic, philosophy	Inverse deduction
Connectionists	Neuroscience	Backpropagation
Evolutionaries	Evolutionary biology	Genetic programming
Bayesians	Statistics	Probabilistic inference
Analogizers	Psychology	Kernel Machines (SVM)



# Symbolists

- As in mathematical operations

Addition       $2 + 2 = ?$

Subtraction       $2 + ? = 4$

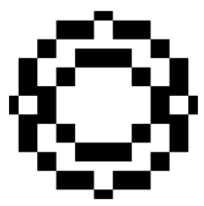
- Inverse deduction (induction)

Deduction

Socrates is human  
+ Humans are mortal  
= ?

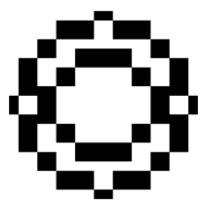
Induction

Socrates is human  
+ ??  
= Socrates is mortal



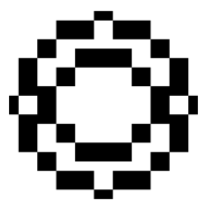
# Symbolists

- Logic programming
  - Prolog
  - Inductive and Abductive logic programming
- Expert systems
  - Knowledge bases, Inference Engines,
- Decision Trees
  - C4.5 the most popular, but has many variants, including regression trees and random forests
- Functional programming – Lisp ...



# Connectionists

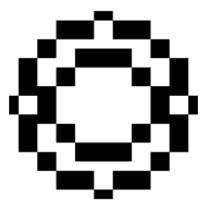
- World is not logic!
- Draws on Biology, more specifically Biological Neurons
  - Model neural cells, McCulloch and Pitts (1943)
- Milestones
  - 1957: Perceptron invented by Frank Rosenblatt.
  - 1968: Minsky and Papert publish the book *Perceptrons*, criticizing single-layer perceptrons.
  - 1986: Backpropagation (re)invented, and connectionist research restarts
  - 2006: Hinton *et al.* publish A fast learning algorithm for deep belief nets, which rejuvenates interest in Deep Learning. Important points of Deep Learning success and transfer learning
  - 2017: AlphaGo defeats reigning Go world champion, using DL



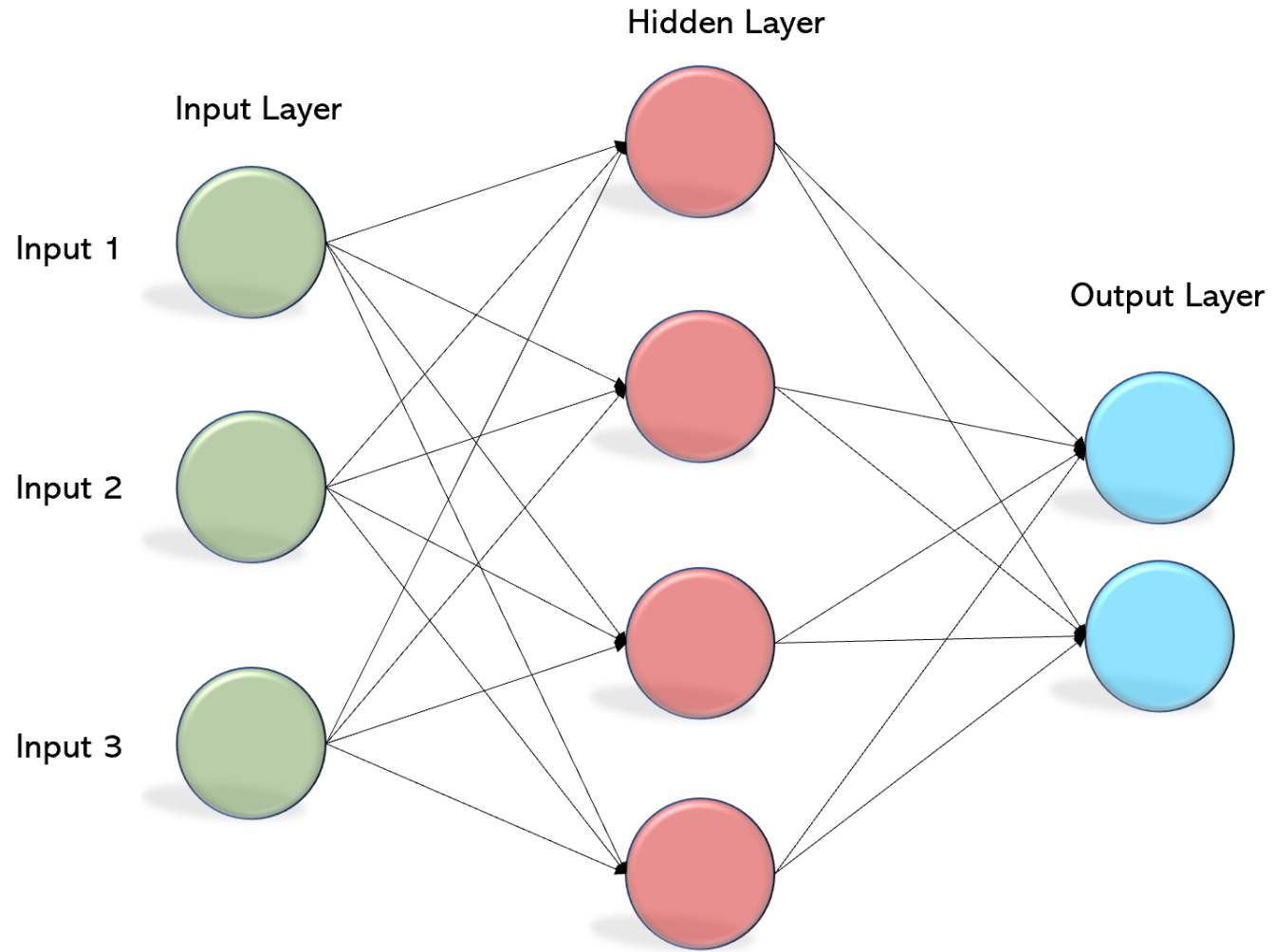
# Connectionists

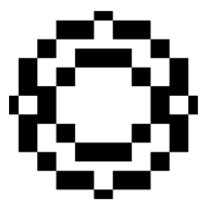
- Main algorithms
  - Perceptron (simple perceptron)
  - Multilayer Perceptron (MLP)
    - Varying structures, including feedback and time-delay
  - DeepLearning Networks
    - Convolutional Neural Networks
  - Hopfield Networks
  - Boltzman Networks





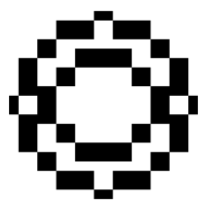
# Connectionists: MLP



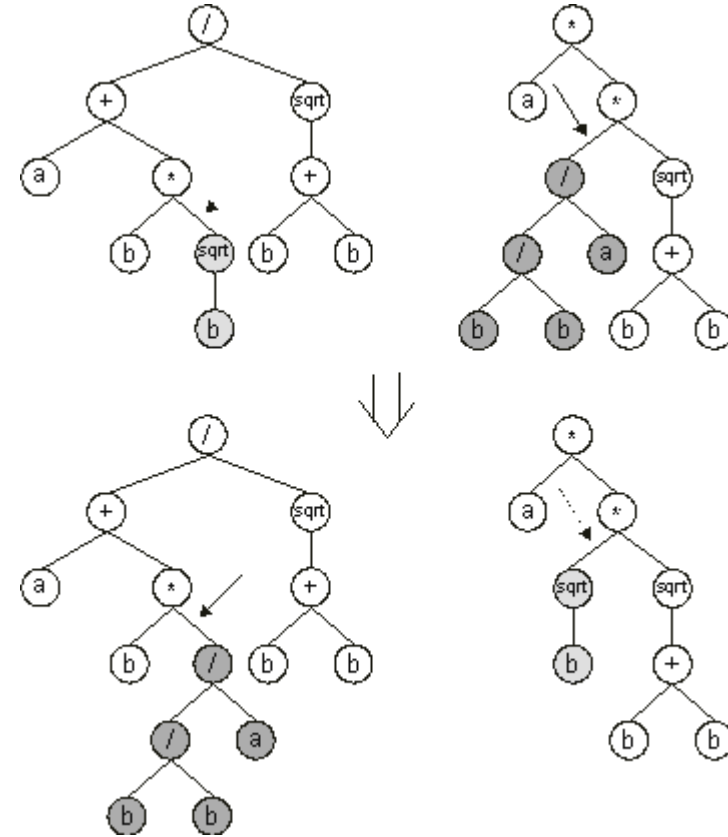
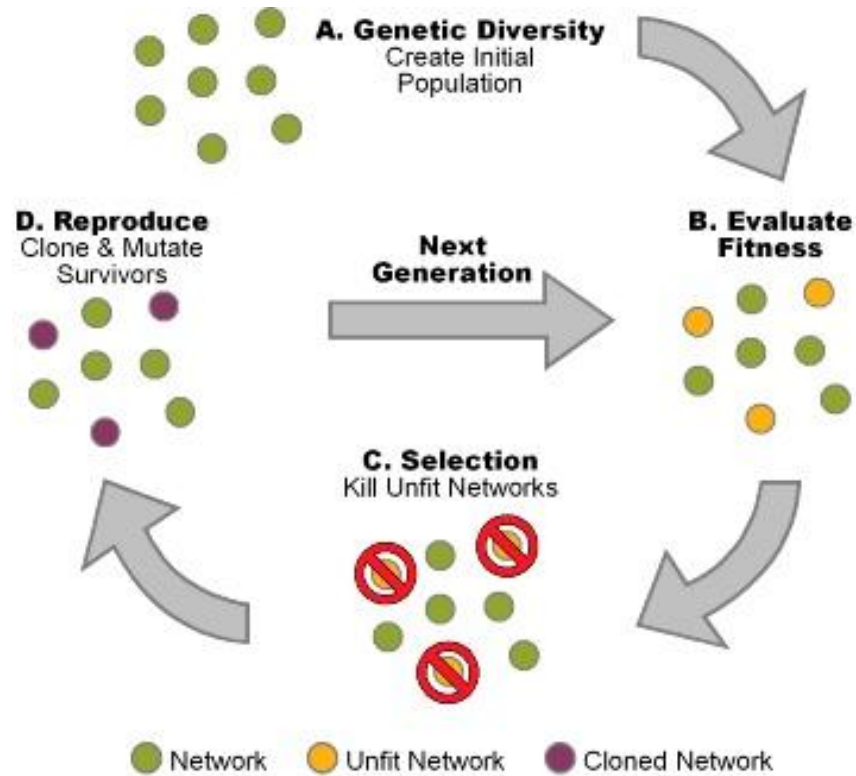


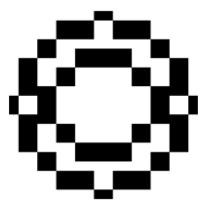
# Evolutionary

- Draws on Biology, the Theory of Evolution and Natural Selection
  - Models natural selection, crossover, mutations
- Milestones:
  - 1964: Lawrence Fogel published on *evolutionary programming*
  - 1975: John Holland invents *genetic algorithms*.
  - 1981: Richard Forsyth coins *genetic programming* and demonstrates it using trees of actions.
  - 1991: Dorigo writes a thesis on *Ant Colony Optimization*
  - 1995: Kennedy & Eberhart invent *Particle Swarm Optimization*



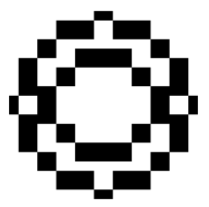
# Evolutionaries: GA and GP





# Bayesians

- Draws on Statistics
  - Bayes Theorem, and evidence
- Everything is uncertainty, so compute probabilities
- Milestones:
  - 1953: Monte Carlo Markov Chain (MCMC) invented. Bayesian inference finally becomes tractable on real problems.
  - 1968: Hidden Markov Model (HMM) invented.
  - 1988: Judea Pearl authors Probabilistic Reasoning in Intelligent Systems, and creates the discipline of probabilistic graphical models (PGMs).
  - 2000: Pearl authors Causality: Models, Reasoning, and Inference, and creates the discipline of causal inference on PGMs



# Bayesians: algorithms

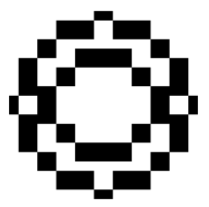
- Bayesian Classifiers
  - NB, MAP, ML
- Bayesian Belief Networks
- Variants of Markov Models

The diagram illustrates Bayes' Theorem with the following components:

- LIKELIHOOD** (orange text): the probability of "B" being TRUE given that "A" is TRUE. An arrow points to the  $P(B|A)$  term in the numerator.
- PRIOR** (teal text): the probability of "A" being TRUE. An arrow points to the  $P(A)$  term in the numerator.
- POSTERIOR** (green text): the probability of "A" being TRUE given that "B" is TRUE. An arrow points to the  $P(A|B)$  term in the denominator.
- The probability of "B" being TRUE** (pink text): An arrow points to the  $P(B)$  term in the denominator.

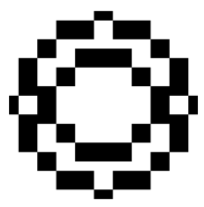
$$P(A|B) = \frac{P(B|A) P(A)}{P(B)}$$

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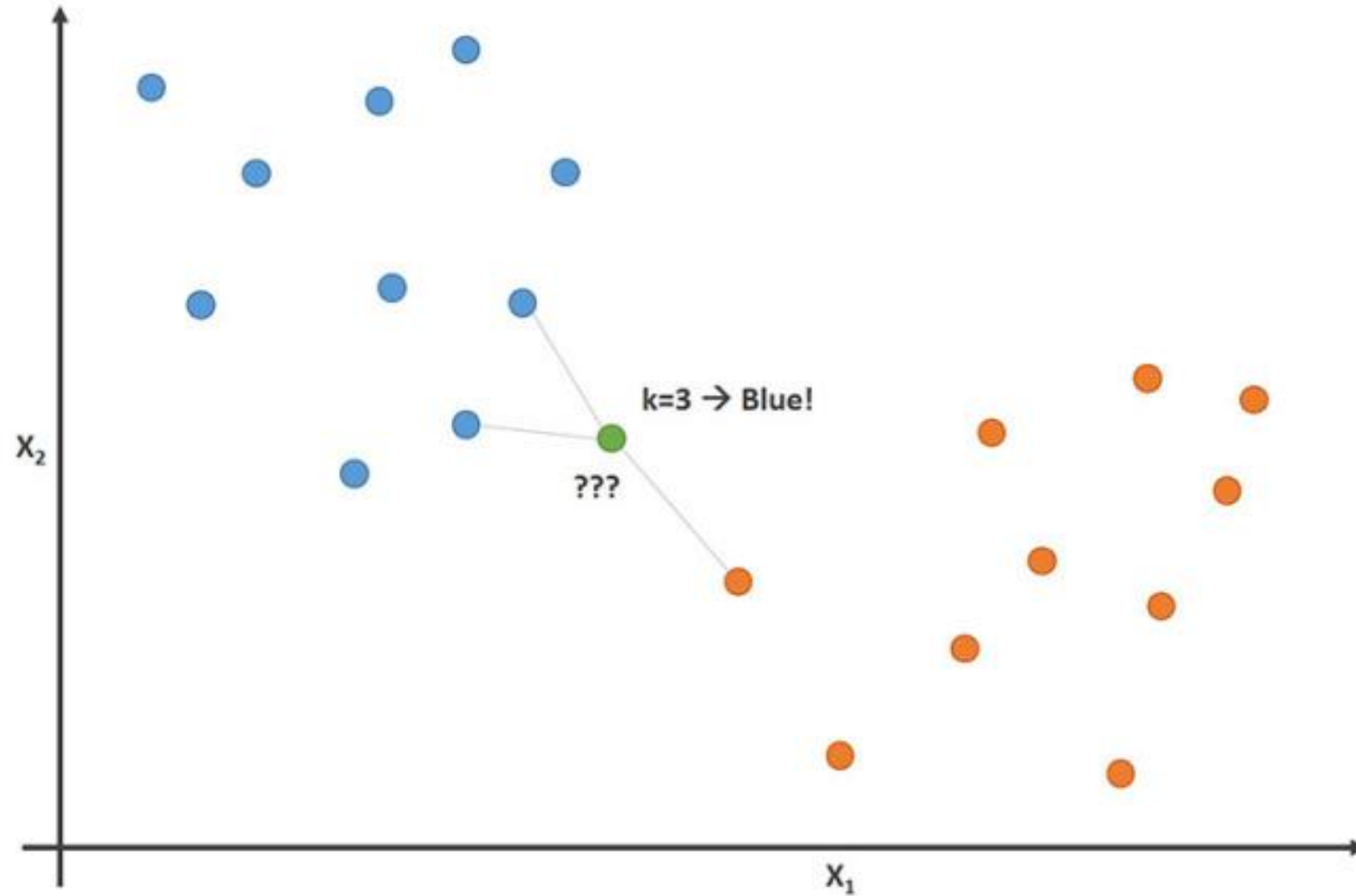


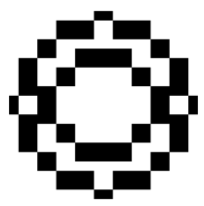
# Analogizers

- Everything we learn is reasoning by analogy
- Draws on Human behaviour and Statistics
- Milestones:
  - 1968: k-nearest neighbor algorithm increases in popularity.
  - 1979: Douglas Hofstadter publishes Godel, Escher, Bach.
  - 1982. Kohonen's book on Self-Organization and Associative Memory
  - 1992: Support Vector Machines (SVMs) invented



# Analogizers: kNN and SVM

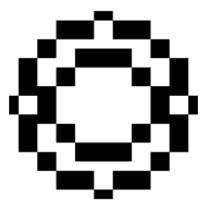




# Can we unify all tribes?

Tribe	Origins	Master Algorithm
Symbolists	Logic, philosophy	Inverse deduction
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Evolutionaries	Evolutionary biology	Genetic programming
Bayesians	Statistics	Probabilistic inference
Analogizers	Psychology	Kernel Machines (SVM)





# Components of ML

- Representation
  - Probabilistic logic (Markov logic Networks)
  - Each rule has a weight
- Evaluation
  - Posterior probability
  - User-defined objective function
- Optimization
  - Formula discovery: Genetic programming
  - Weight learning: Backprop



# Questions?