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# Simulierte Evolution: Hands-On-Starter Kit

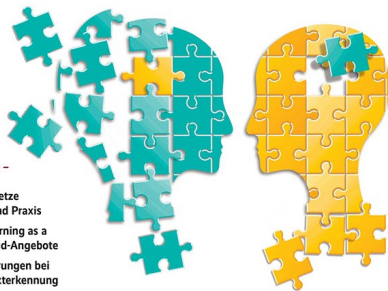
**JFS 2022**

Heiko Spindler  
Freelancer

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# Machine Learning

Verstehen, verwenden, verifizieren



**Data Science –  
Status quo**

Neuronale Netze  
in Theorie und Praxis

Machine Learning as a  
Service: Cloud-Angebote

Herausforderungen bei  
Bild- und Texterkennung

Blick in die Blackbox

**TensorFlow, Keras & Co.**  
ML-Frameworks  
und -Bibliotheken

Programmiersprachen:  
Python, Scala, C++

**Hardware pusht ML**  
GPUs und CPUs  
für Machine Learning

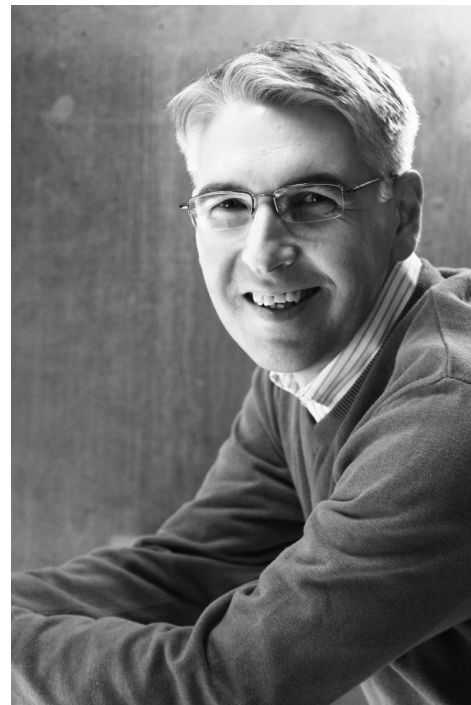
Neuromorphe Chips für  
neuronale Netze

**Mensch trifft KI**  
Rechtsfragen bei  
Künstlicher Intelligenz  
Verantwortungsvoller  
Einsatz und Ethik

## Heiko Spindler

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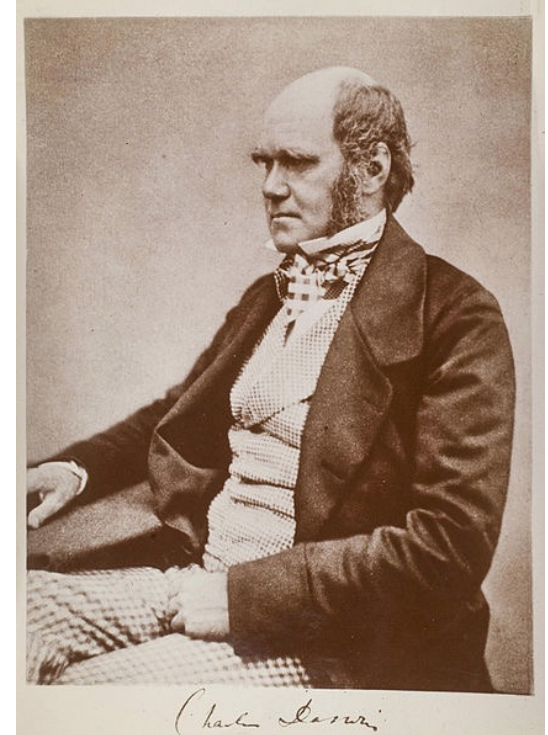
# Evolution

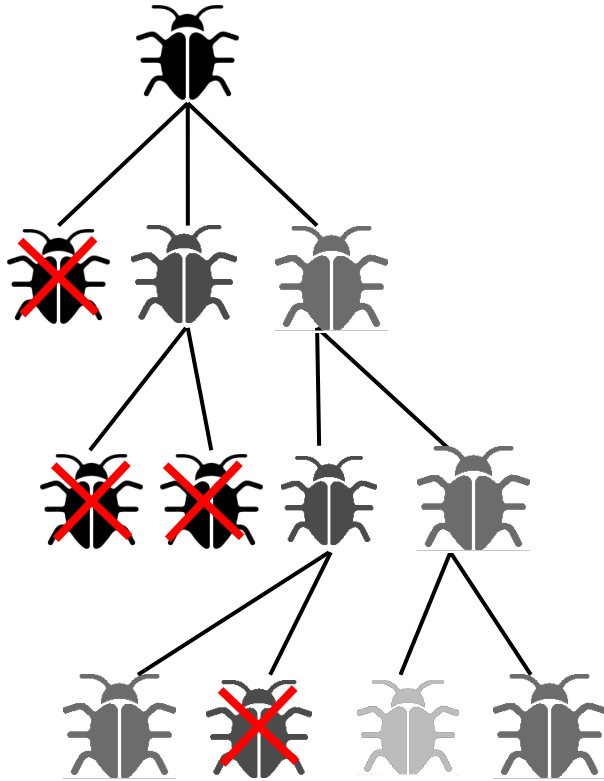
... is the process by which traits that enhance survival and reproduction become more common in successive generations:

Variation exists within populations of organisms: morphology, physiology, and behavior

Different traits confer different rates of survival and reproduction (differential fitness)

Traits can be passed from generation to generation (heritability of fitness).





**Mutation creates variation**

**Not every offspring survives**

**Favorable mutations are more likely to survive and reproduce**

# — Evolution

- not only in nature!

Do you know other areas where Evolution takes place?



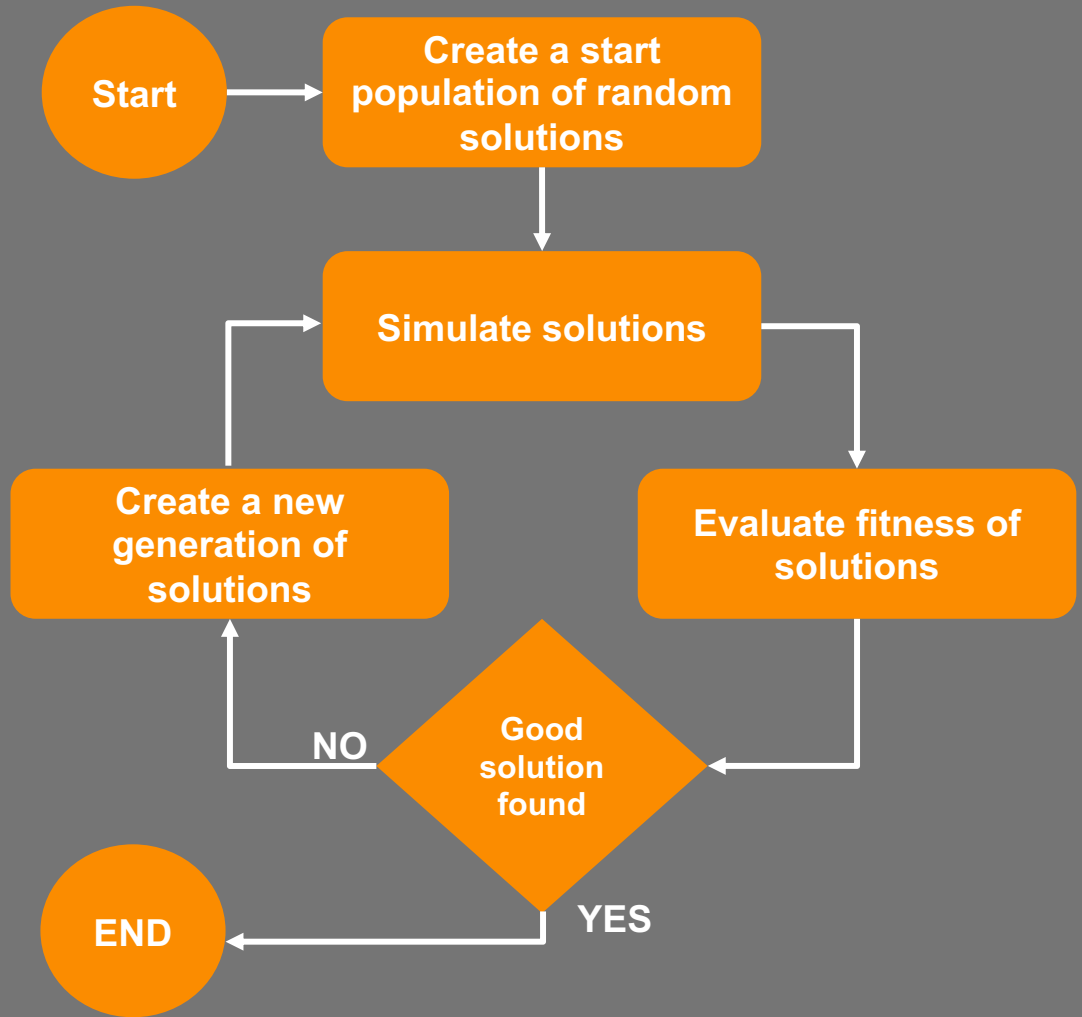


# What do we need to simulate Evolution?

My Plan:

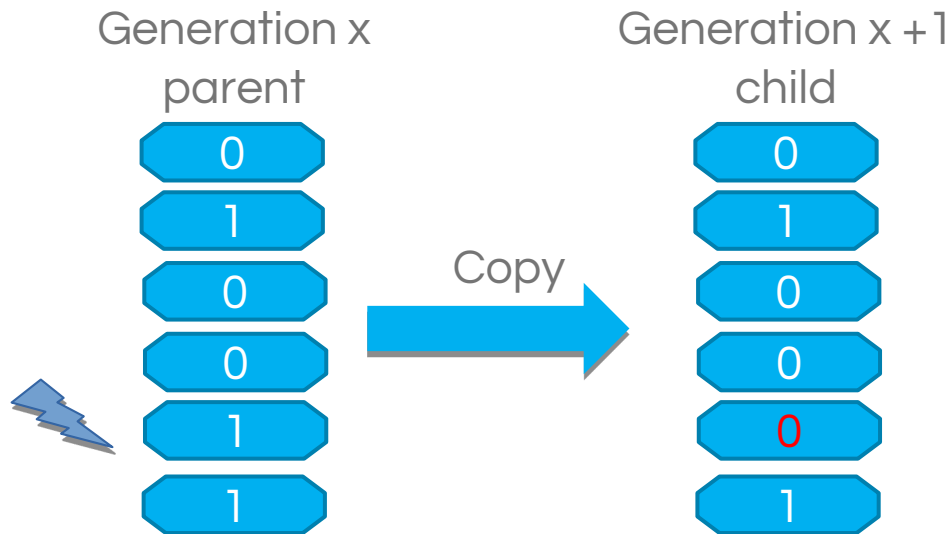
- Data structure for genes
- Implementation of genetic operations
- Implementation of the process
- Simulation of solutions
- Fitness function

# Process of Evolution





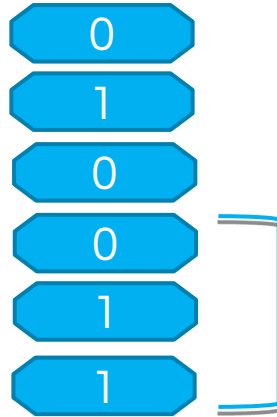
# Mutation



Changes a value at a random position

# Inversion

Generation x  
parent



Generation x + 1  
child



The sequence changes.

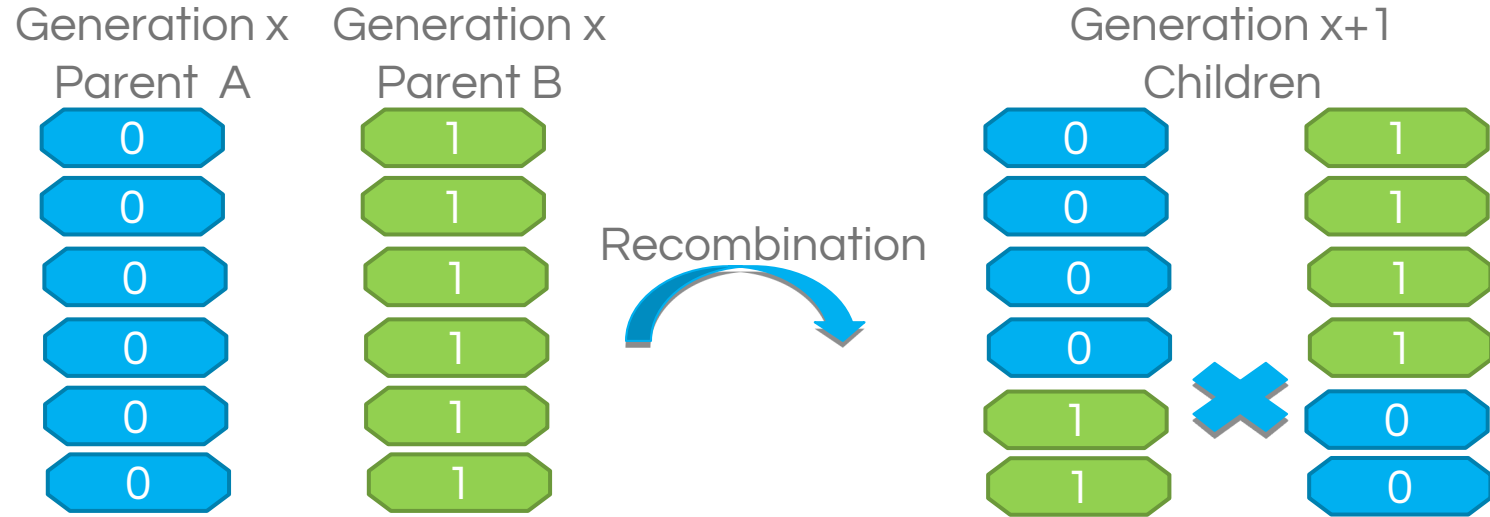
# Recombination

Two solution mate

Fast spreading of  
good traits



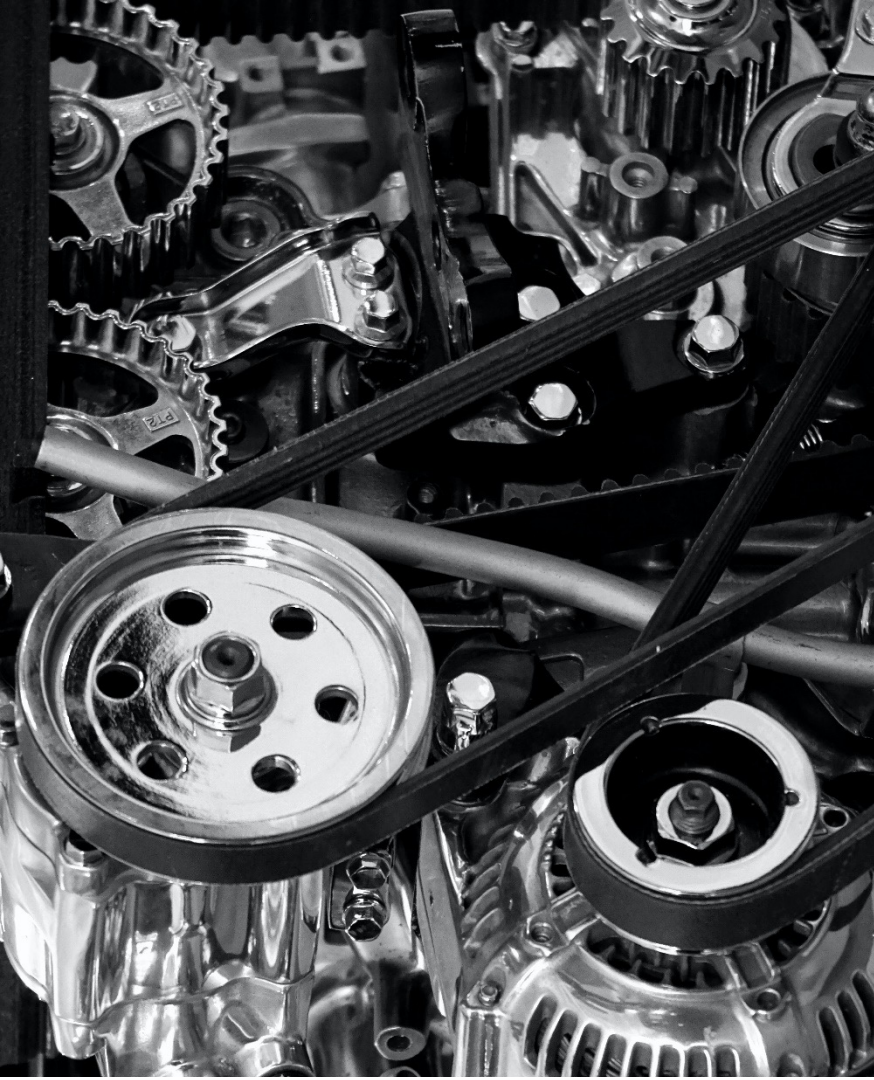
# Recombination / Cross-Over





# To Dos

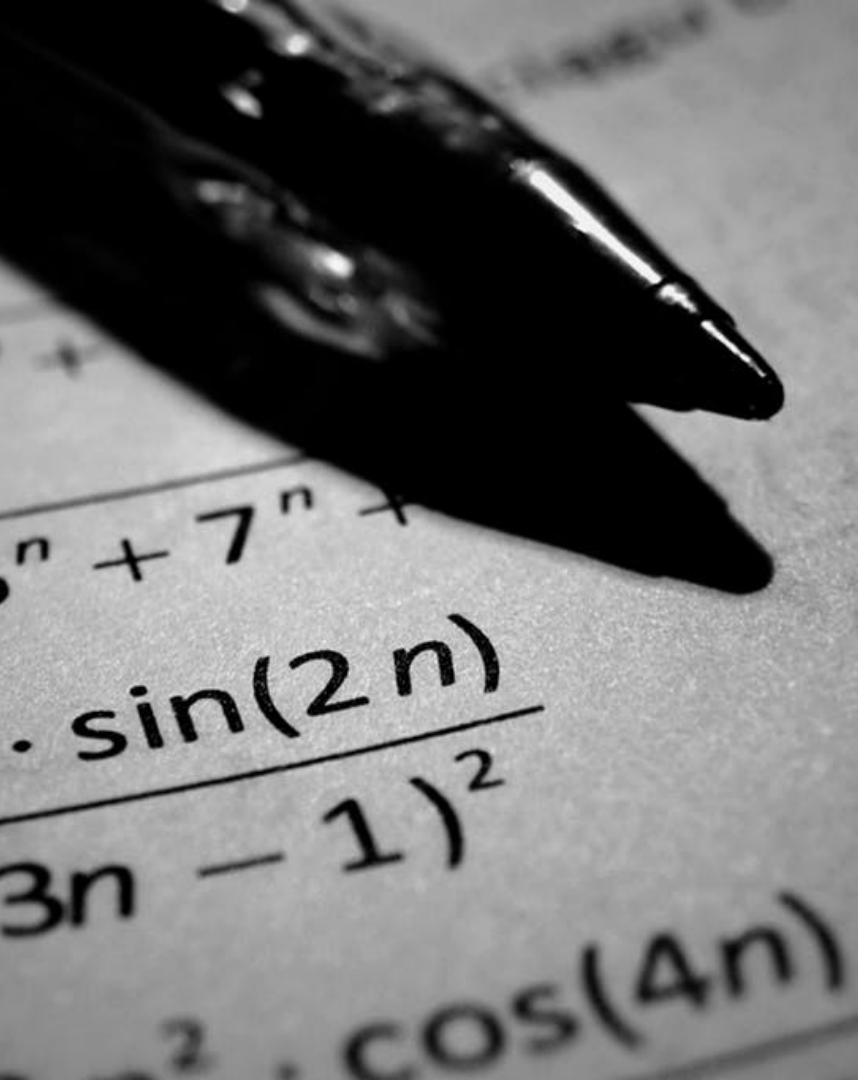
- Define an encoding / decoding for the genome
- Define a Fitness-Function
- Define good parameters for the evolution process:
  - Population size
  - Mutation rate
  - Cross over rate



# Frameworks can help

- Create initial populations (random)
- Run the process
  - Calculate the fitness
  - Sort solutions by fitness
  - Apply genetic operators
  - Distribute the work





# Define a Fitness Function

- Simple to calculate
- Maps a solution to a number  
(Higher number: better solution)

# Frameworks

JGAP: <http://jgap.sourceforge.net>

~~Apache Commons Math:~~ <http://commons.apache.org>

Jenetics: <http://jenetics.io>

## Features

- Java, Open Source
- Data structures
- Genetic operations
- Interfaces for fitness functions
- Process of simulated evolution

# Sample Coding



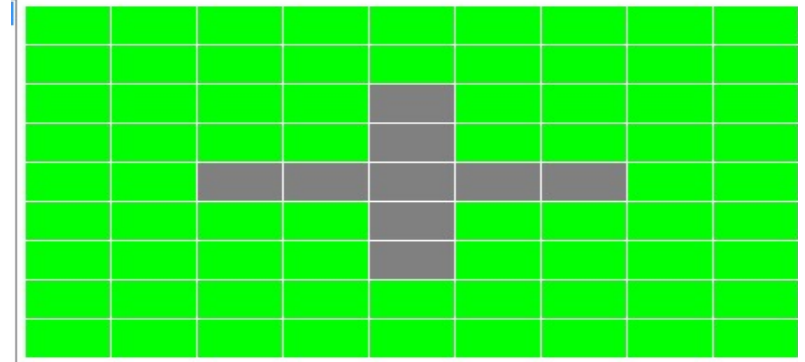


# The challenge

A Robot should visit and mark all cells of a given board (9 x 9 cells).

Some cells are blocked (gray).

The memory holds up to 30 commands.



# Commands

Command	Description
<b>Move</b> <left, right, up, down>	Moves the robot
<b>Set</b>	Marks the cell where the robot is located
<b>Nop</b>	„No Operation“
<b>Goto</b> <Command No.>	Continues the execution of the program at another position
<b>If</b> <left, right, up, down> = <free, marked, border> <b>Goto</b> <Command No.>	Conditional jump depends on the state of the current cell

# Mapping of Commands to Genes



Commands are stored in 4 integer values

Not used parameters are ignored



# Fitness Function

Priority 1: „Get the job done“:

- Fitness is given by the percentage of marked cells (0.0 to 100.0).

...

# Fitness Function

Priority 1: „Get the job done“:

- Fitness is given by the percentage of marked cells (0.0 to 100.0).

Priority 2: „Be efficient“:

- Every command execution consumes energy
- The robot gets energy for new marked cell
- The energy left after marking all cells defines the fitness.

**Main parameters to  
drive evolution**



# Population Size

- 50 up to 1000 Individuals.
- In a large population, a slight improvement may not prevail and is lost in the crowd.

## Mutation Rate

- High values result in much destruction.
- Too low values mean too little creativity and new ideas.

## Crossing Over Rate

- High Rate: Process converges quickly but fixes (too) early on specific solutions.
- Niedrige Rate: Gute Individuen (insbesondere nur leichte Verbesserungen) setzen sich evtl. nur langsam oder gar nicht durch.

# Ideas for Improvement

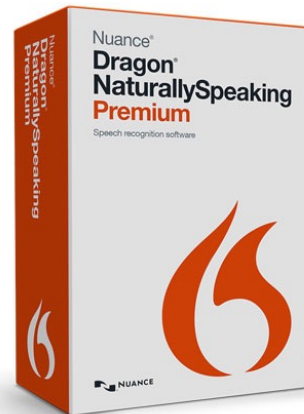
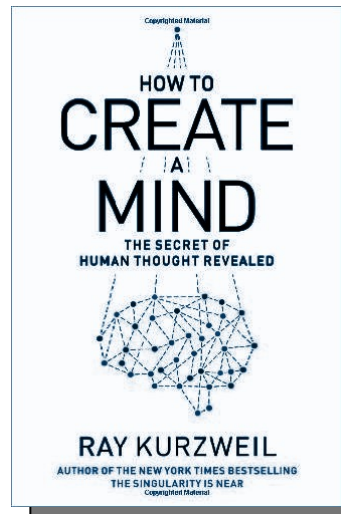
- Adaptive mutation rate:  
Long periods of stagnation lead to an increase in the mutation rate.
- Elitism:  
The best solution(s) will always survive (copied to next generation).
- Island Model:  
Separated populations breed different solutions and share the best individuals from time to time.



# Simulated Evolution in Action

Ray Kurzweil, 2016

**“How to Create a Mind:  
The Secret of Human  
Thought Revealed”**





# Q & A

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