

Ένωση Πληροφορικών Ελλάδας



Το μέλλον του AI: Τι αναμένουμε να γίνει
καλύτερο, φθηνότερο, πιο αποδοτικό ενεργειακά

Χάρης Γεωργίου (MSc, PhD)

Ένωση Πληροφορικών Ελλάδας

Στόχοι:

- Πρώτος “καθολικός” φορέας εκπροσώπησης πτυχιούχων Πληροφορικής.
- Αρμόδιος φορέας εκπροσώπησης επαγγελματιών Πληροφορικής.
- Αρμόδιος επιστημονικός “συμβουλευτικός” φορέας για το Δημόσιο.
- Αρωγός της Εθνικής Ψηφιακής Στρατηγικής & Παιδείας της χώρας.

<https://www.epe.org.gr>



Τομείς παρέμβασης

Ποιοι είναι οι κύριοι τομείς παρεμβάσεων της ΕΠΕ;

- 1 Εθνική Ψηφιακή Στρατηγική & Οικονομία
- 2 Εργασιακά (ΤΠΕ), Δημόσιος & ιδιωτικός τομέας
- 3 Παιδεία (Α', Β', Γ')
- 4 Έρευνα & Τεχνολογία
- 5 Έργα & υπηρεσίες ΤΠΕ
- 6 Ασφάλεια συστημάτων & δεδομένων
- 7 Ανοικτά συστήματα & πρότυπα
- 8 Χρήση ΕΛ/ΛΑΚ
- 9 Πνευματικά δικαιώματα
- 10 Κώδικας Δεοντολογίας (ΤΠΕ)
- 11 Κοινωνική μέριμνα (ICT4D)





Harris Georgiou (MSc, PhD) – <https://github.com/xgeorgio/info>

- R&D: Associate post-doc researcher and lecturer with the University Athens (NKUA) and University of Piraeus (UniPi)
- Consultant in Medical Imaging, Machine Learning, Data Analytics, Signal Processing, Process Optimization, Dynamic Systems, Complexity & Emergent A.I., Game Theory
- HRTA member since 2009, LEAR / scientific advisor
- HRTA field operator (USAR, scuba diver)
- Wilderness first aid, paediatric (child/infant)
- Humanitarian aid & disaster relief in Ghana, Lesvos, Piraeus
- Support of unaccomp. minors, teacher in community schools
- Streetwork training, psychological first aid & victim support
- 2+4 books, 200+ scientific papers/articles (and 6 marathons)

Επισκόπηση

- Περιεχόμενα:
 - Μέρος I: Τεχνητή Νοημοσύνη – Βασικές Έννοιες
 - Μέρος II: Μελλοντικές Προοπτικές
- Σχετικό υλικό:
 - «Turing machines explained visually» – https://www.youtube.com/watch?v=-ZS_zFg4w5k
 - «Game of life: computer with display» – <https://www.youtube.com/watch?v=Wfuhbl8HE7s>
 - Computer History Museum (CHM) – <https://www.youtube.com/@ComputerHistory>
 - «Artificial Intelligence | 60 Minutes Full Episodes» – <https://www.youtube.com/watch?v=aZ5EsdnpLMI>
 - «Queens puzzle solver in LISP», @ApneaCoding – https://youtu.be/_1CRCyklUto
 - «BAM neural network in Arduino», @ApneaCoding – <https://youtu.be/RkM-rpSVD4I>

Μέρος Ι: Τεχνητή Νοημοσύνη – Βασικές Έννοιες

1. Τι είναι;
2. Γιατί μας ενδιαφέρει;
3. Ποιος τη σχεδιάζει-δημιουργεί;
4. Νοημοσύνη στη Φύση

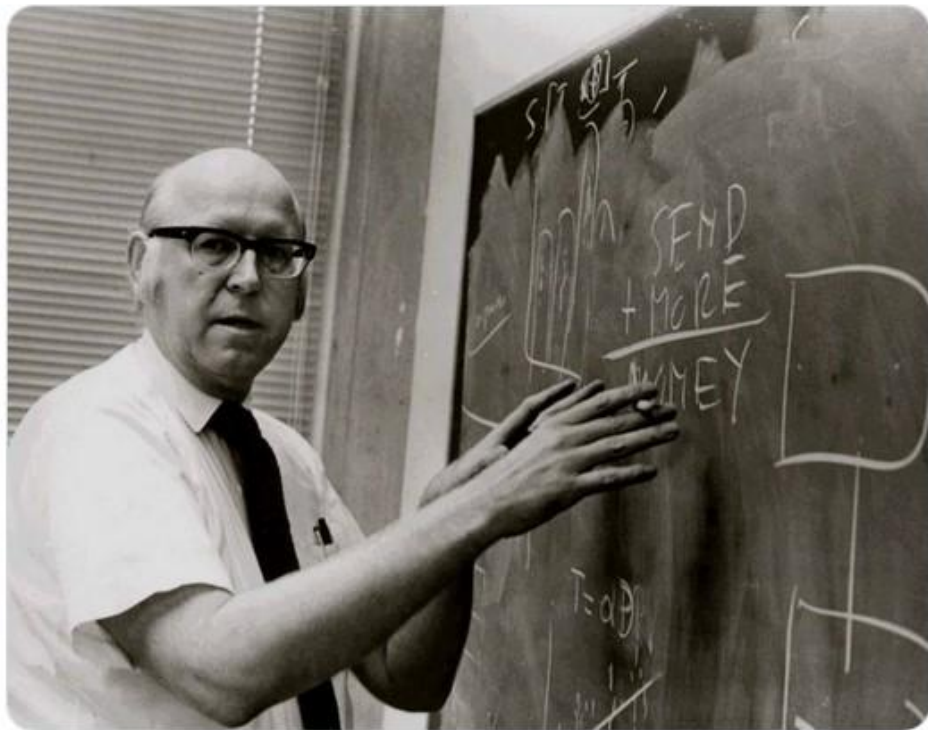




MIT CSAIL  @MIT_CSAIL · 13h

Born #otd in 1927: Allen Newell, who in 1955 — before the term “AI” even existed — co-wrote what’s widely considered the world’s first AI program, Logic Theorist. This work proved 38 out of 52 classic math theorems.

...



MEMORANDUM
RM-3739-RC
JUNE 1963

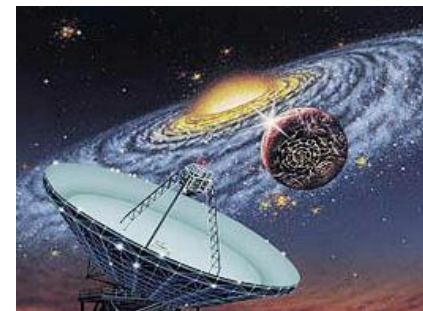
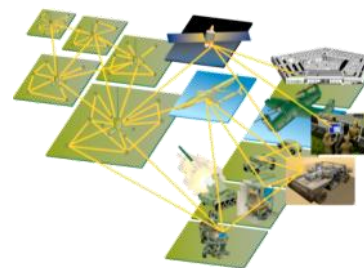
IPL-V PROGRAMMERS' REFERENCE MANUAL

Edited by Allen Newell

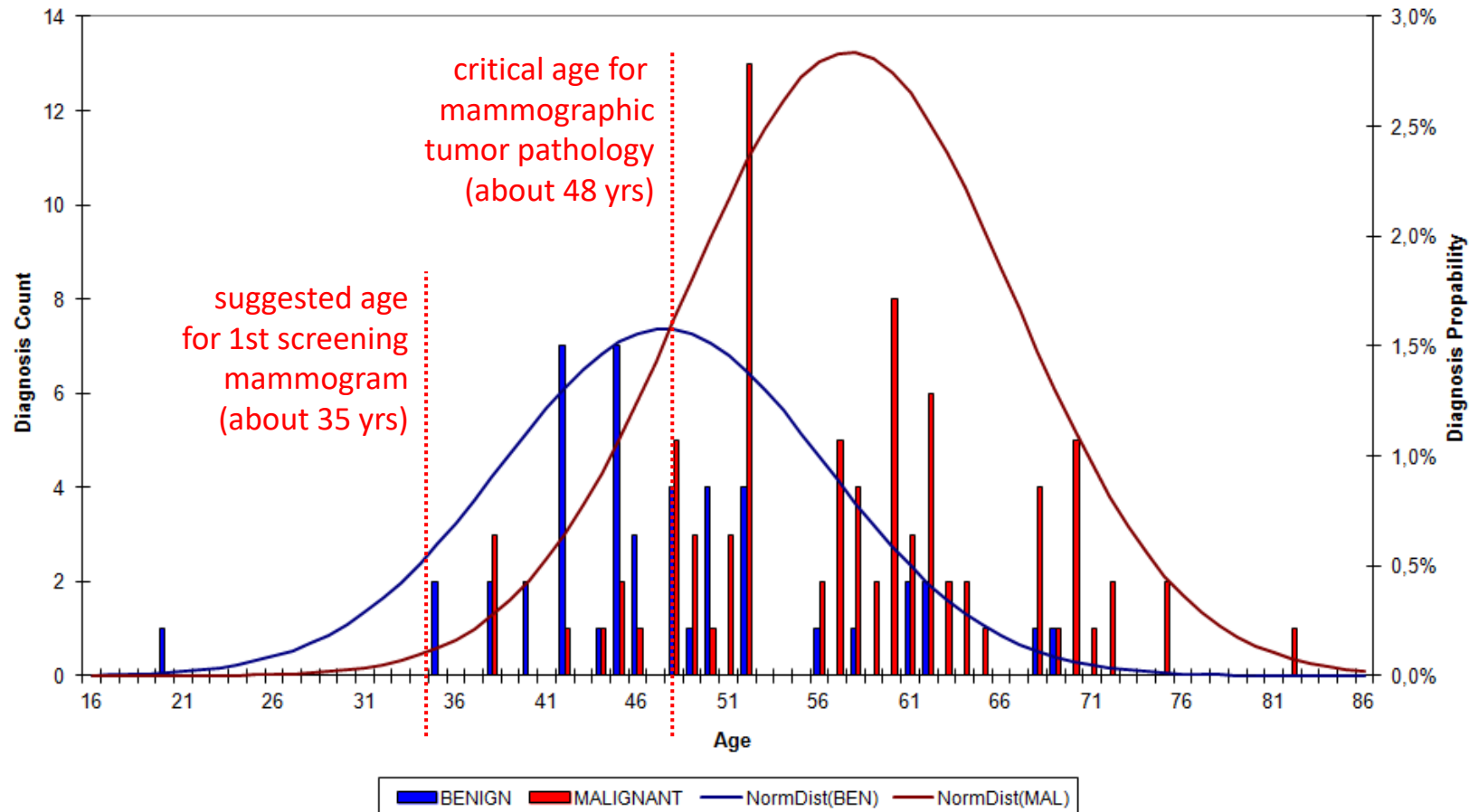
The RAND Corporation
1700 MAIN ST • SANTA MONICA • CALIFORNIA

Δεδομένα παντού ...

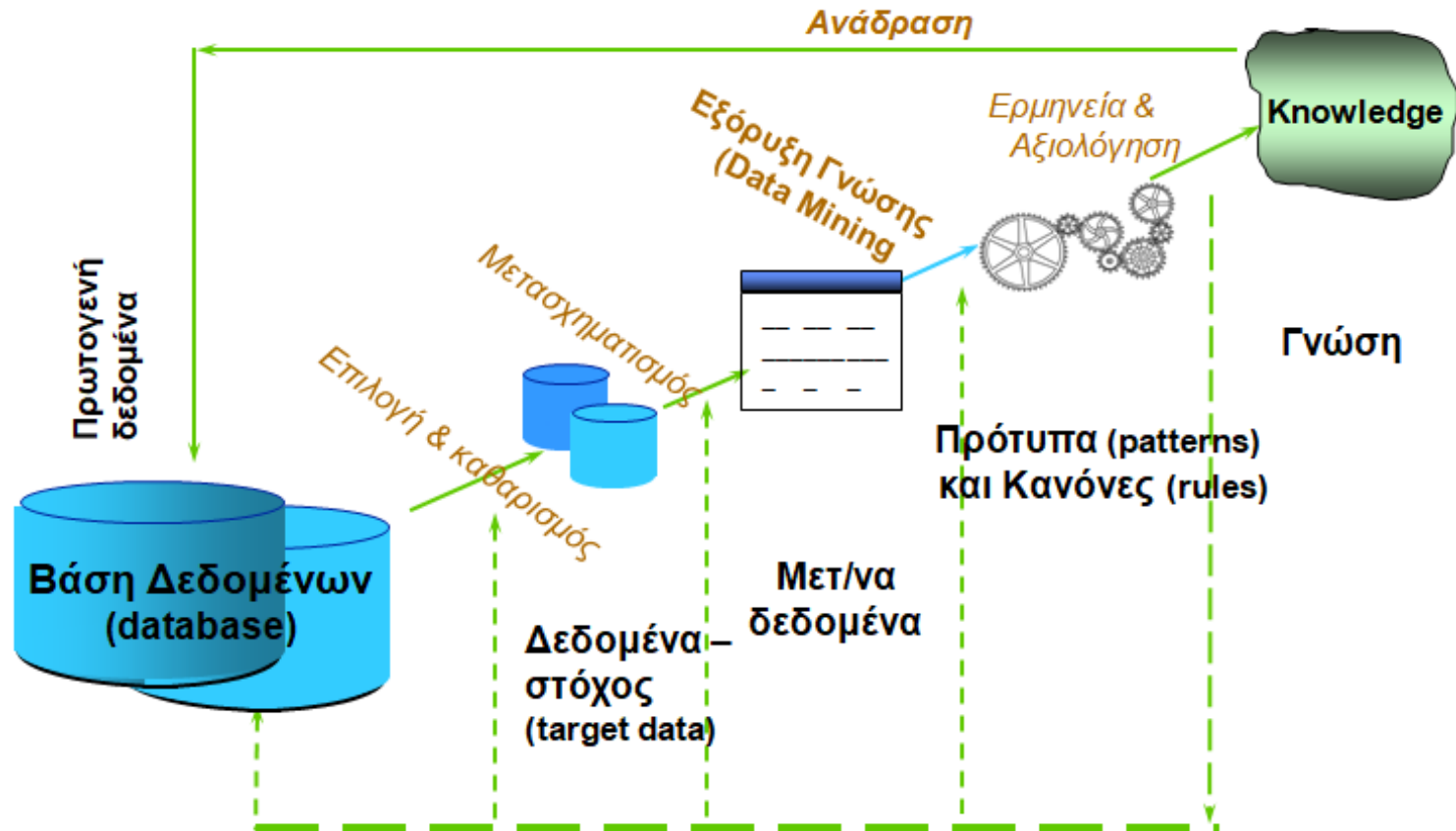
- Παράγονται όλο και περισσότερα δεδομένα:
 - Τραπεζικά, τηλεπικοινωνιακά, ...
 - Επιστημονικά δεδομένα: αστρονομικά, βιολογικά κλπ.
 - Κείμενα στο web κ.α.
- Αποθηκεύονται όλο και περισσότερα δεδομένα:
 - Γρήγορη / φθηνή τεχνολογία αποθήκευσης
 - Ικανά ΣΔΒΔ για μεγάλες ΒΔ



Age Distributions vs Benign/Malignant

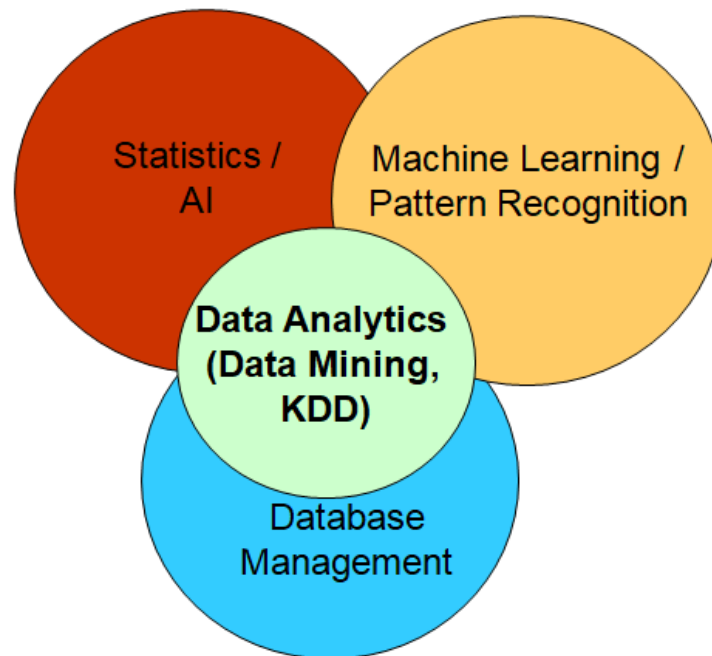


Η “σκάλα” της διαδικασίας KDD



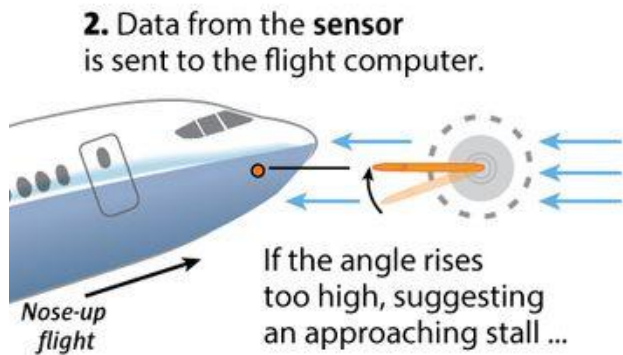
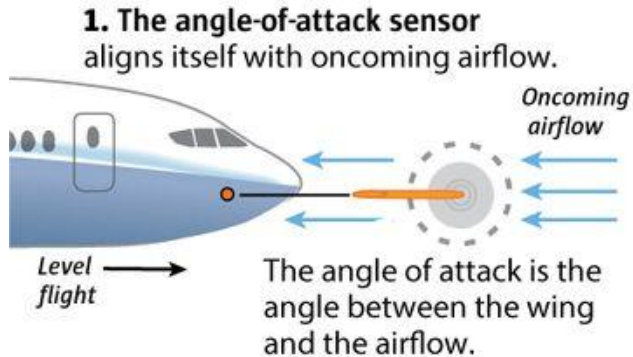
Σχετικά επιστημονικά πεδία

- Στατιστική / «Τεχνητή Νοημοσύνη», Μηχανική Μάθηση / Αναγνώριση Προτύπων, Διαχείριση Βάσεων Δεδομένων
- Οι παραδοσιακές τεχνικές επεξεργασίας δεδομένων που μας προσφέρουν αυτές οι επιστημονικές περιοχές μπορεί να είναι ανεφάρμοστες λόγω:
 - του μεγάλου όγκου,
 - των πολλών διαστάσεων,
 - της ετερογένειας των δεδομένων,
 - των απαιτήσεων επεξεργασίας,
 - ...

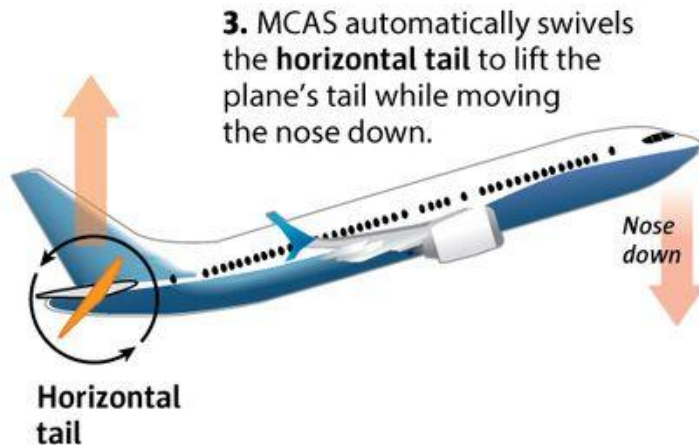


Worst Software Failures: Boeing 737 MAX

How the MCAS (Maneuvering Characteristics Augmentation System) works on the 737 MAX



... the MCAS activates.



Sources: Boeing, FAA, Indonesia National Transportation Safety Committee, Leeham.net, and The Air Current

Reporting by DOMINIC GATES,
Graphic by MARK NOWLIN / THE SEATTLE TIMES

Worst Software Failures: Tesla car crash



*The information in this preliminary report is
It will be supplemented or corrected du*

On Friday, March 23, 2018, about 9:27 a.m., Pacific electric-powered passenger vehicle, occupied by US Highway 101 (US-101) in Mountain View, S approached the US-101/State Highway (SH-85) i from the left, which was a high-occupancy-vehicle

According to performance data downloaded from driver assistance features traffic-aware cruise control Tesla refers to as “autopilot.” As the Tesla approached lanes of US-101 from the SH-85 exit ramp, it merged. Tesla continued traveling through the gore area at a speed of about 71 mph.² The crash attenuated barrier. The speed limit on this area of roadway is the traffic-aware cruise control speed was set to rotated the Tesla counterclockwise and caused a side Tesla was involved in subsequent collisions with Audi A4 (see figure 1).

PRELIMINARY REPORT

HIGHWAY

A preliminary review of the recorded performance data showed the following:

- The Autopilot system was engaged on four separate occasions during the 32-minute trip, including a continuous operation for the last 18 minutes 55 seconds prior to the crash.
- During the 18-minute 55-second segment, the vehicle provided two visual alerts and one auditory alert for the driver to place his hands on the steering wheel. These alerts were made more than 15 minutes prior to the crash.
- During the 60 seconds prior to the crash, the driver’s hands were detected on the steering wheel on three separate occasions, for a total of 34 seconds; for the last 6 seconds prior to the crash, the vehicle did not detect the driver’s hands on the steering wheel.
- At 8 seconds prior to the crash, the Tesla was following a lead vehicle and was traveling about 65 mph.
- At 7 seconds prior to the crash, the Tesla began a left steering movement while following a lead vehicle.
- At 4 seconds prior to the crash, the Tesla was no longer following a lead vehicle.
- At 3 seconds prior to the crash and up to the time of impact with the crash attenuator, the Tesla’s speed increased from 62 to 70.8 mph, with no precrash braking or evasive steering movement detected.



The New Caledonian crow uses twigs and branches to extricate grubs and insects from inside trees (Credit: Alamy)

Dolphin social intelligence: complex alliance relationships in bottlenose dolphins and a consideration of selective environments for extreme brain size evolution in mammals

[Richard C Connor](#)^{1,*}

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PMCID: PMC2346519 PMID: [17296597](https://pubmed.ncbi.nlm.nih.gov/17296597/)

Abstract

Bottlenose dolphins in Shark Bay, Australia, live in a large, unbounded society with a fission–fusion grouping pattern. Potential cognitive demands include the need to develop social strategies involving the recognition of a large number of individuals and their relationships with others. Patterns of alliance affiliation among males may be more complex than are currently known for any non-human, with individuals participating in 2–3 levels of shifting alliances. Males mediate alliance relationships with gentle contact behaviours such as petting, but synchrony also plays an important role in affiliative interactions. In general, selection for social intelligence in the context of shifting alliances will depend on the extent to which there are strategic options and risk. Extreme brain size evolution may have occurred more than once in the toothed whales, reaching peaks in the



Octopuses Capable of Hand-Eye Coordination

By Helen Albert, CosmosMagazine.com

May 30, 2011

LONDON: Octopuses are able to use visual cues to guide a single arm to a location, a complex movement that was not thought possible due to their lack of a rigid body structure, say researchers.

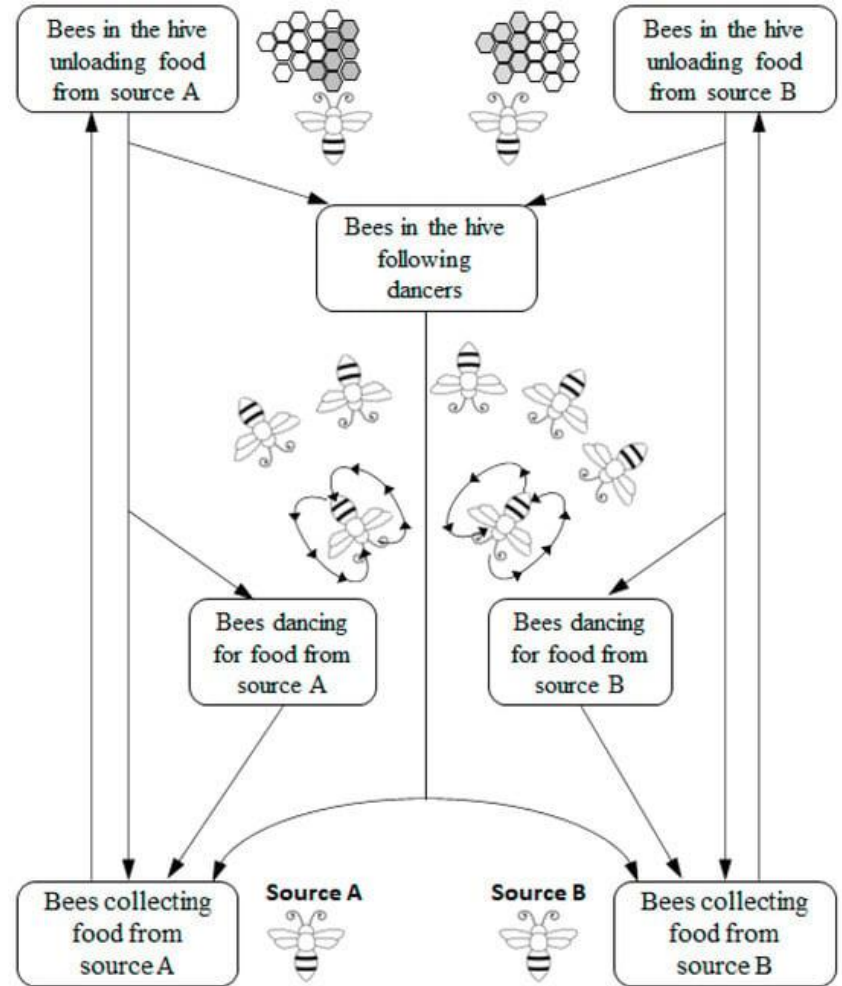
The octopus' arm is made up primarily of muscle with no skeletal support, so octopuses were previously believed to have a low level of body awareness and only limited control over their limbs. However, this study has shown for the first time that they can direct a single arm in a complex movement to a target location.

"Octopuses have a central nervous system that is advanced for an invertebrate, but simple compared to a vertebrate, yet it is capable of controlling a much more 'difficult' arm," said lead study author Tamar Gutnick, a researcher at the Hebrew University of Jerusalem in Israel.

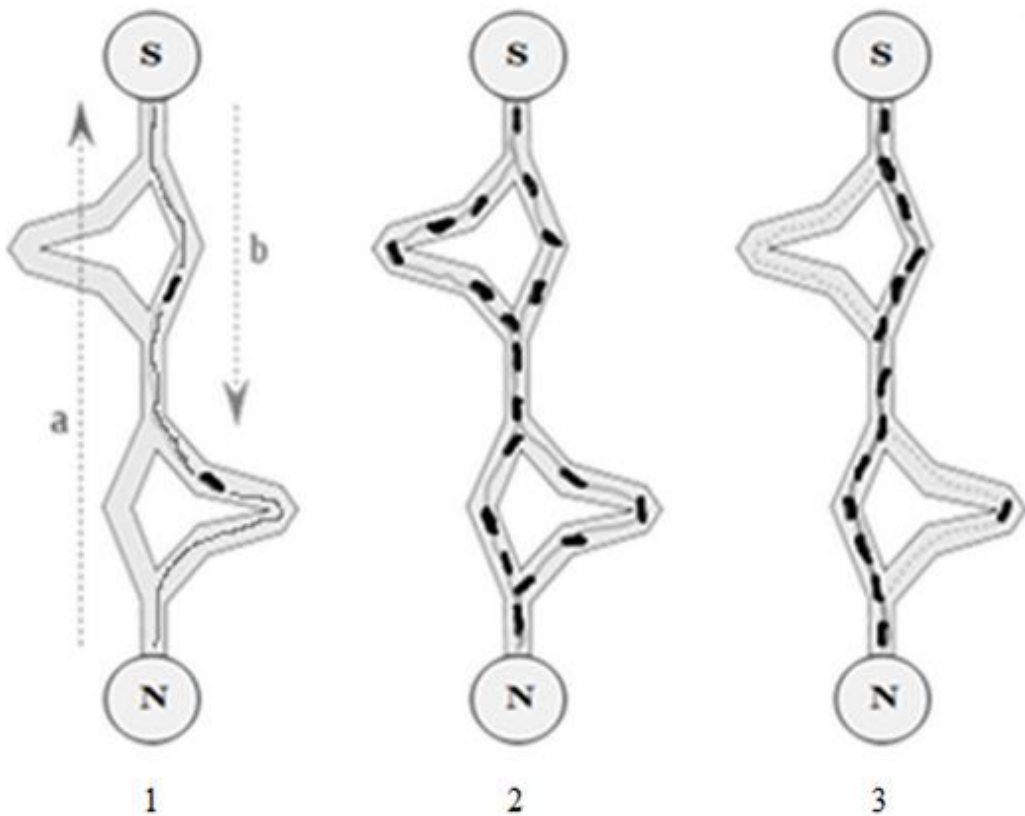
"Because of the unique body plan of the octopus its ability to control a single arm in a complex movement is quite amazing."



Bee Colony Optimization (ACO)



Ant Colony Optimization (ACO)





Swarm Intelligence (SI)








Particle swarm optimization and RBF neural networks for public transport arrival time prediction using GTFS data

[Eva Chondrodima](#)^a  , [Harris Georgiou](#)^a, [Nikos Pelekis](#)^b, [Yannis Theodoridis](#)^a


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<https://doi.org/10.1016/j.jjime.2022.100086> 

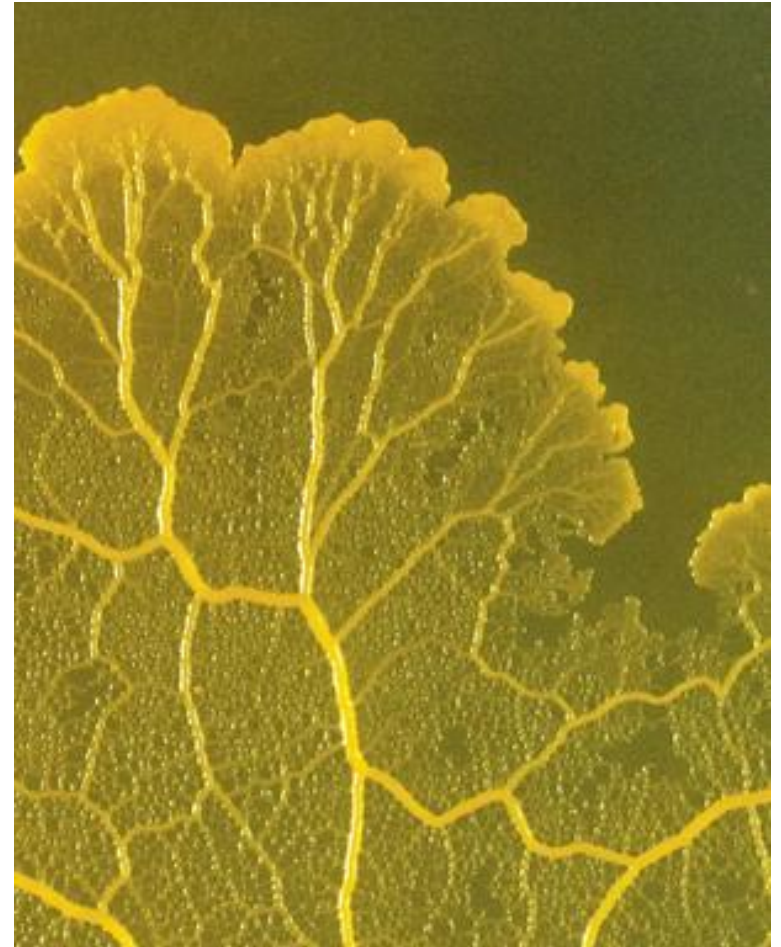
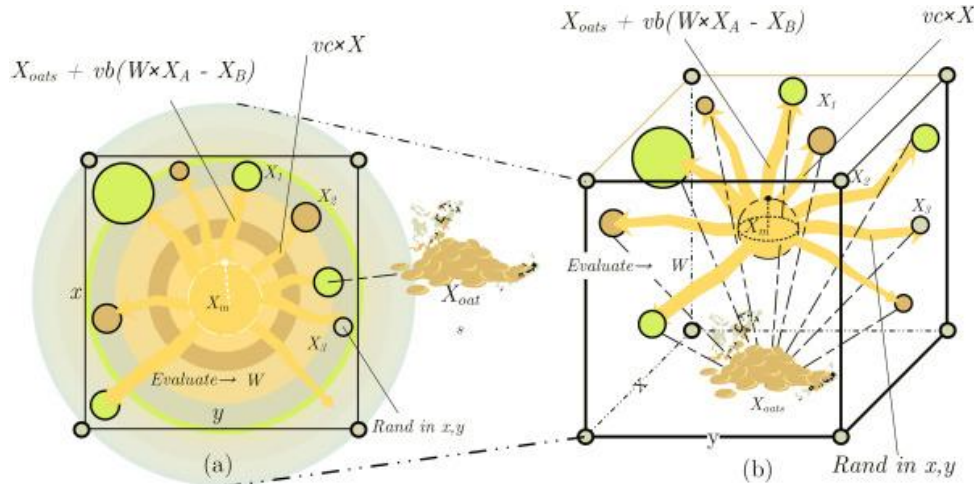
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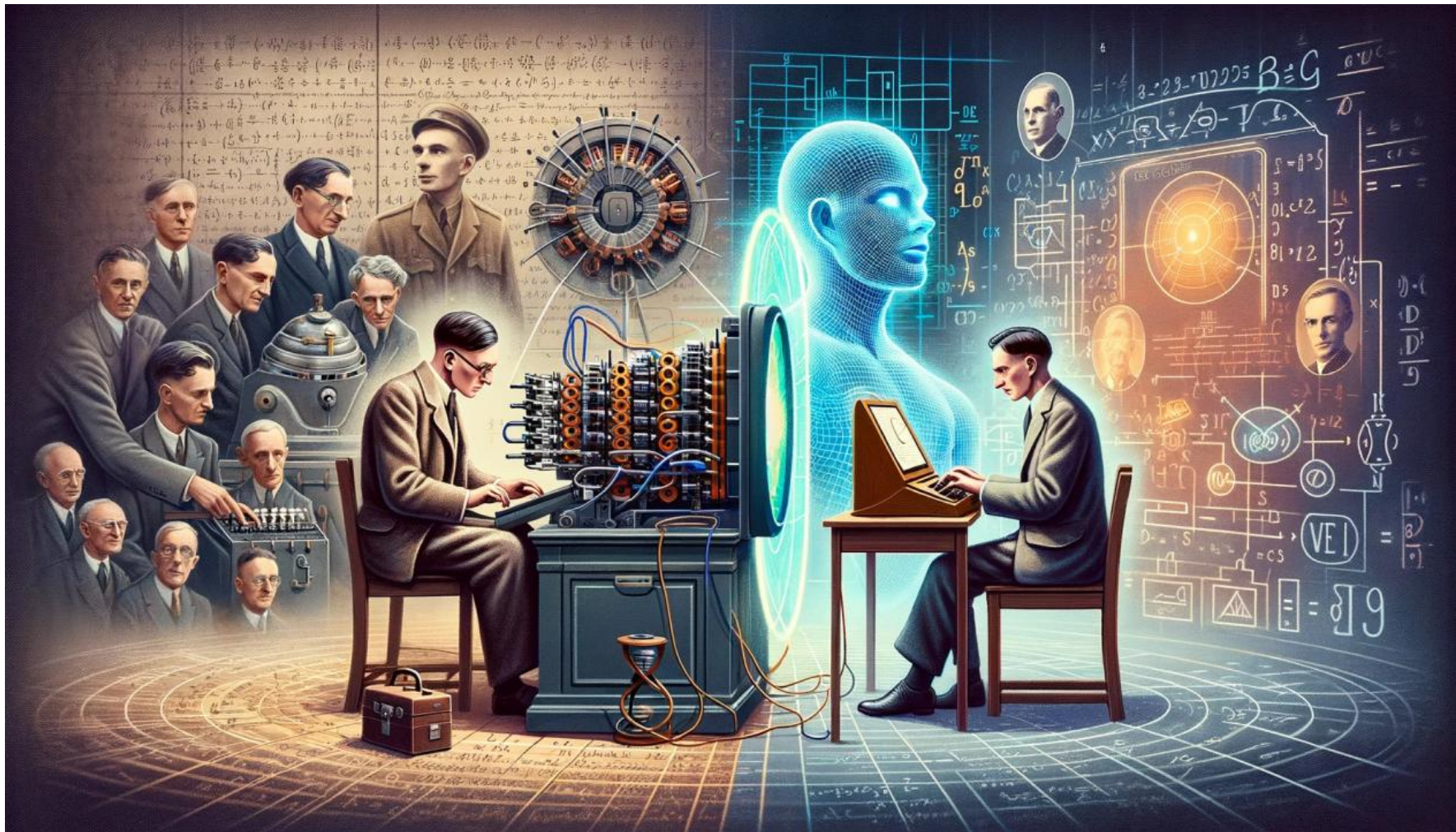
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Slime mould algorithm: A new method for stochastic optimization

Shimin Li ^a✉, Hailing Chen ^a✉, Mingjing Wang ^e✉, Ali Asghar Heidari ^{b c}✉, Seyyedali Mirjalili ^d✉





Μέρος II: Μελλοντικές προοπτικές

1. Turing Completeness
2. Προσομοιώνοντας τη Φύση
3. Τι έρχεται στο μέλλον;
4. Πως θα τα καταφέρουμε;



What Does Turing Complete Mean?

Key characteristics that define a Turing complete system



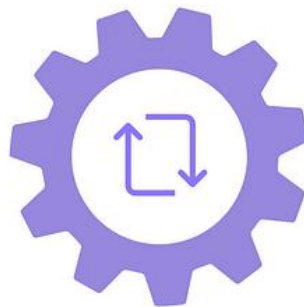
Sequence

Execute a series of computational steps **in the order** they are provided.



Conditionals

Execute different computational paths conditionally **based on certain criteria**.



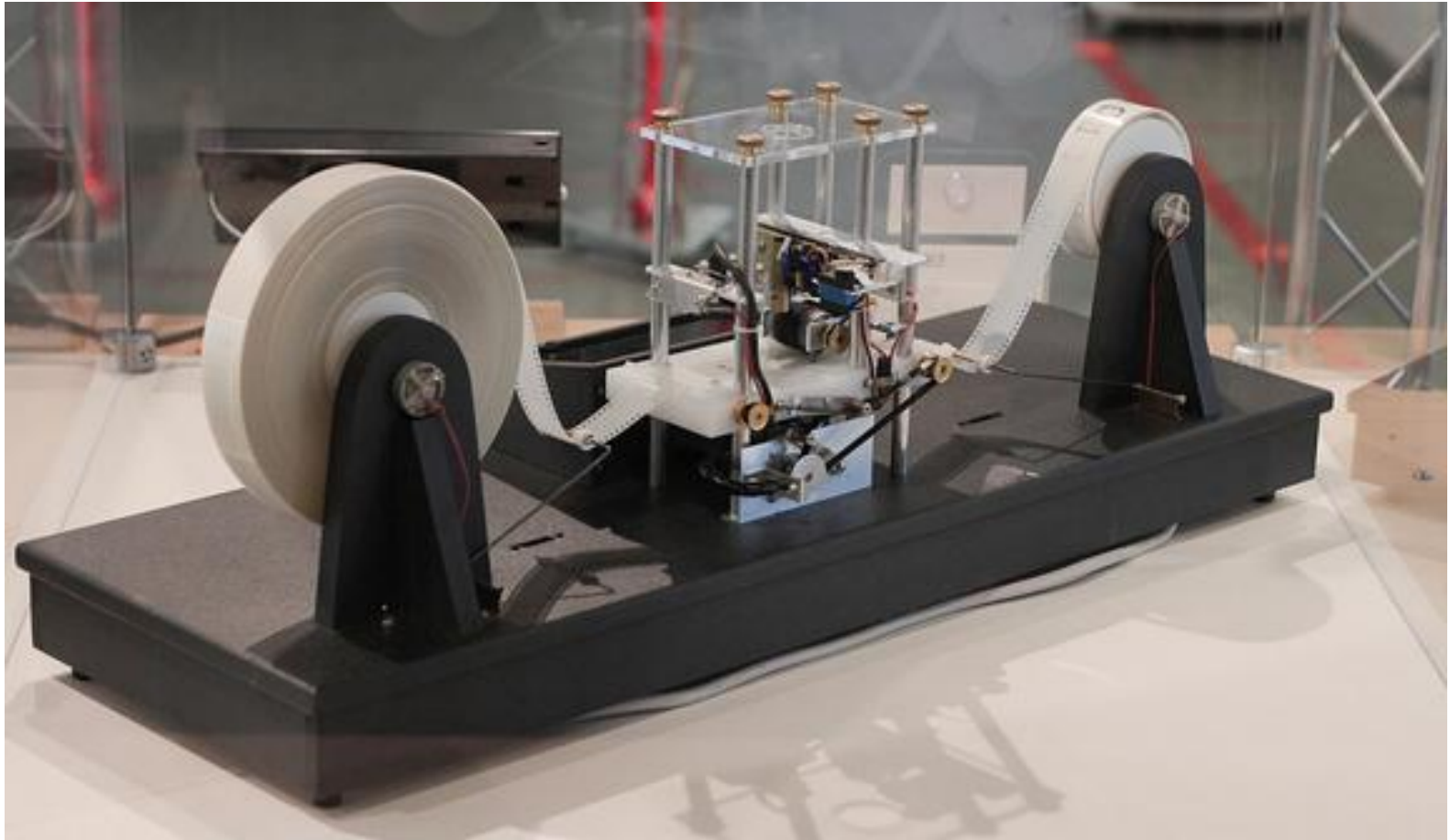
Iteration

Repeat computational sub-processes **over and over**.



Store Data

Store intermediate results for **later use in memory**.

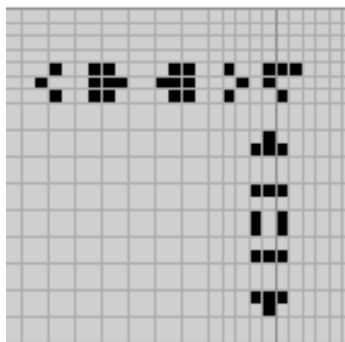


Details: https://en.wikipedia.org/wiki/Turing_machine

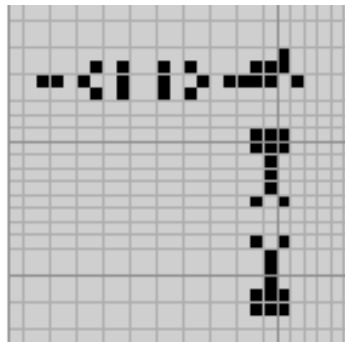
Counter machines

- ▶ Can be thought in one of two ways.
- ▶ Same structure as multistack machine.
- ▶ Each stack in the machine is replaced by a counter.
- ▶ Counter hold any nonnegative integer.
- ▶ The move of a machine depends on the
 - 1.current state
 - 2.current input symbol
 - 3.if one of the value of the counter is zero

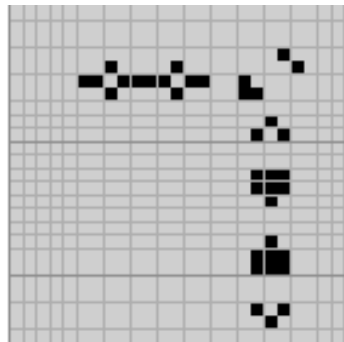




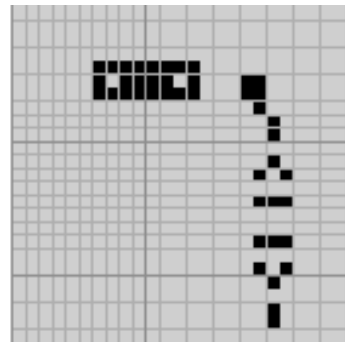
(a)



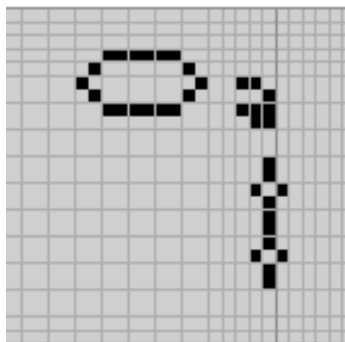
(b)



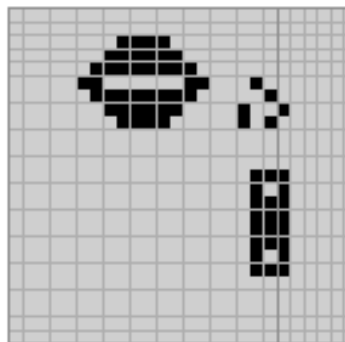
(c)



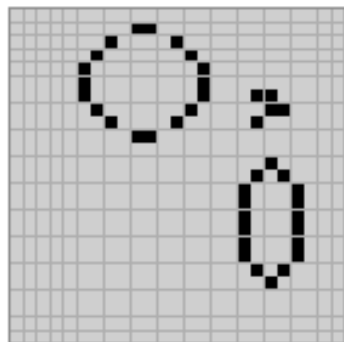
(d)



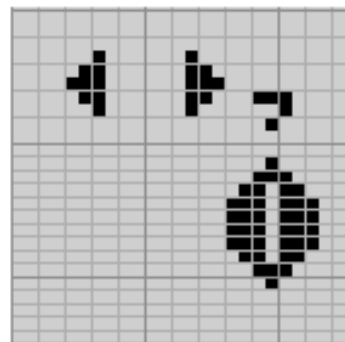
(e)



(f)



(g)



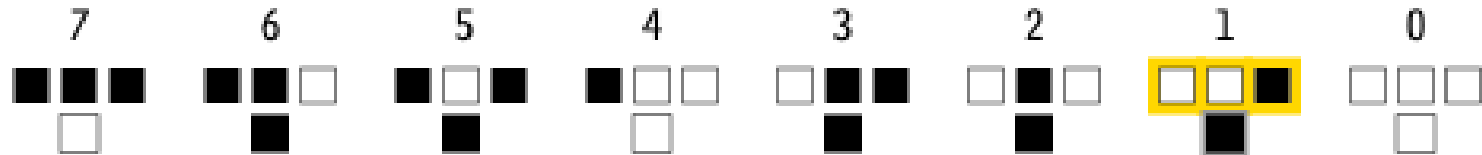
(h)

Details: https://en.wikipedia.org/wiki/Conway%27s_Game_of_Life

current automaton contents



rule 110 (01101110)



the next generation of the automaton



SUBLEQ - A One Instruction Set Computer (OISC)

29 May 2020 Lawrence Woodman

#Programming #SUBLEQ

SUBLEQ has to be one of the easiest architectures to implement in either software or hardware and this is the main reason for its design as a teaching aid. It has only one instruction, hence why it is called a One Instruction Set Computer (OISC), which isn't the best name considering that most processors have one instruction set. URISC is good, but perhaps One Instruction Computer (OIC) would be more accurate.

SUBLEQ (SUBtract and Branch if Less than or Equal

The SUBLEQ instruction stands for *SUBtract and Branch if Less than or Equal to zero*. Because there is only one instruction, only two are specified, which consist of 3 memory addresses that are

; Outputs "HELLO, WORLD!\n"

.equ OUT -1

.equ HALT -1

loop: sble hello OUT ; Ou

sble #-1 loop ; In

sble #-1 checkEnd+1 ; In

checkEnd: sble z hello HALT ; Ha

sble z z loop ; Ju

hello: .asciiz "HELLO, WORLD!\n"

z: .word 0

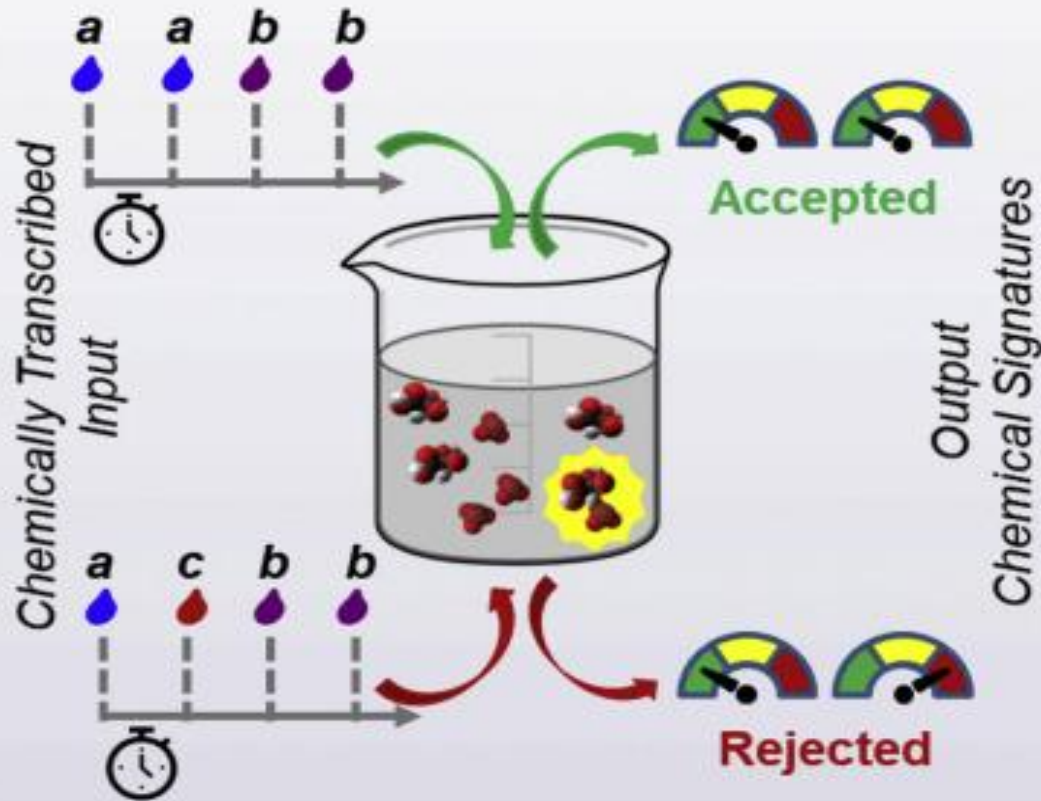
Brainfuck Programming Language

- created in 1993 by Urban Müller
- Only 8 instructions
 - Müller's Amiga compiler was 240 bytes in size
 - x86/Linux by Brian Raiter had 171 Bytes!

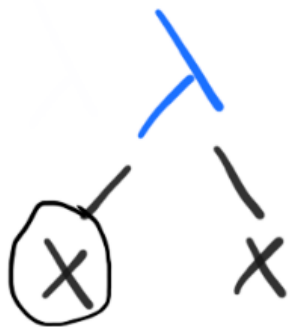
```
+++[>+++++[>++++>++++>++++>+  
++>+<<<<<-]>++++>+++>+++>+<<<<<  
-]>>.>.>+.>.>-----.
```

```
<      Goto T1  
[      While T1 is nonzero  
  >      Goto T2  
  [->+>+<<]  Copy T2 to T3 and T4  
  >>      Goto to T4  
  [-<<+>>]  Move T4 to T2  
  <      Goto to T3  
  [-<<<+>>>] Accumulate T3 onto T0  
  <<      Goto to T1  
  -      Decrement T1  
]
```

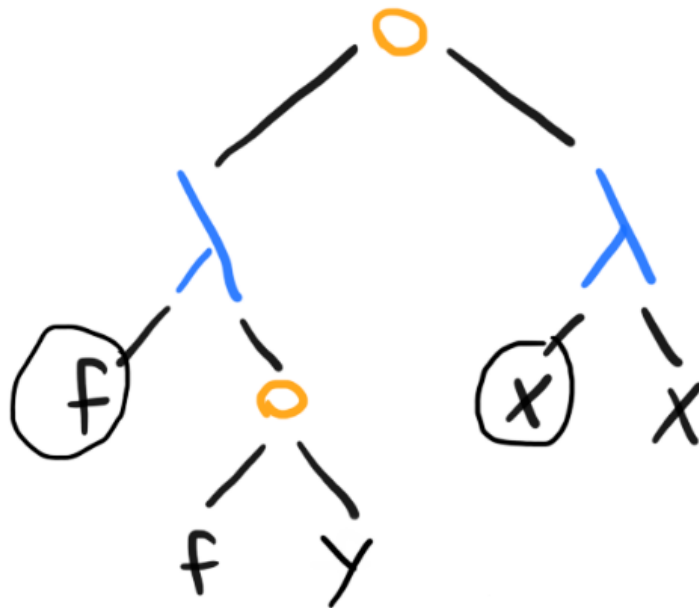
How chemistry computes



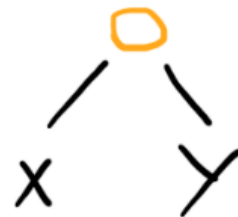
λ -Calculus (A. Church, 1936)



$(\lambda x. x)$



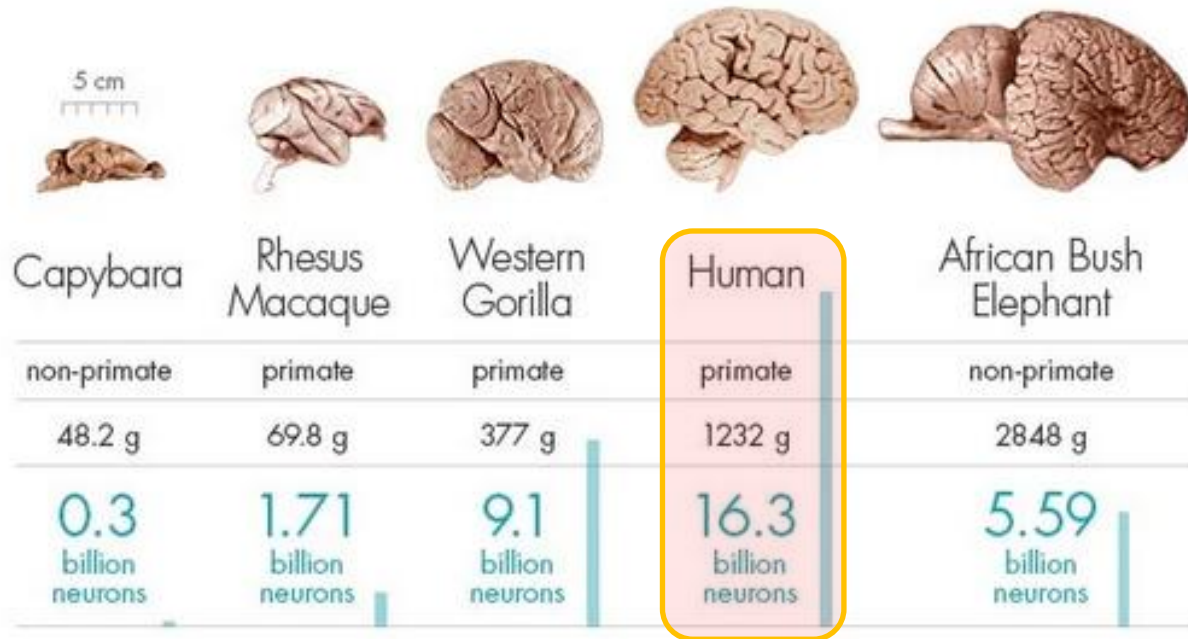
$((\lambda f. (f y)) (\lambda x. x))$



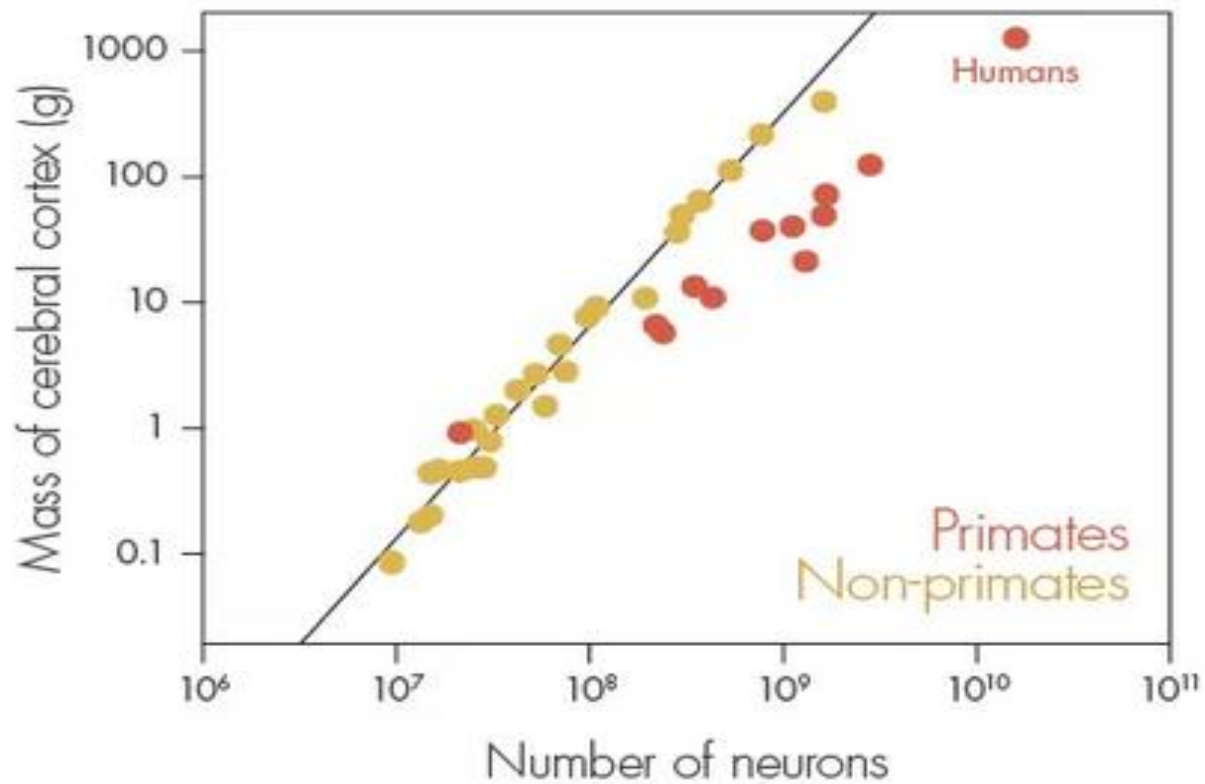
$(x y)$

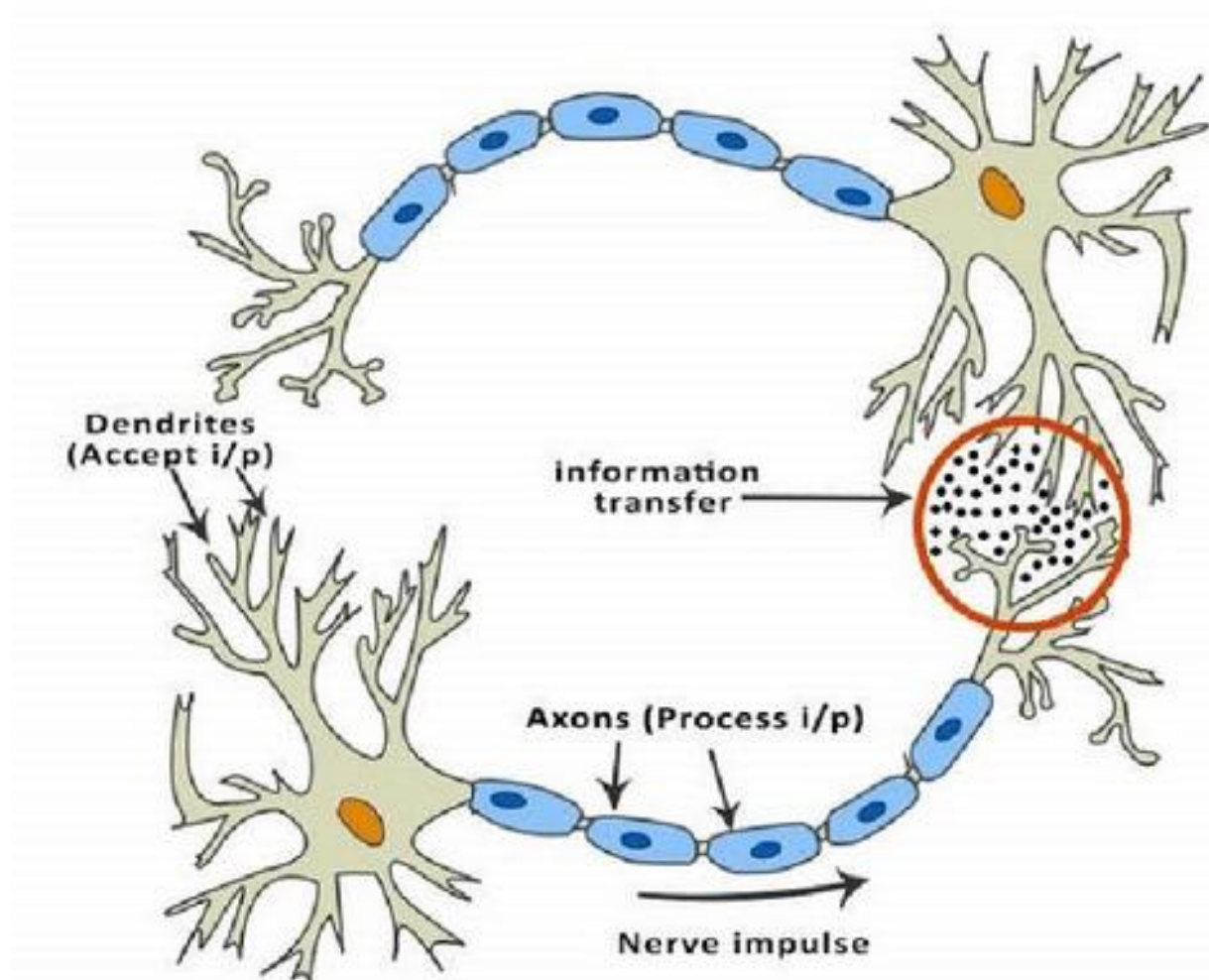
BRAIN SIZE AND NEURON COUNT

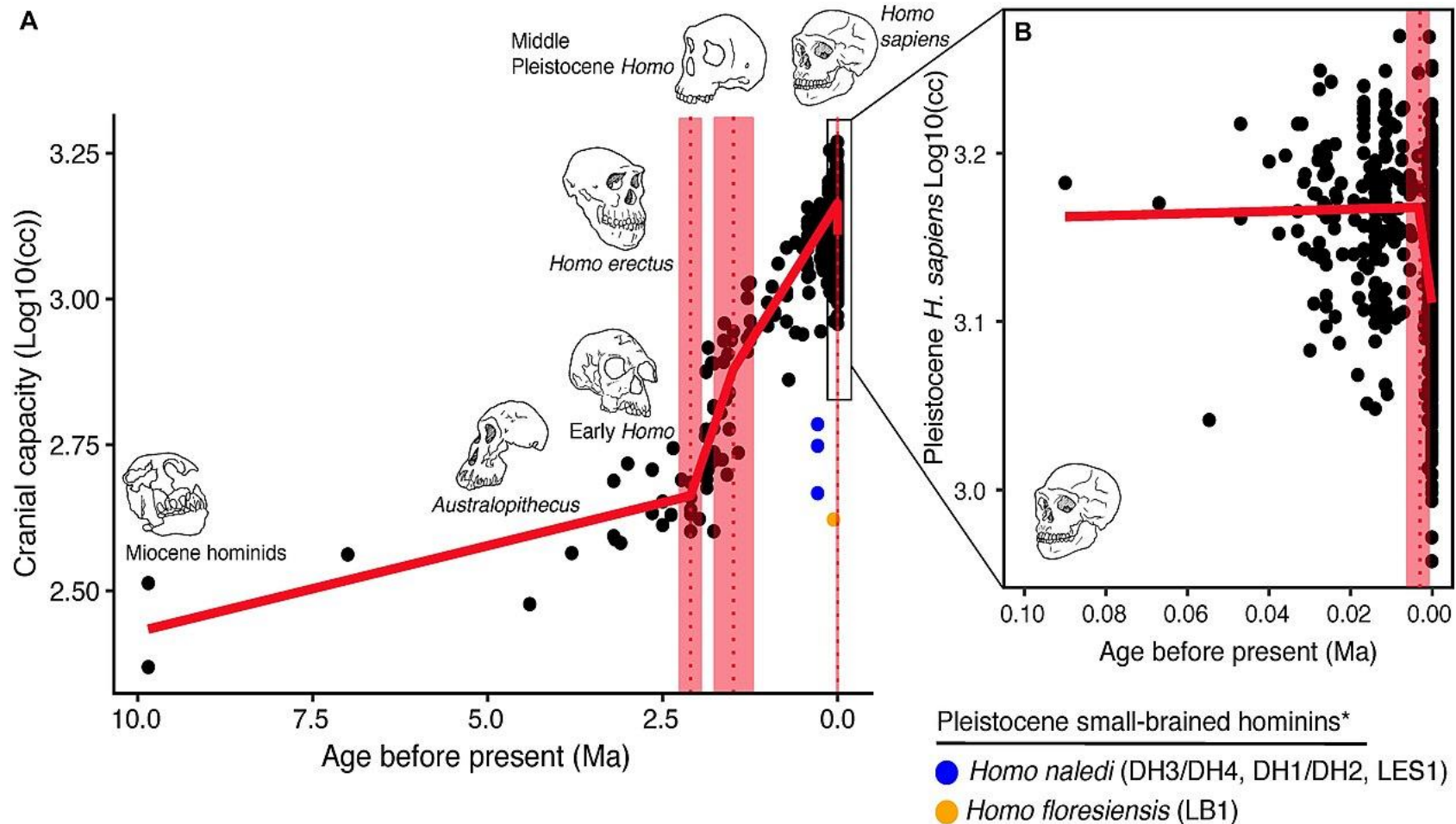
Cerebral cortex mass and neuron count for various mammals.



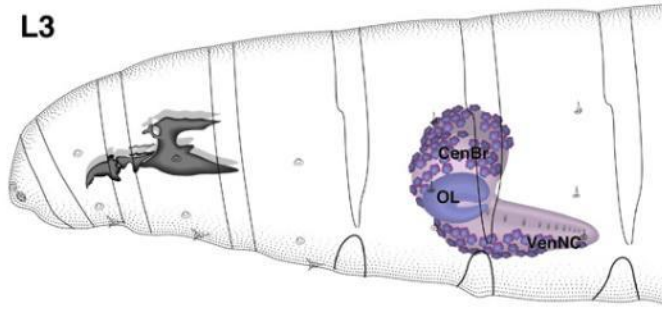
BRAIN DENSITY



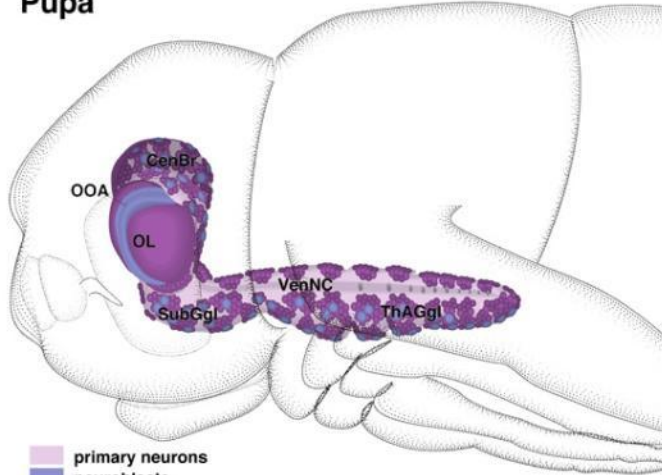




L3

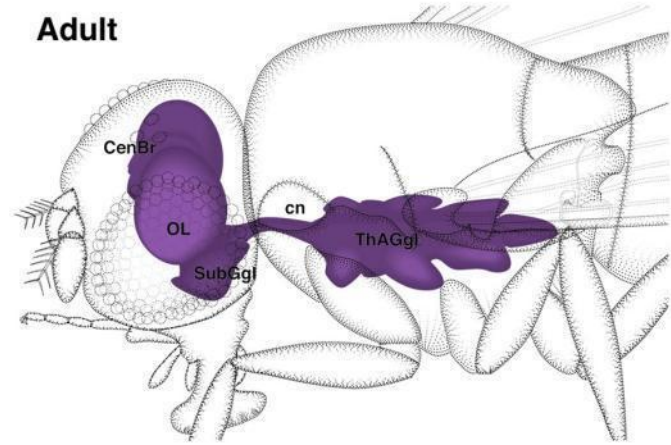


Pupa

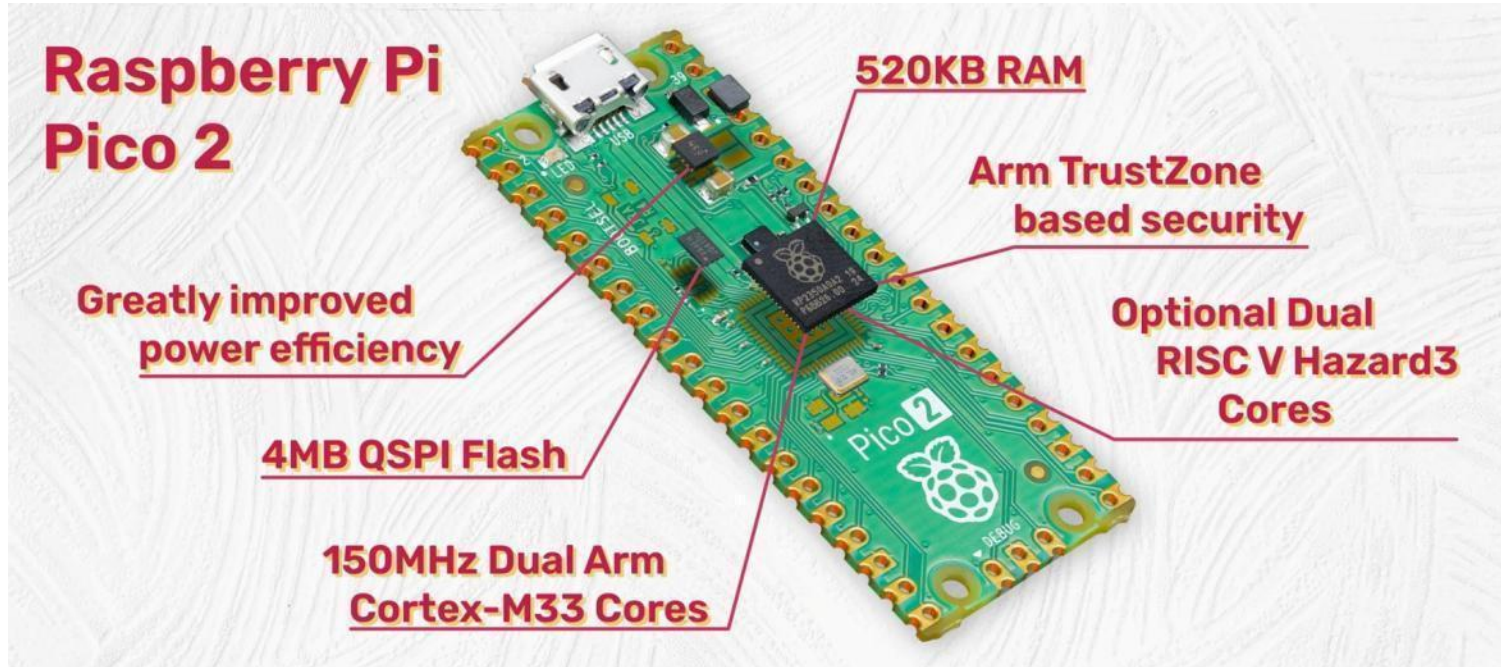


primary neurons
neuroblasts
secondary neurons

Adult



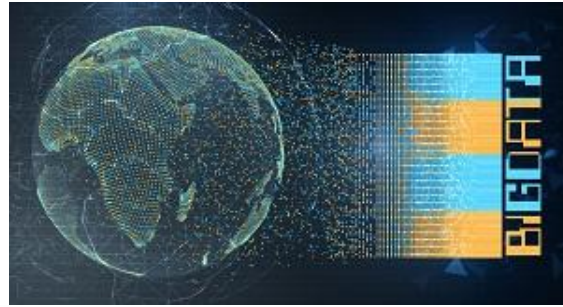
- Earthworm: $n=302$ ($s=7,500$)
- Sea slug: $n=18,000$
- Fruit fly: $n=150,000$
- Ant: $n=250,000$
- Honey bee: $n=960,000$ ($s=10^9$)
- Cockroach: $n=10^6$
- Brown rat: $n=31 \times 10^6$ ($s=3 \times 10^6$)



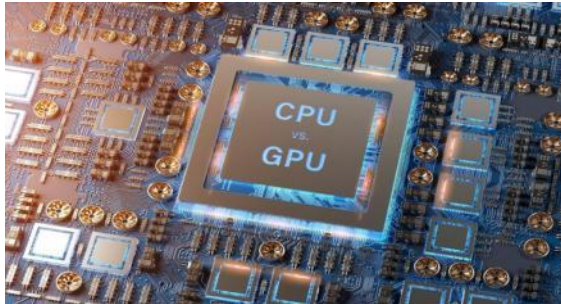
- 520KB RAM \approx 133,120 “synapse” weights (32-bit FP variables)
- FP arithmetic (FxDadds) \approx 300 per sample @ 44100 khz
- FP benchmark \approx 1.5 MFLOPS @ 133 MHz (\approx 11 Hz “brain”)
- *RPi 5 (SBC): 4 cores / x16 memory / x37 speed*

Τι έχει αλλάξει τα τελευταία 10-15 χρόνια;

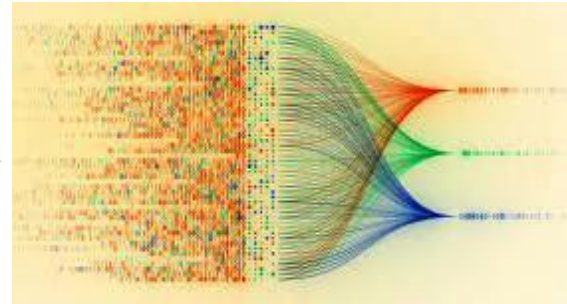
Big (Open) Data

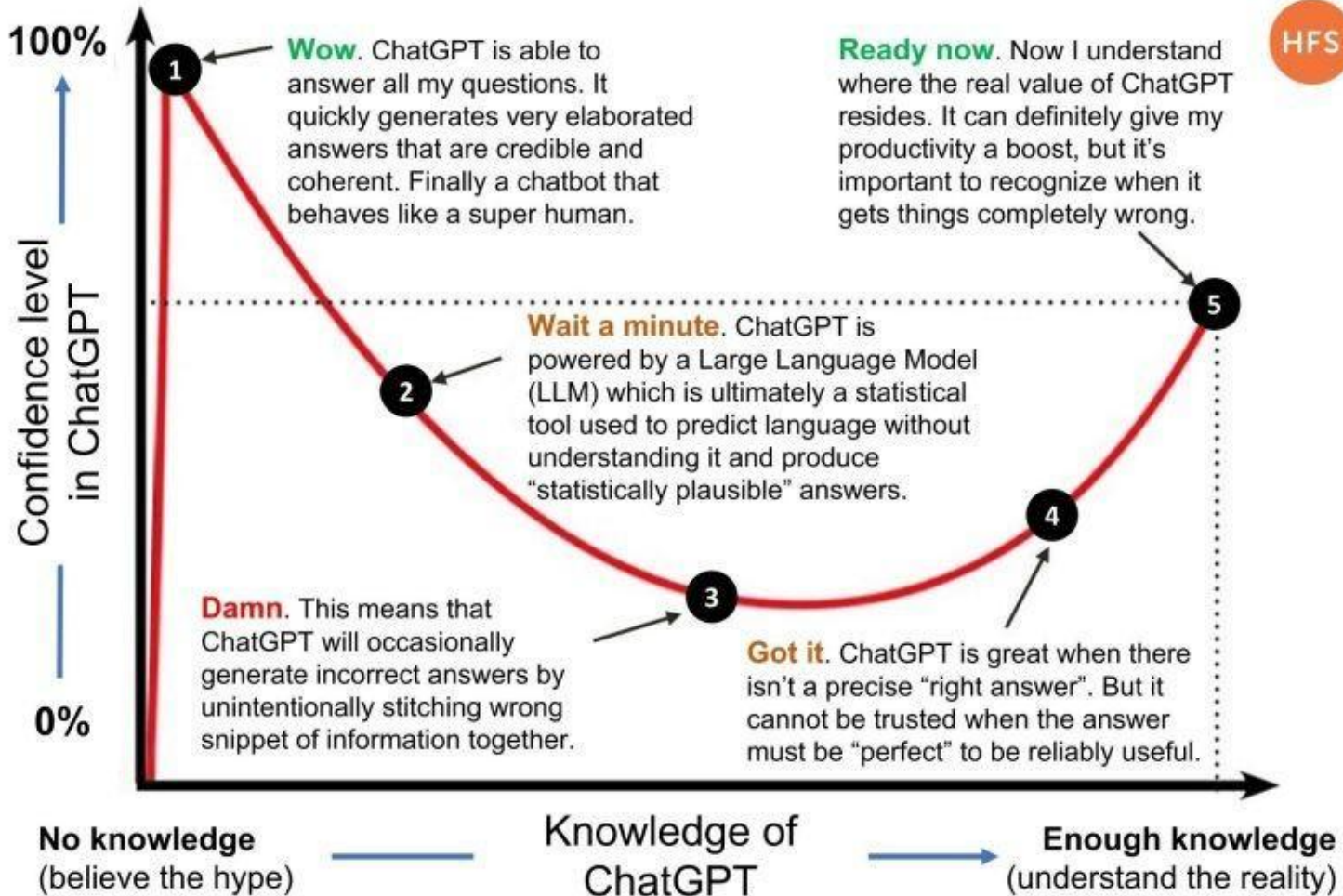


GPU accelerators



Deep Learning,
Transformers,
LLMs, ...

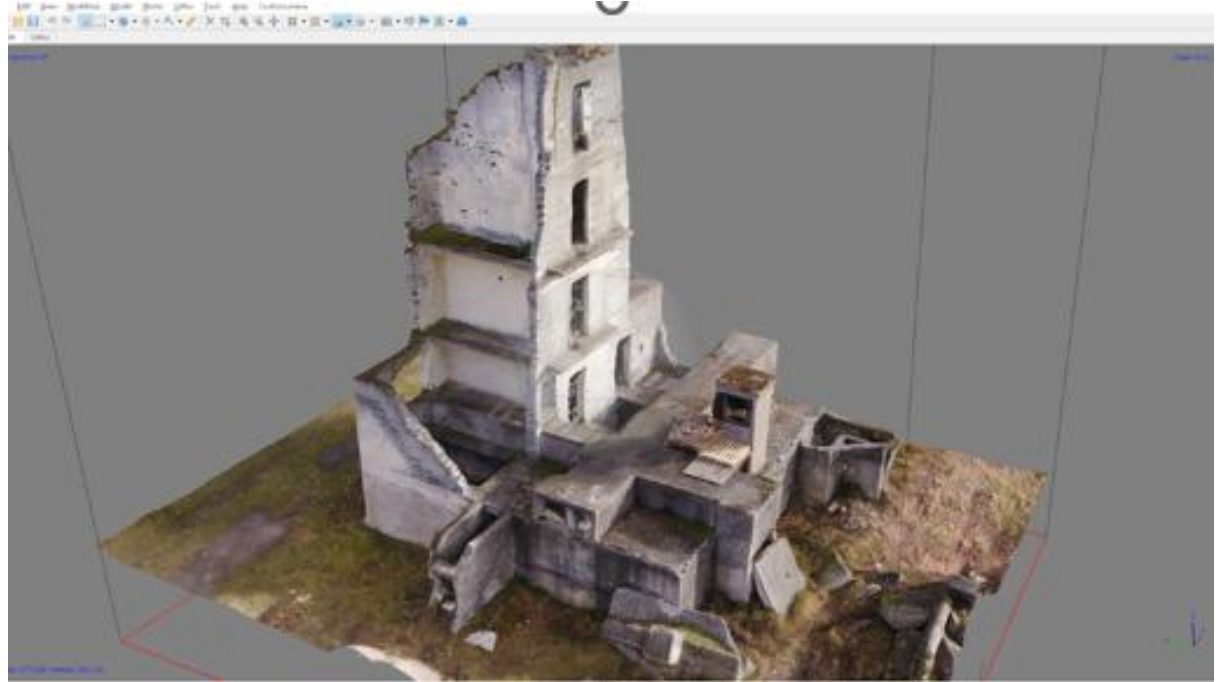




Source: A. AlQuraini, "ChatGPT as seen through the Dunning-Kruger Curve" (26/2/2023)

Modelling Swarm Drones:

Creation of orthophoto and 3-D model of disaster area with 5 pre-programmed drones

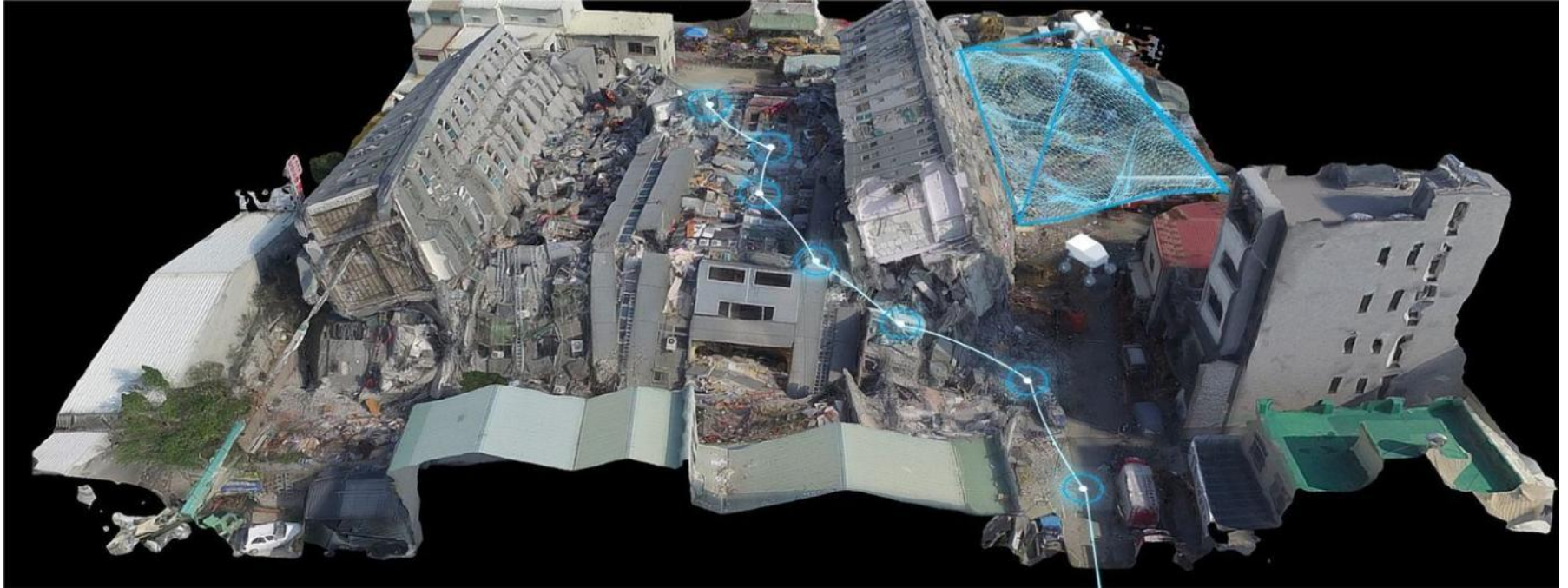


Credits: CURSOR project (EU H2020)

Earthquake in Turkiye



Live 3-D optimal route planning: Risk assessment and FR team navigation inside the hotzone



Credits: INTREPID project (EU H2020)

Future trends: Body part detection & assessment

Data-Driven Skin Detection in Cluttered Search and Rescue Environments

Yogeshwar Singh Dadwal[✉], *Student Member, IEEE*, Satish Kumar, *Member, IEEE*, and H. K. Sardana[✉], *Member, IEEE*

Abstract—Locating human victims in cluttered urban search and rescue (USAR) environments is still a challenge. In this paper, we present an approach to generate *skin objectness* windows to assist human rescuers. We introduce the term *skin objectness* to denote the task of extracting windows in the scene with a high probability of skin presence for locating victims. Unlike naive skin segmentation approaches, the presented algorithm accounts for both color and spatial information to extract regions of interest and at the same time, rejects the background clutter. We use temporal information of the video sequence to make the skin objectness windows more reliable. To selectively boost skin regions, the RGB skin pixels are transformed to Gabor space to generate a transformation matrix. The matrix is used to generate skin. Further, the Bayesian inference and temporal cues from previous frames are used to generate skin objectness windows. It has real-time applications in image retrieval, action classification, and disaster relief. The proposed method in cluttered environments.

Index Terms—Distinctiveness, Gabor filters, color image analysis

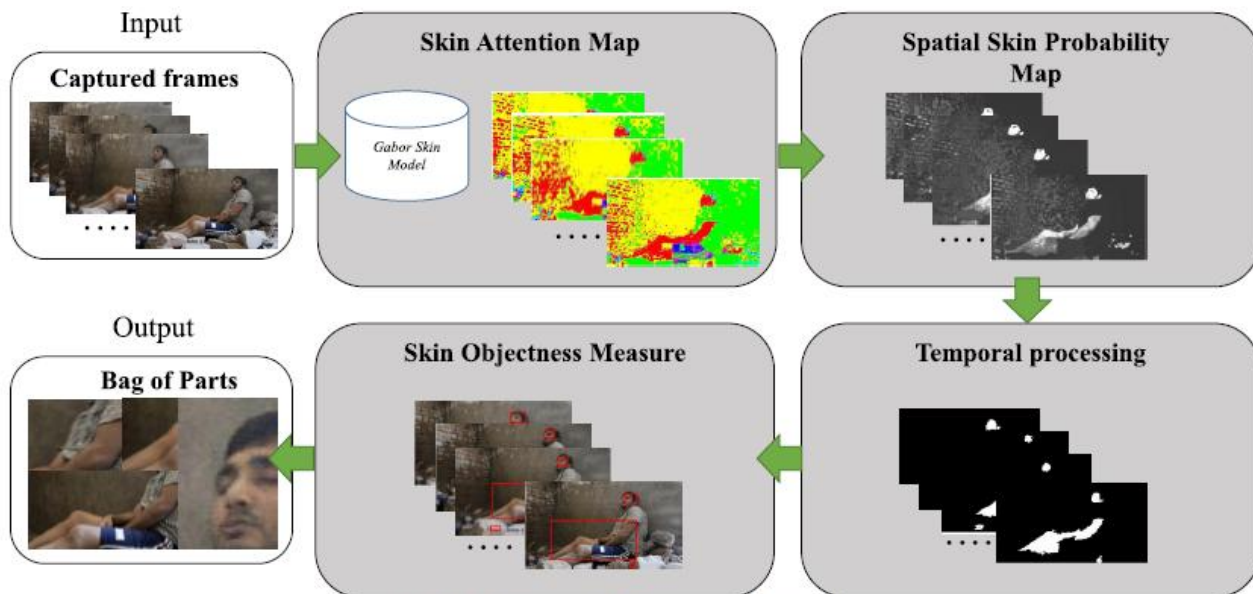


Fig. 1. Framework of the proposed algorithm.

FR safety and remote sensing:

Mapping and navigation inside the hotzone with sensors, AR and UxVs



Τροφή για σκέψη...

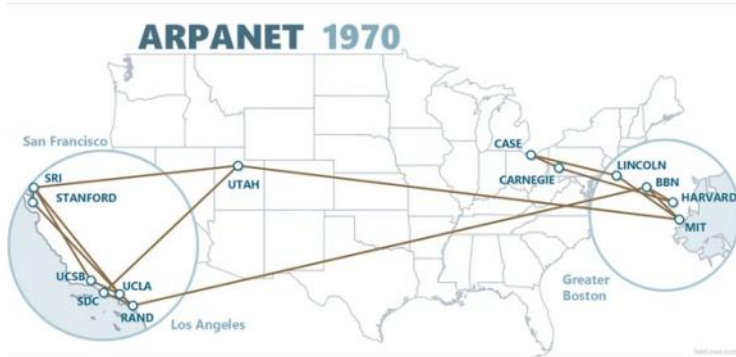
- A.I. = «πολλαπλασιαστής»
- Generative A.I.
- Large Language Models (LLM)
- Transformers
- Big (Open) Data
- Cloud → Edge → IoT
- EU A.I. Act, USA Exec. Order

- Η ελληνική πραγματικότητα
- Brain drain
- «Έλλειψη ταλέντων»...
- «Πληροφορικάριοι»...
- Ψηφιακός αναλφαβητισμός
- «Φορολόγηση των ρομπότ»
- «Dark Internet forest»

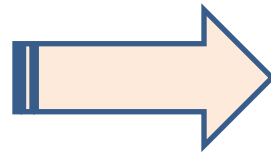
Είμαστε έτοιμοι ? → Φυσικά όχι (όπως πάντα άλλωστε)

Τι χρειάζεται για να πετύχει;

1. Αφοσίωση και όρεξη + Επιστημονική επάρκεια
2. Απαραίτητοι πόροι, κυρίως δημόσιοι στο ξεκίνημα
3. Κυρίως όραμα για το που πάμε και με τι στόχο



23 χρόνια





Source: "13 minutes to the Moon" (podcast) – <https://www.bbc.co.uk/programmes/w13xttx2>

Σύνοψη

- Περιεχόμενα:
 - Μέρος I: Τεχνητή Νοημοσύνη – Βασικές Έννοιες
 - Μέρος II: Μελλοντικές Προοπτικές
- Σχετικό υλικό:
 - «Turing machines explained visually» – https://www.youtube.com/watch?v=-ZS_zFg4w5k
 - «Game of life: computer with display» – <https://www.youtube.com/watch?v=Wfuhbl8HE7s>
 - Computer History Museum (CHM) – <https://www.youtube.com/@ComputerHistory>
 - «Artificial Intelligence | 60 Minutes Full Episodes» – <https://www.youtube.com/watch?v=aZ5EsdnpLMI>
 - «Queens puzzle solver in LISP», @ApneaCoding – https://youtu.be/_1CRCyklUto
 - «BAM neural network in Arduino», @ApneaCoding – <https://youtu.be/RkM-rpSVD4I>

```

MOVE 1 TO DATA-C(N-T).
ADD 1 TO N-CHANGED.
GO TO LOOP-SCAN.
SELECT-CL2.
ADD DATA-X(N-T) TO SUM2-X.
ADD DATA-Y(N-T) TO SUM2-Y.
ADD 1 TO N-CL2.
IF DATA-C(N-T) EQUAL 2 GO TO LOOP-SCAN.
MOVE 2 TO DATA-C(N-T).
ADD 1 TO N-CHANGED.

```

```

91 id : Integer := 0; -- target ID (counter)
92 det : Integer := 0; -- detection slots in sequence
93 pur : Integer := 0; -- rel. power of detection
94 pur0 : Integer := detLimit; -- rel. power baseline (adapt
95 disp : Boolean := False; -- target reporting (flag)
96
97 begin
98 -- process the FOV slots --
99 for p in 1..(seekerData'length)-1 loop
100 -- rel. power is current detection 'step'
101 pur := abs(seekerData(p+1)-seekerData(p));
102 if pur >= detLimit then
103 -- detection valid, continue analysis
104 if pur > pur0+detLimit then
105 -- strong new 'step' from baseline (new target)
106 pur0 := pur; -- update the baseline
107 det := 0; -- reset the run-length
108 disp := False; -- enable target reporting
109 end if;
110
111 det := det + 1;
112
113 if
114
115 d:
116 p:
117 d:

```

Ανάλυση Ιατρικής Εικόνας μέσω Η/Υ

- Εικόνα (2-D): Επικλυτόμενες δομές ιστών
- Τομογραφία (3-D): Όγκος πληροφοριών
- Διαφορετικές τεχνολογίες απεικόνισης
- Διαφορετικά διαγνωστικά χαρακτηριστικά
- Η διαγνωστική πληροφορία συνήθως δεν είναι καλώς ορισμένη (θόρυβος, ασάφειες δομών)
- Η διαγνωστική διαδικασία είναι συνήθως ασαφής, πολύπλοκη και βασίζεται στην εμπειρία (ιατρός)

⇒ Η χρήση Η/Υ επιτρέπει την αυτόματη επεξεργασία και ενοποίηση (τομογραφία) μεγάλου όγκου δεδομένων απεικόνισης

⇒ ...αλλά εξακολουθεί να έχει σημαντικούς περιορισμούς ως προς τη σημασιολογική ερμηνεία τους (διαγνωστική πληροφορία)

Παράδειγμα τριδιάστατης ανατομικής μαγνητικής τομογραφίας (γυναικα) - Wikipedia.org

- Hamming (7,4) error correction codes in **R**
- Kmeans clustering in **COBOL**
- Bi-directional Associative Memory (BAM) in **Arduino/C**
- Linear Regression in **SQL, Matlab**
- ...

YouTube:

@ApneaCoding



<https://www.youtube.com/@apneacoding>

<https://www.facebook.com/apneacoding>

Github:

@xgeorgio



<https://github.com/xgeorgio>

<http://apneacoding.blogspot.com>

Ένας ψηφιακός κόσμος γεμάτος γνώση για όλους

Σύμφωνα με το Καταστατικό της Ένωσης Πληροφορικών Ελλάδας, ένας από τους βασικούς σκοπούς της λειτουργίας της είναι η προώθηση της γνώσης και χρήσης των πληροφορικών αγαθών από το κοινωνικό σύνολο και η εξάλειψη της τεχνοφοβίας και του "αναλφαριθμισμού" στην Πληροφορική.



<https://courses.epe.org.gr>

Σχετικά με τα ανοικτά μαθήματα της Ένωσης Πληροφορικών Ελλάδας:

- ✓ Τα μαθήματα πραγματοποιούνται εξ ολοκλήρου διαδικτυακά, ζωντανά μέσω της πλατφόρμας Zoom.
- ✓ Η συμμετοχή σε όλα τα μαθήματα είναι ελεύθερη για οποιονδήποτε από οπουδήποτε στην Ελλάδα ή στο εξωτερικό.
- ✓ Δεν υπάρχει οικονομικό κόστος ή άλλες προϋποθέσεις συμμετοχής.
- ✓ Οι Εισηγητές είναι μέλη της Ένωσης Πληροφορικών Ελλάδας και πραγματοποιούν τα μαθήματα εθελοντικά.
- ✓ Τα μαθήματα μαγνητοσκοπούνται και παραμένουν διαθέσιμα για σύγχρονη παρακολούθηση στο Αρχείο Μαθημάτων.
- ✓ Η εκπαίδευση που παρέχεται μέσω των ανοικτών διαδικτυακών μαθημάτων είναι άτυπη και δεν παρέχονται βεβαιώσεις παρακολούθησης στους συμμετέχοντες.



Ερωτήσεις



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<https://methodd.substack.com>