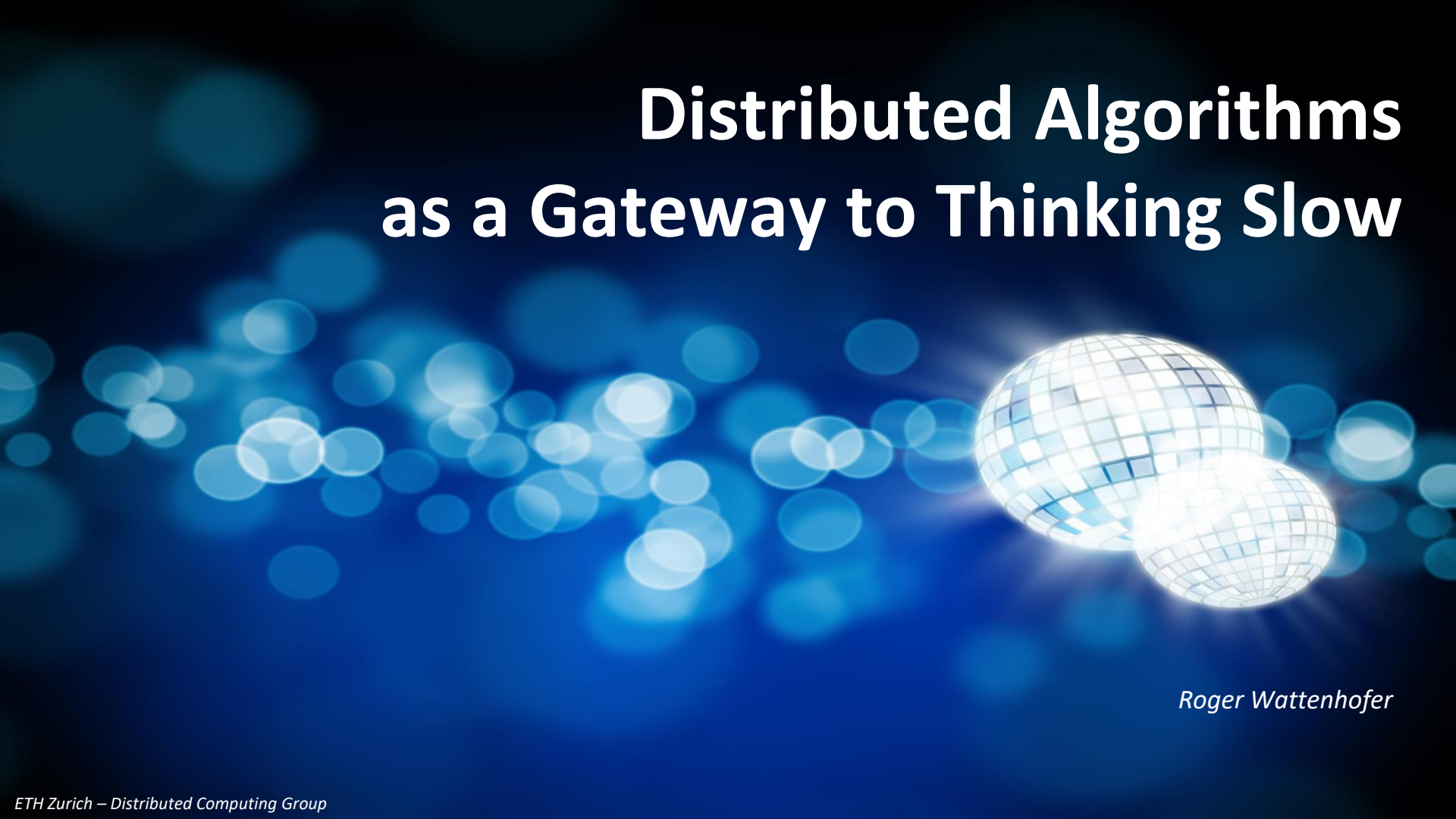
Distributed Algorithms as a Gateway to Thinking Slow



Roger Wattenhofer

ETH Zurich – Distributed Computing Group


ML Party

Deep Learning is Robust to Massive Label Noise

David Rolnick [*1] Andreas Veit [*2] Serge Belongie [2] Nir Shavit [3]

Machine Learning with Adversaries: Byzantine Tolerant Gradient Descent

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QSGD: Communication-Efficient SGD via Gradient Quantization and Encoding

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Byzantine Fault-Tolerant Distributed Machine Learning using D-SGD and Norm-Based Comparative Gradient Elimination (CGE)

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Nitin Vaidya
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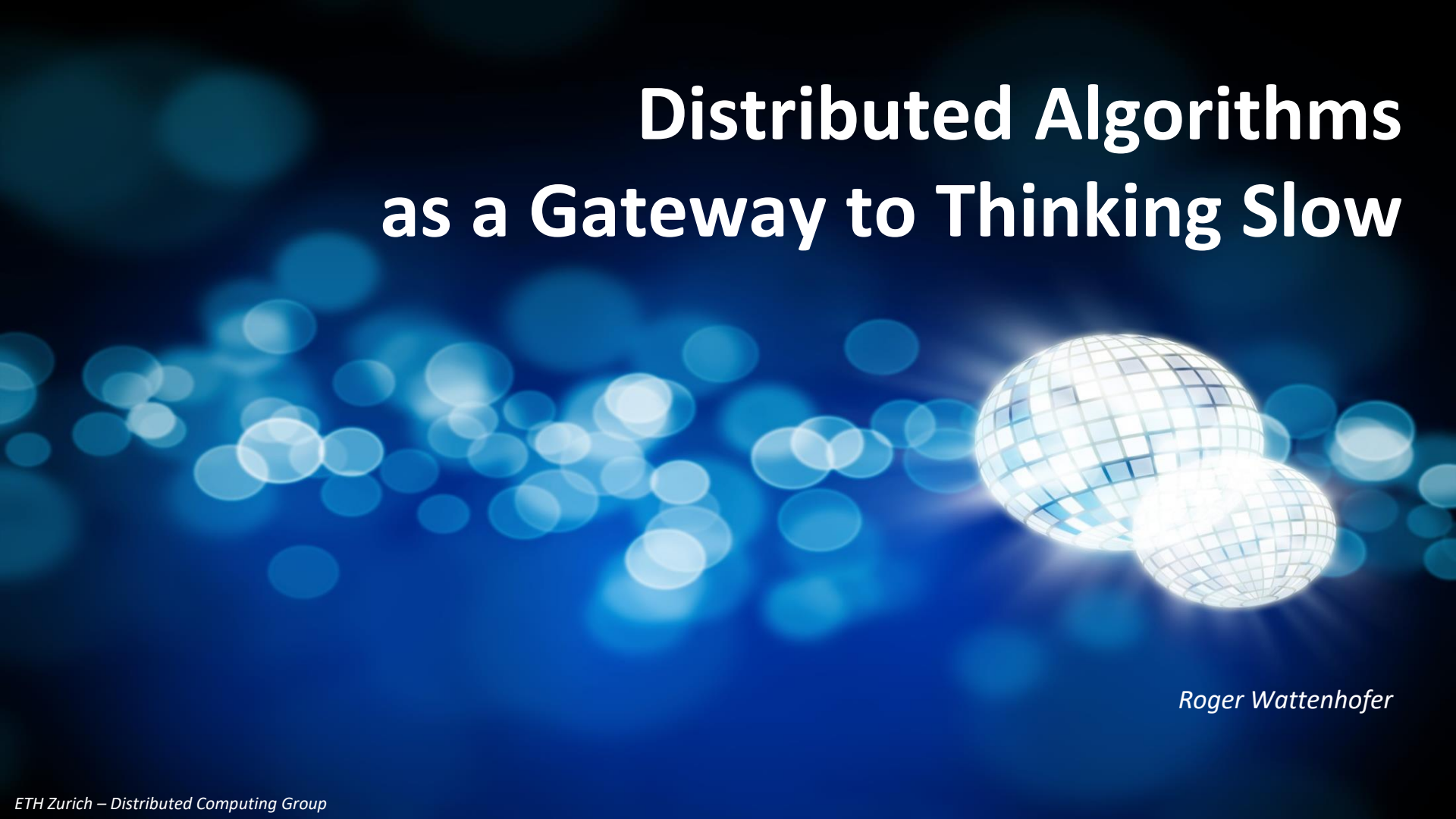
Distributed Algorithms as a Gateway to Thinking Slow



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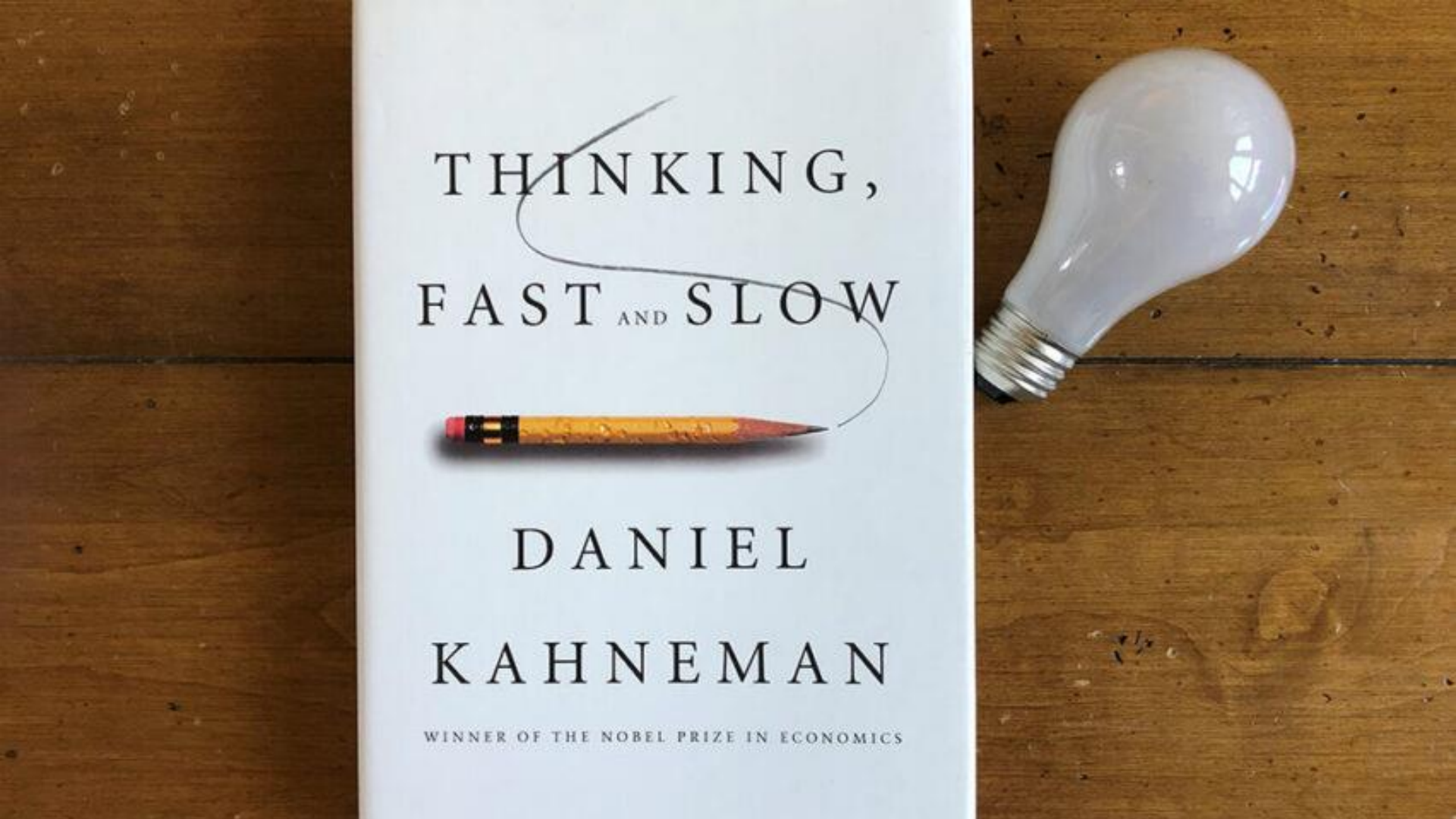
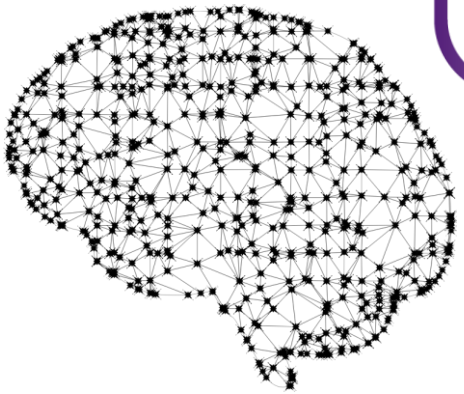
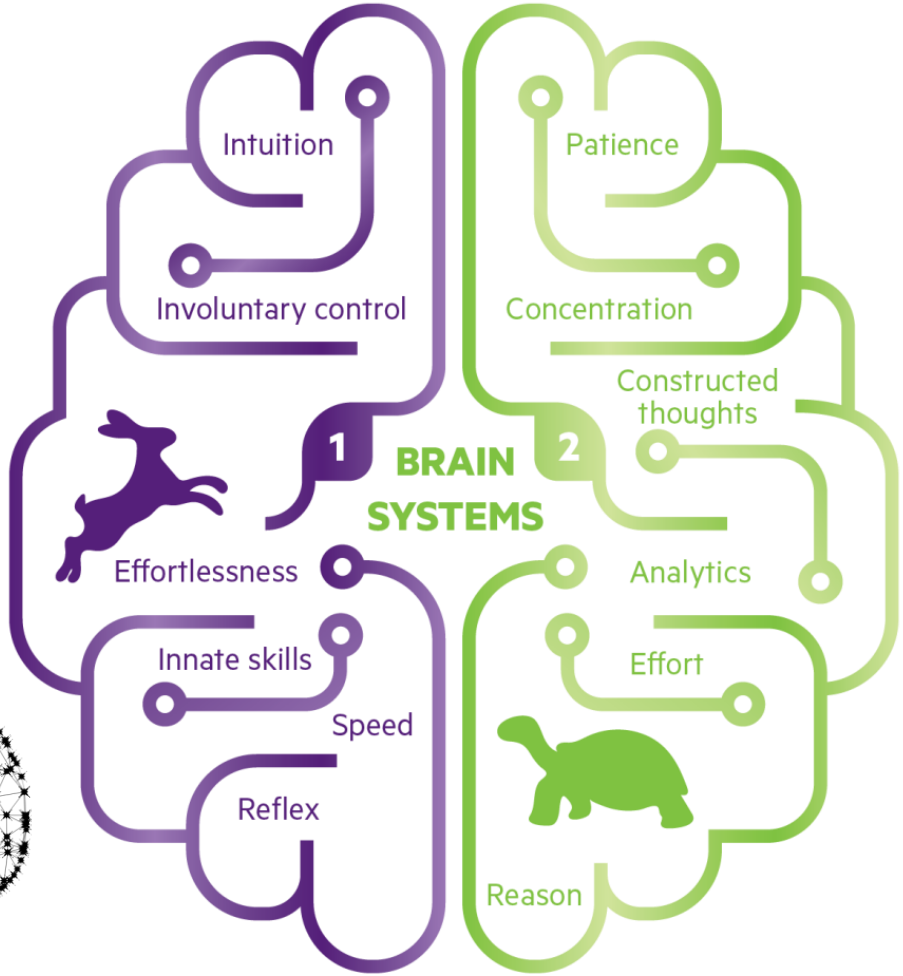

THINKING,
FAST AND SLOW
DANIEL
KAHNEMAN
WINNER OF THE NOBEL PRIZE IN ECONOMICS

Machine Learning



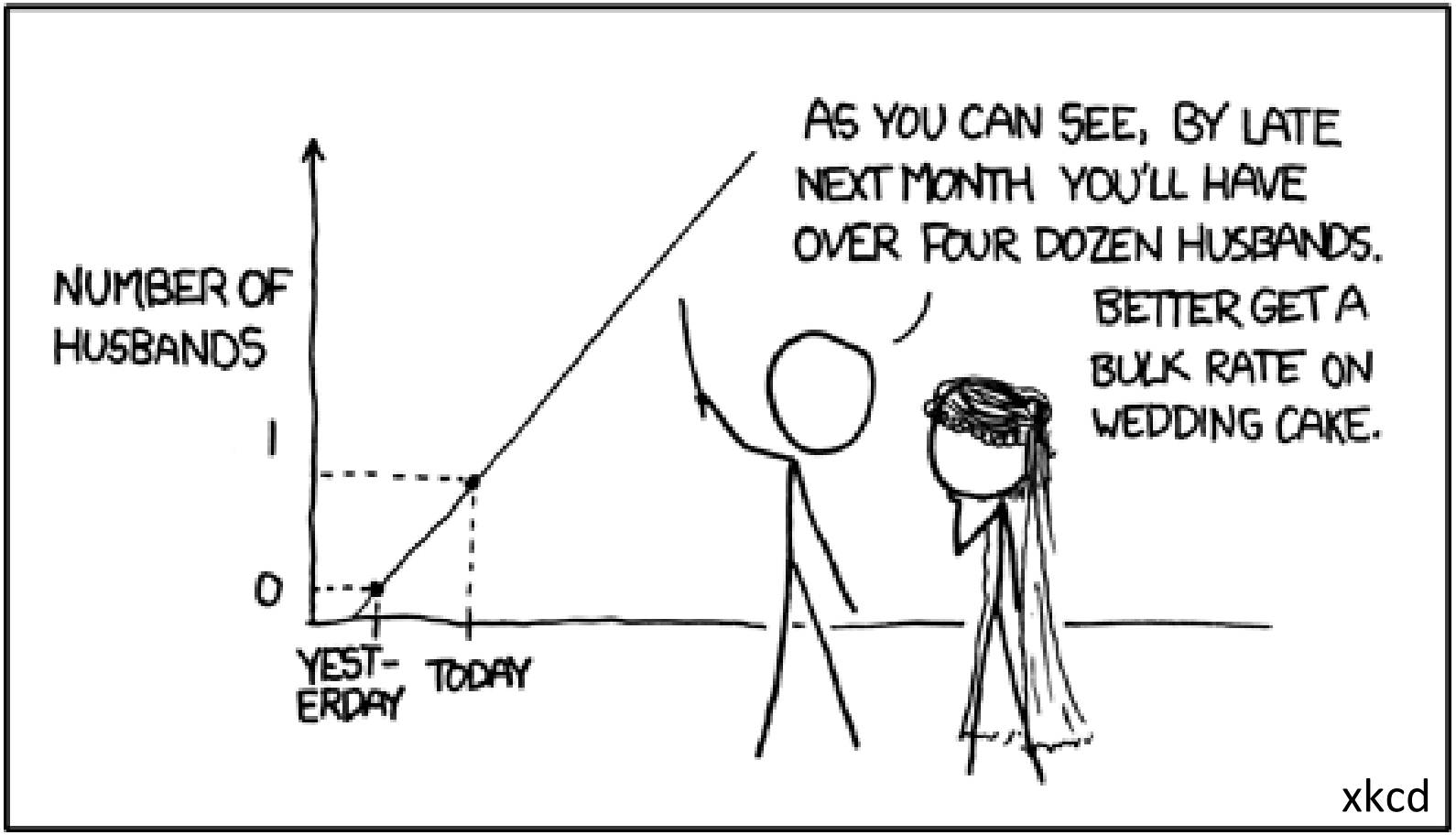


?

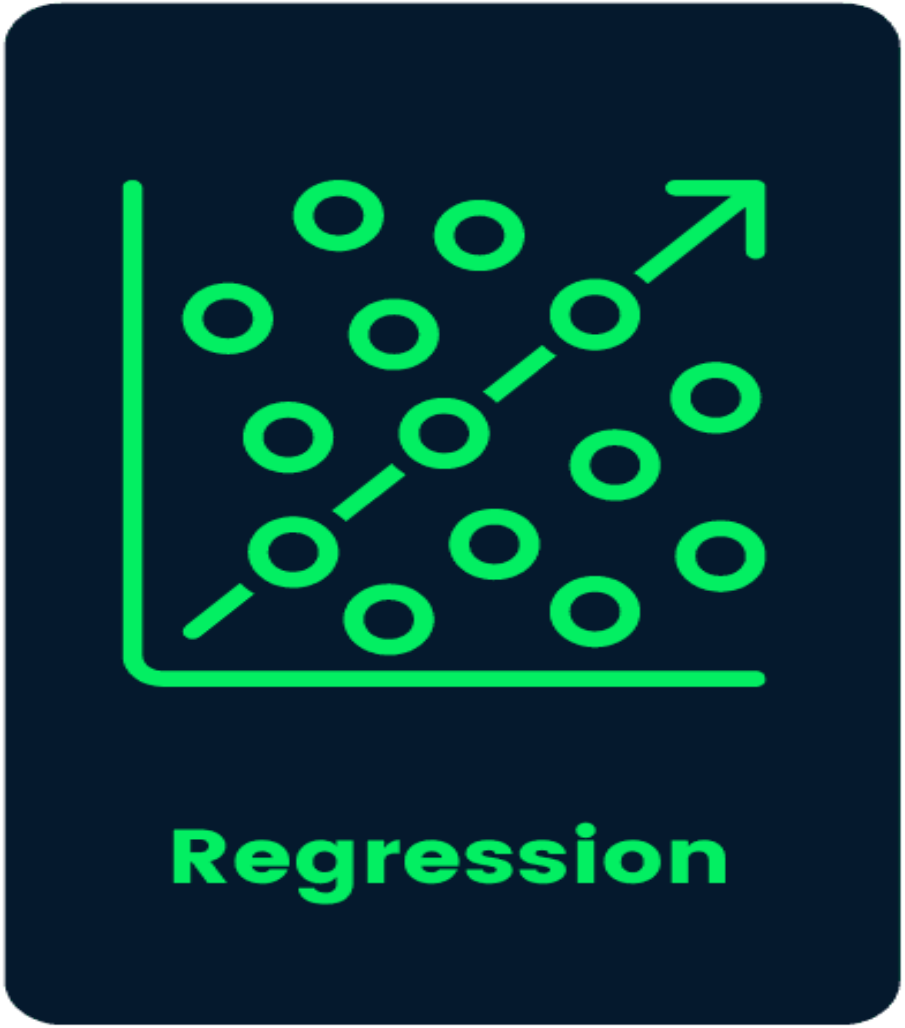
Classification

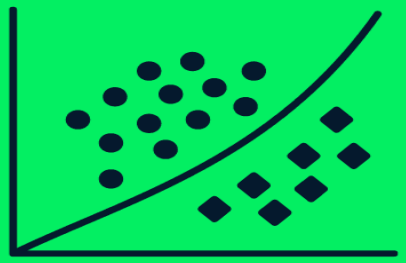
Chihuahua
vs.
Muffin

Linear Regression

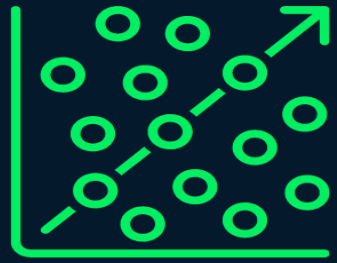



Classification

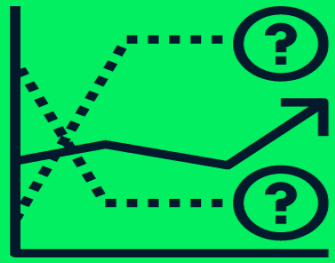

Regression



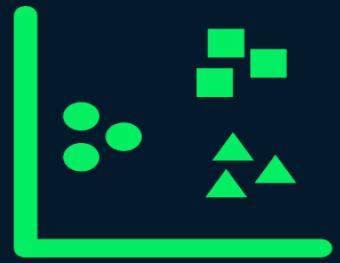
Classification



Regression



Forecast

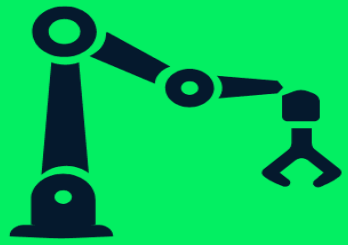


Clustering

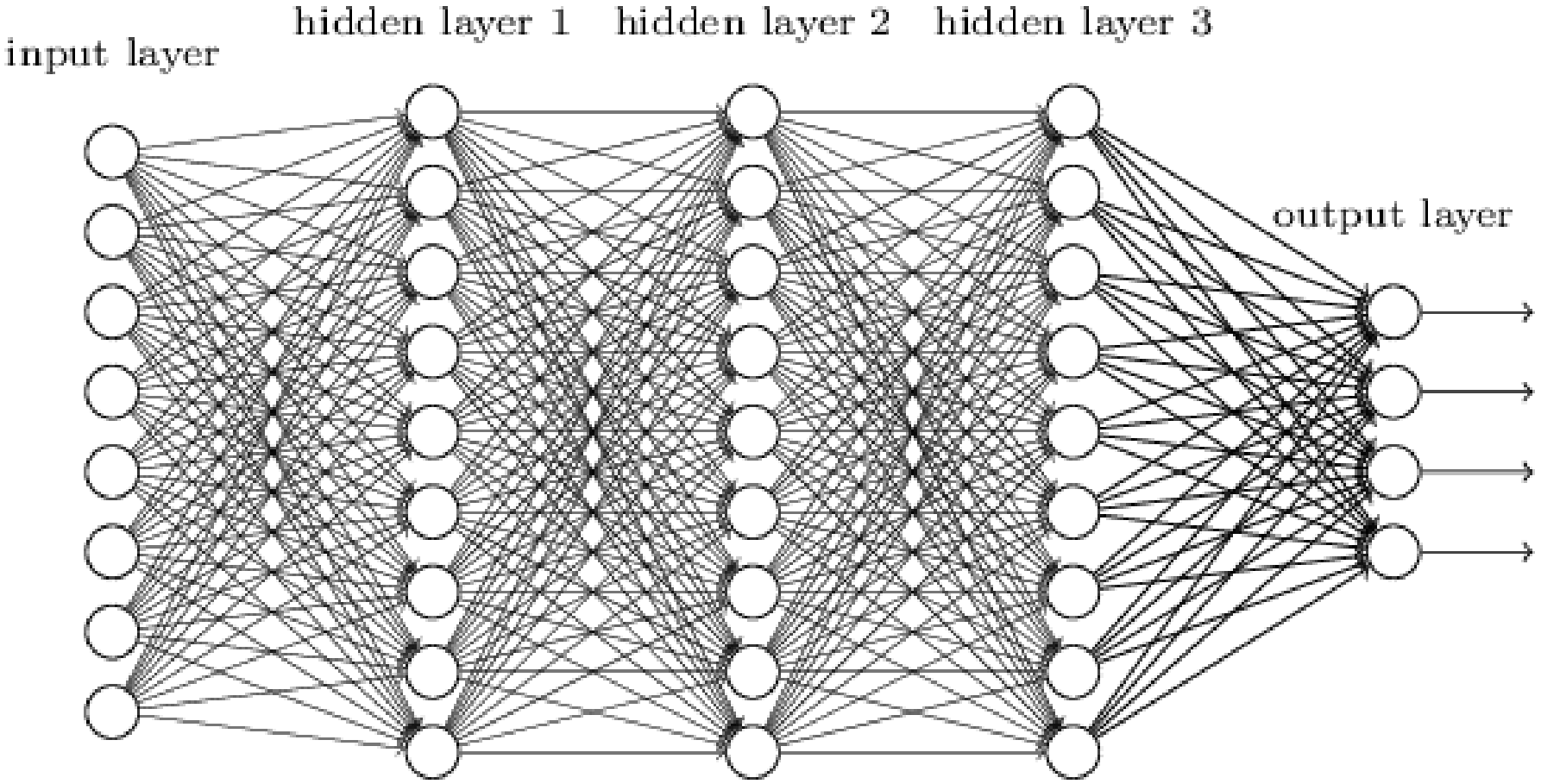
Machine Translation

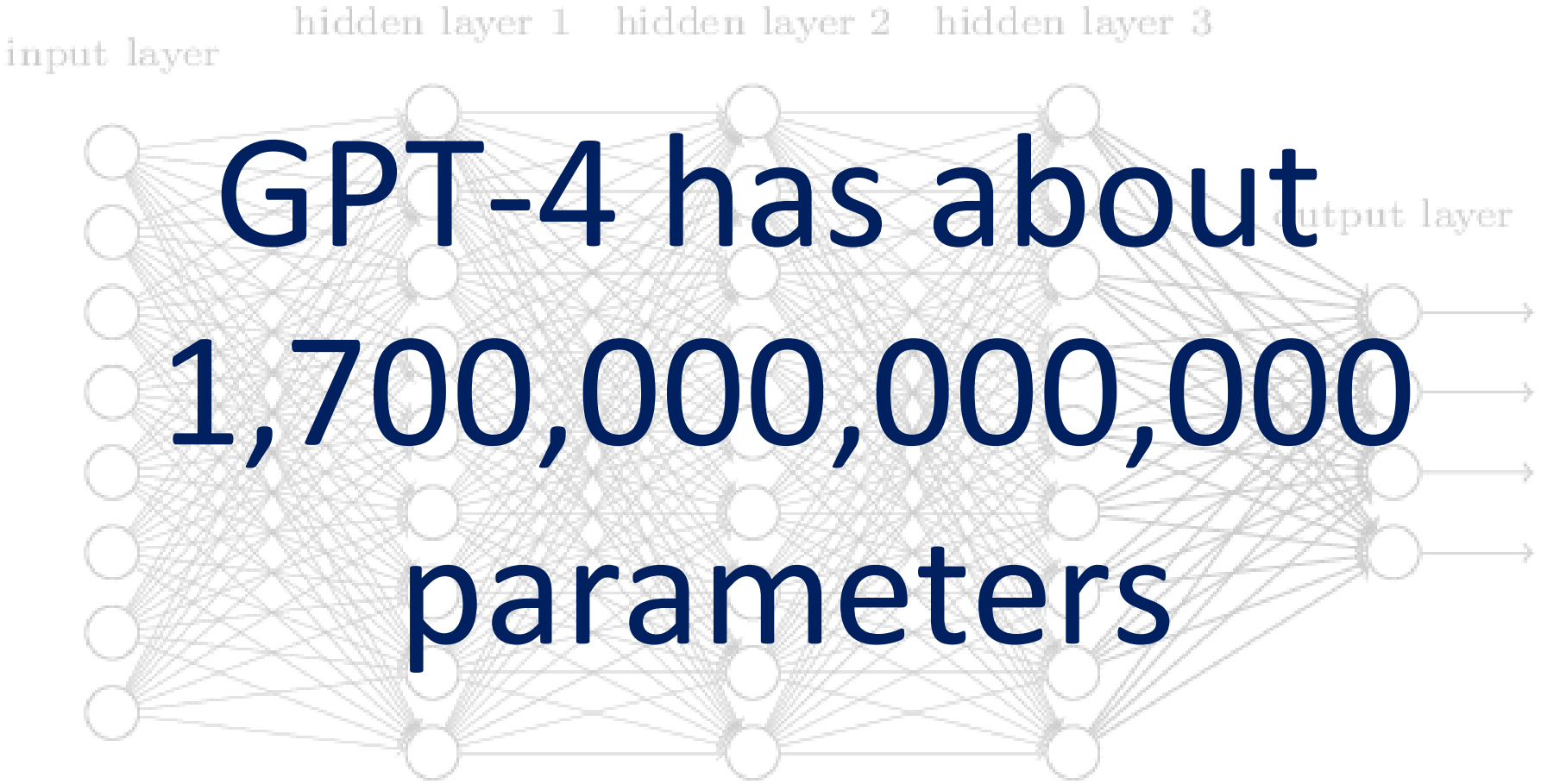
Computer Vision

Generative Art



Reinforcement learning

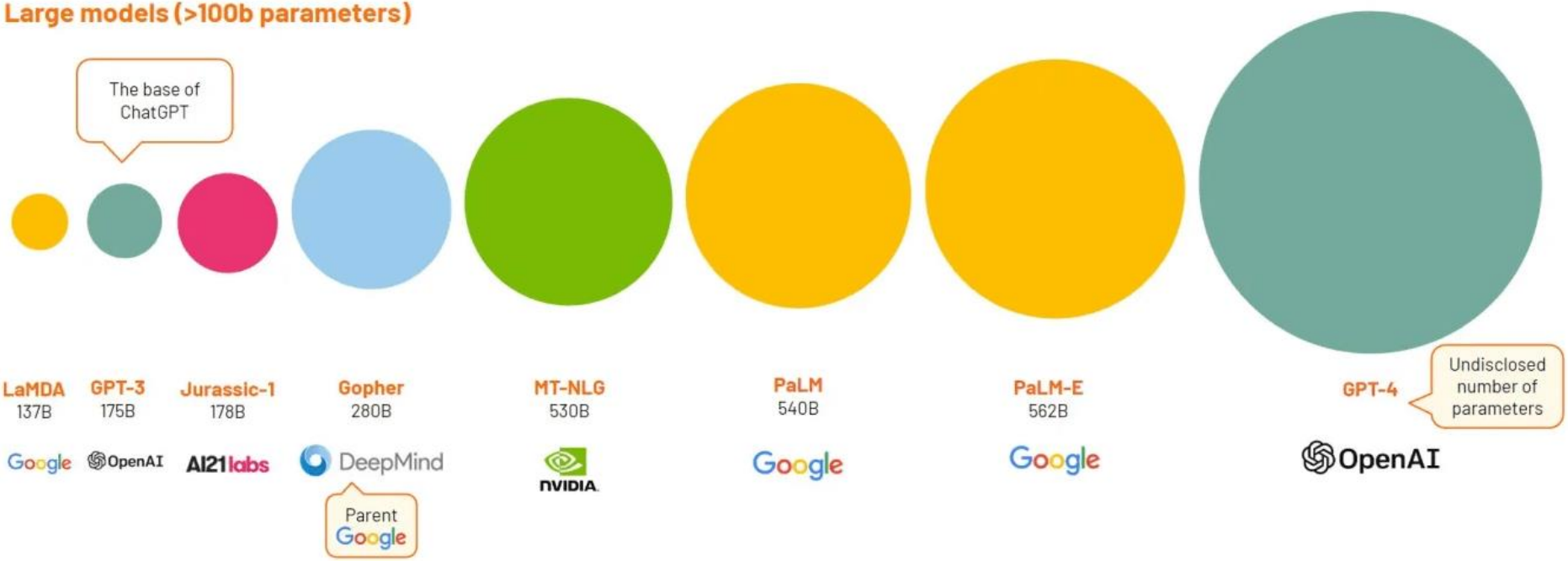

input layer
hidden layer 1
hidden layer 2
hidden layer 3
output layer

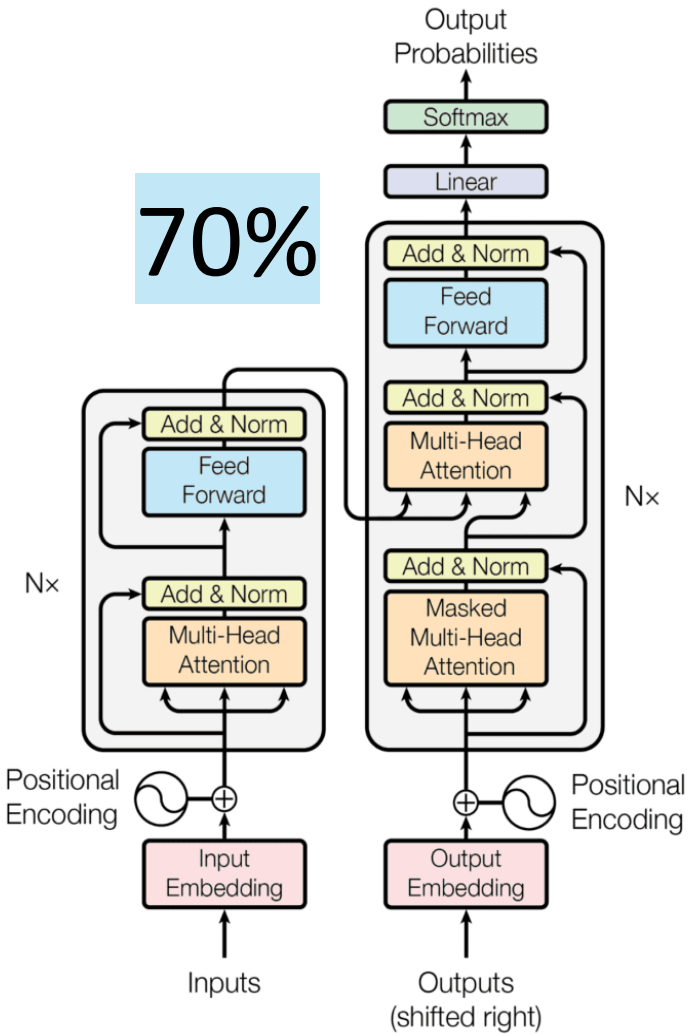


GPT-4 has about 1,700,000,000,000 parameters

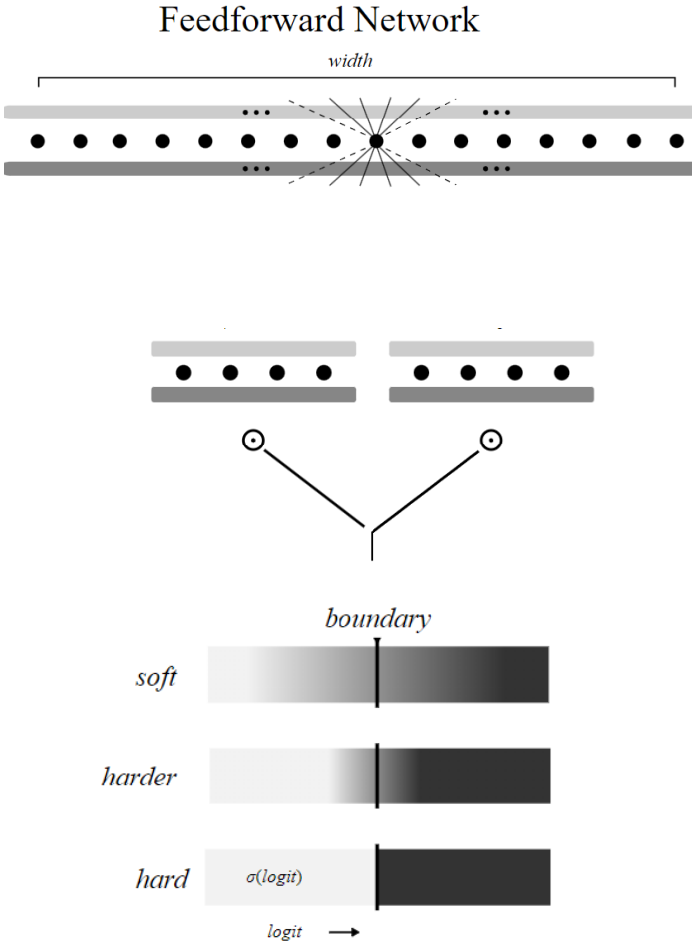
Small models (<= 100b parameters)

Large models (>100b parameters)




Output Probabilities
Softmax
Linear
Add & Norm
Feed Forward
Add & Norm
Multi-Head Attention
Add & Norm
Masked Multi-Head Attention
Add & Norm
Feed Forward
Add & Norm
Multi-Head Attention
Nx
Nx
Positional Encoding
Positional Encoding
Input Embedding
Output Embedding
Inputs
Outputs (shifted right)

70%


Feedforward Network
width
boundary
soft
harder
hard
σ(logit)
logit

Exponentially Faster Language Modeling

Peter Belcak and Roger Wattenhofer
ETH Zürich
{belcak,wattenhofer}@ethz.ch

Model	N_T	N_I/N_T	RTE	MRPC	STSB	SST-2	MNLI	QNLI	QQP	Avg	CoLA	Avg
UltraFastBERT-1x11-long	4095	0.3%	60.7	87.5	86.4	89.9	81.3	89.7	87.6	83.0	35.1	77.7
External Baselines												
OpenAI GPT	3072	100%	56.0	82.3	80.0	91.3	81.4	87.4	70.3	78.8	45.4	75.1
DistilBERT	3072	100%	59.9	87.5	86.9	91.3	82.2	89.2	71.3	81.2	52.1	77.6
BERT-base	3072	100%	66.4	88.9	85.8	93.5	83.4	90.5	71.2	83.0	51.3	79.6

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▲ Exponentially faster language modelling (arxiv.org)

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▲ WithinReason 10 days ago | next [–]

Link to previous paper:

▲ qntty 10 days ago | prev | next [–]

According to scientists, we only use 0.3% of our neural networks. Imagine if we could use 100%.

r/

Beiträge

Gepostet von u/lexected vor 11 Tagen

181

[R] Exponentially Faster Language Modelling

Research

TL;DR: Organize your neurons into a tree to get 78x faster inference (theoretical limit is 341x).

This was demonstrated on BERT-base, where this change preserved 96% of its downstream GLUE performance. For a quick comparison, DistilBERT offers 1.6x acceleration while preserving 97% of GLUE performance.

This is a HuggingFace Featured Paper from 11/21/2023.

Paper: https://arxiv.org/abs/2311.10770

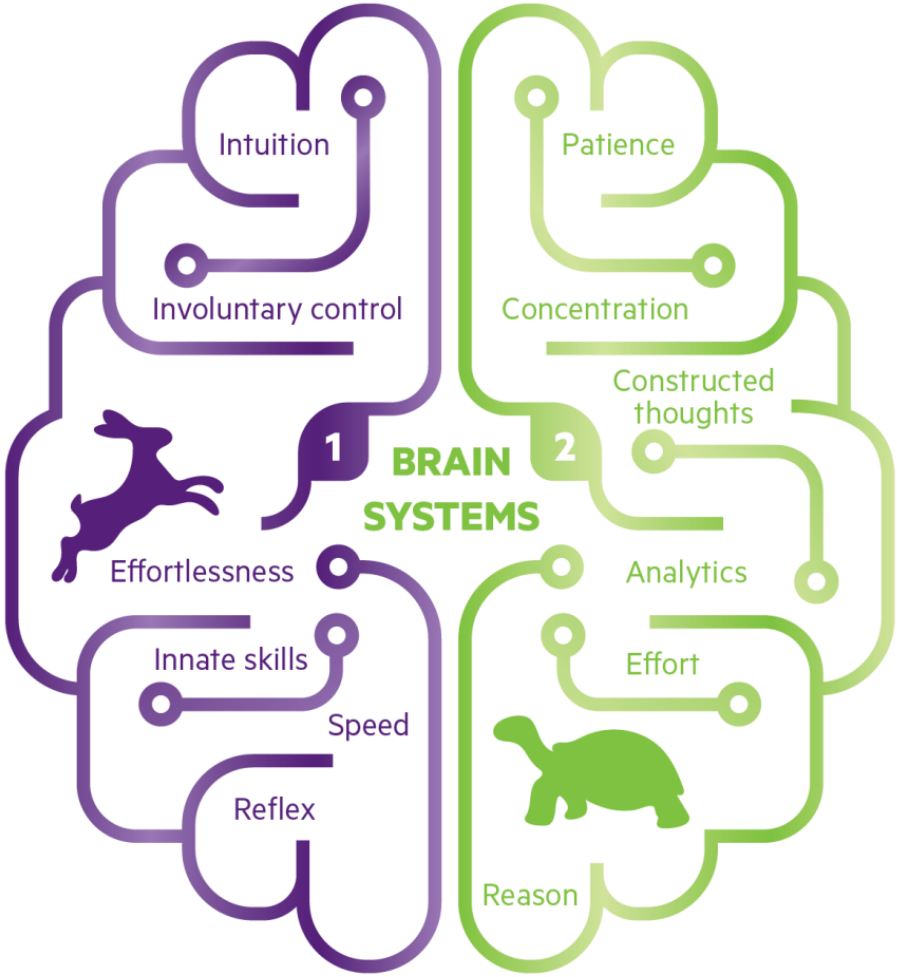
Code: https://github.com/pbelcak/UltraFastBERT

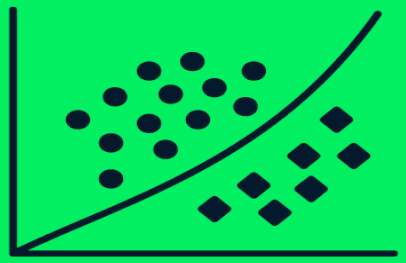
This study introduces the "Ultra Fast BERT" model, designed to theoretically enhance inference speed by up to 341 times. The increase in speed is possible by replacing the standard FeedForward layers within the Attention mechanism.

1/4

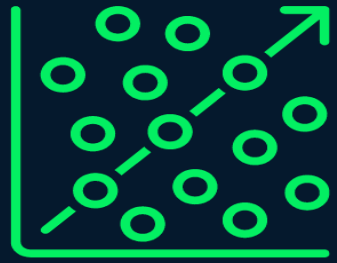
Exponentially Faster Language Modeling

Model	N_T	N_1/N_T	RTE	MRPC	STSB	SST-2	MNLI	QNLI	QQP	Avg	CoLA	Avg
Baselines												
crammedBERT-3072	4095	100.0%	58.8	87.6	85.2	91.9	82.8	90.4	89.0	83.6	45.0	79.3
crammedBERT-4095	3072	100.0%	57.6	89.1	85.9	91.9	81.3	90.9	87.6	83.2	47.9	79.3
UltraFastBERTs												

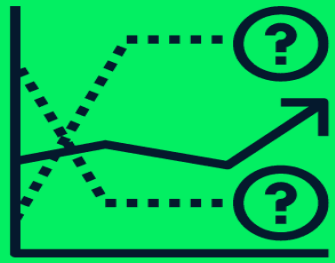

Intuition
Involuntary control
Patience
Concentration
Constructed thoughts
1
BRAIN SYSTEMS
2
Effortlessness
Analytics
Innate skills
Speed
Reflex
Effort
Reason



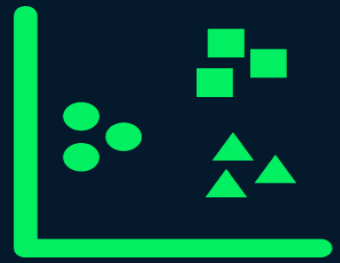
Classification



Regression



Forecast

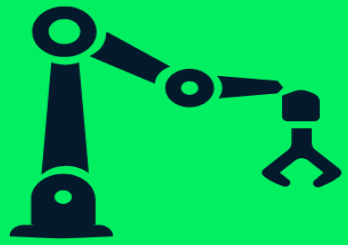


Clustering

Machine Translation

Computer Vision

Generative Art



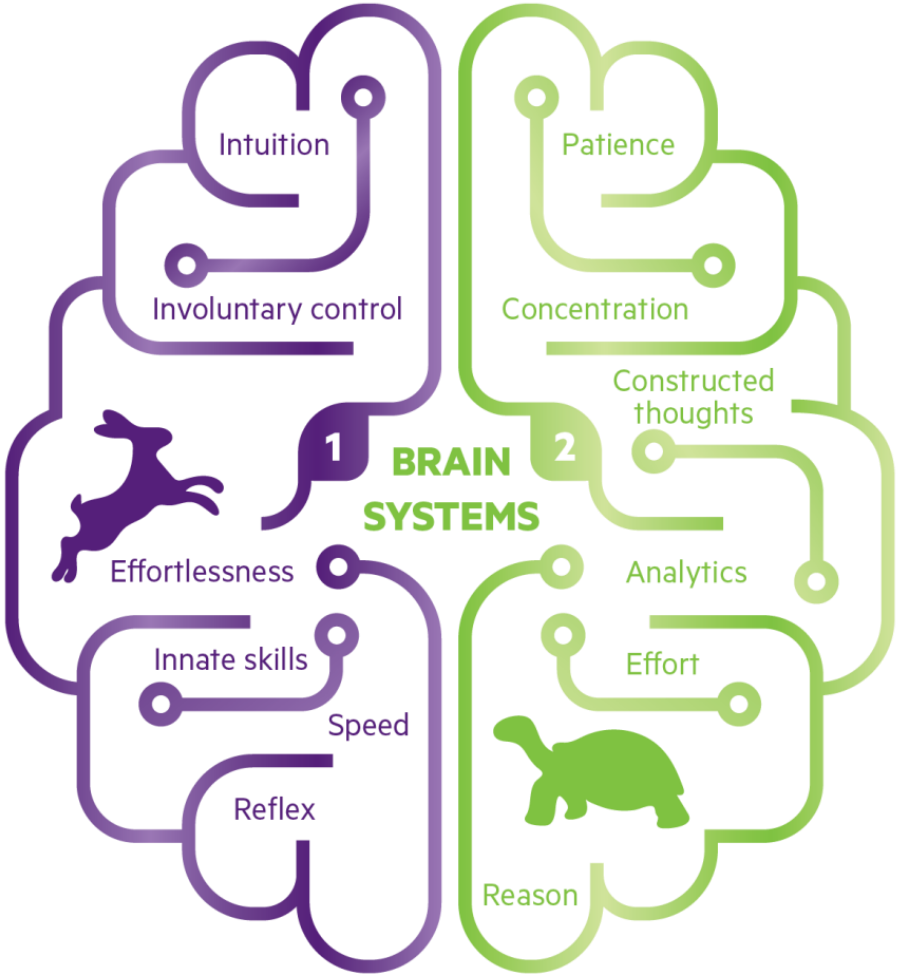
Reinforcement learning

uter
on

Generative Art

Reinfor
lear

Refik Anadol, Unsupervised, MOMA


Intuition
Involuntary control
Patience
Concentration
Constructed thoughts
1
BRAIN SYSTEMS
2
Effortlessness
Analytics
Innate skills
Speed
Reflex
Effort
Reason

User

what is 5 times 7?

Assistant

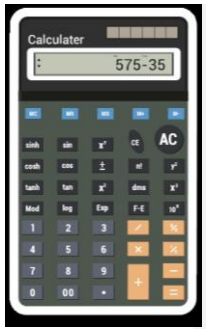
5 times 7 equals 35.

User

what is 384903720 times 538982734?

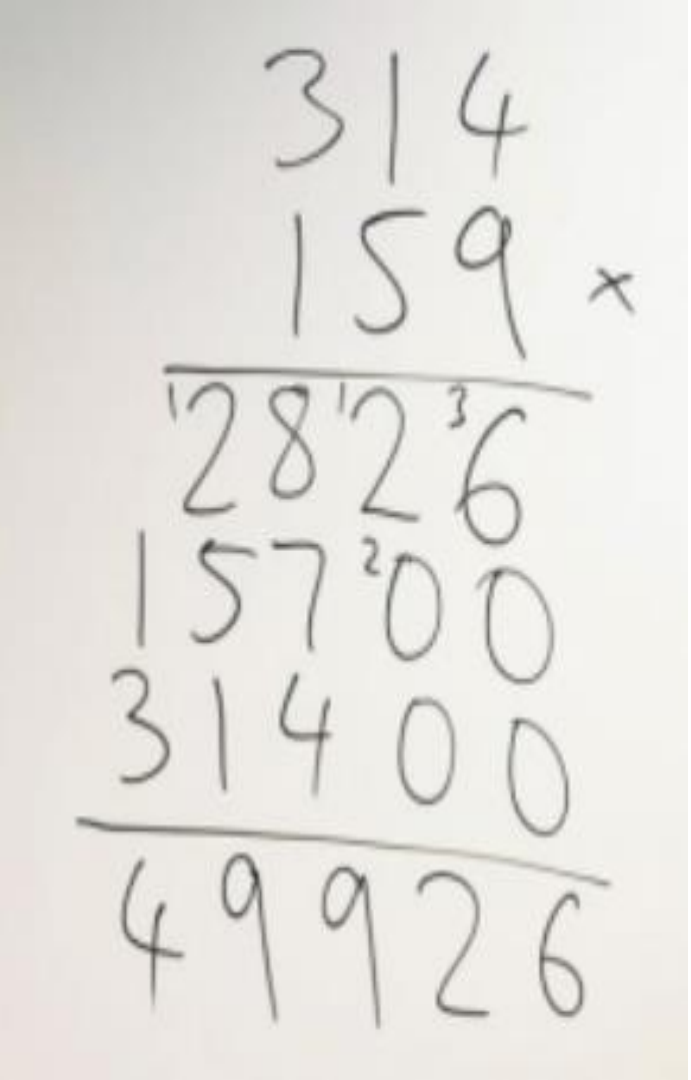
Assistant

384,903,720 times 538,982,734 equals 207,559,081,508,668,480.



384903720 × 538982734 =

207,456,459,332,370,480



User

what is 384903720 times 538982734?

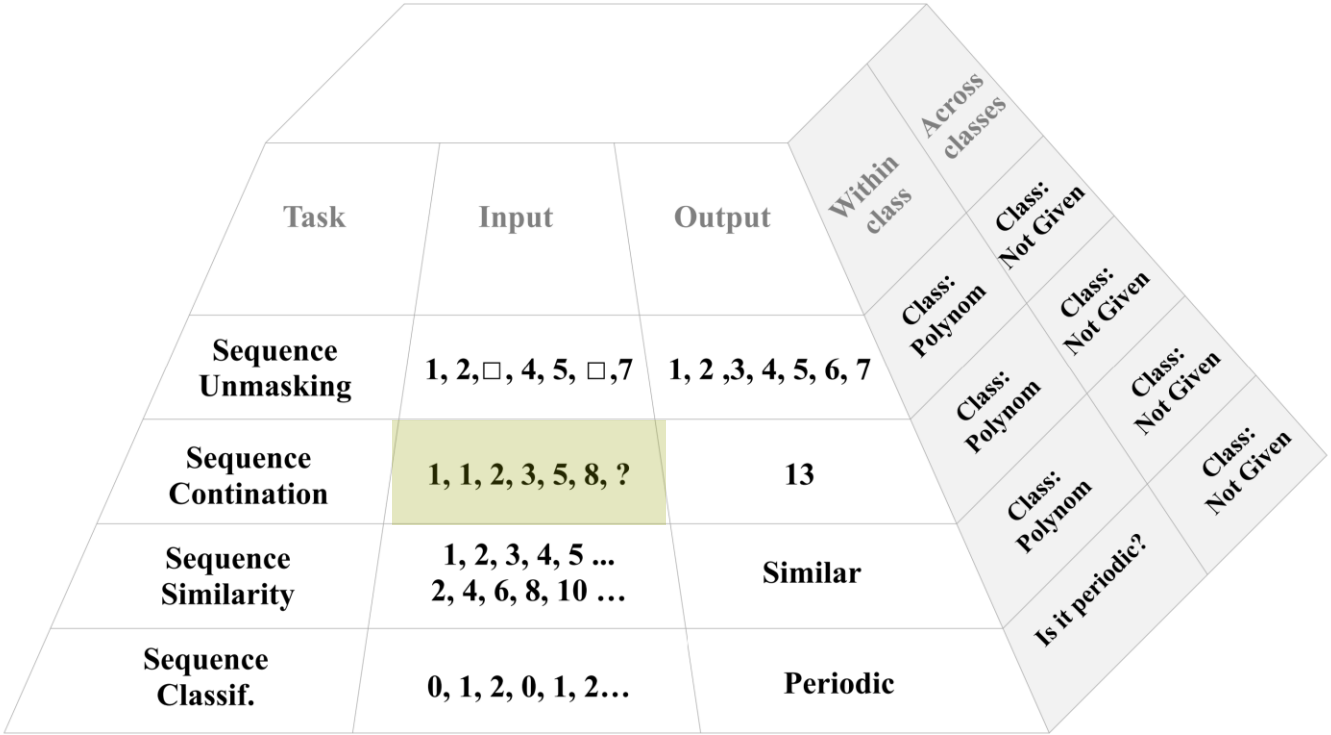
Assistant

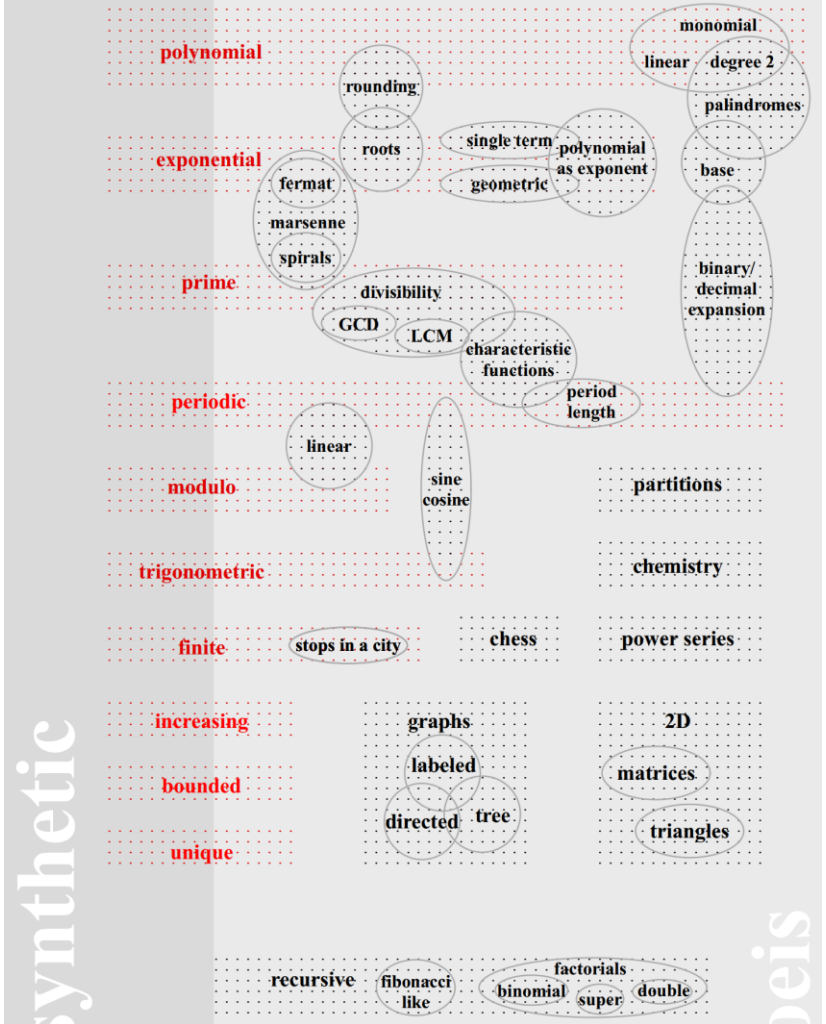
384,903,720 times 538,982,734 equals 207,559,081,508,668,480.

Thinking Slow Benchmark?

FACT: Learning Governing Abstractions Behind Integer Sequences

Peter Belcák, Ard Kastrati, Flavio Schenker, Roger Wattenhofer




synthetic
oeis
polynomial
exponential
prime
periodic
modulo
trigonometric
finite
increasing
bounded
unique
rounding
roots
fermat
marsenne
spirals
divisibility
GCD
LCM
linear
sine
cosine
stops in a city
graphs
labeled
directed
tree
recursive
fibonacci like
single term
geometric
polynomial as exponent
factorials
binomial
super
double
monomial
linear
degree 2
palindromes
base
binary/ decimal expansion
characteristic functions
period length
partitions
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triangles


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ひらめきパズル
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謎解き
with
クロスワード
4
漢字
パズル
コレクション
1
絵むす
コレクション

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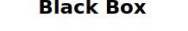
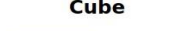

Black Box


Bridges


Cube

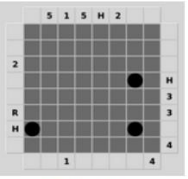

Dominosa

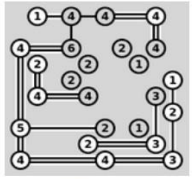

Fifteen

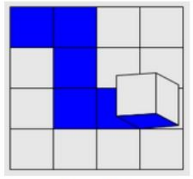

Filling


Flip

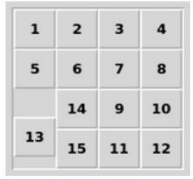

Flood

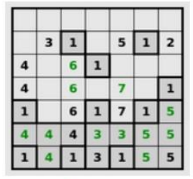

Galaxies

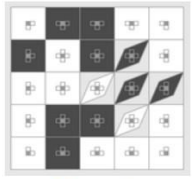

Guess

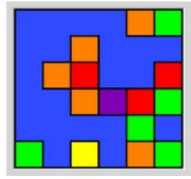

Inertia

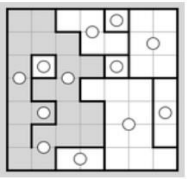

Keen

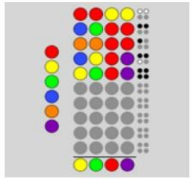

Light Up

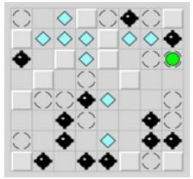

Loopy


Magnets

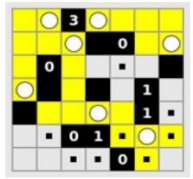

Map

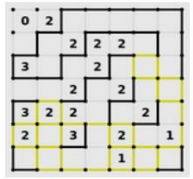

Mines

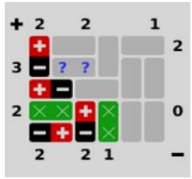

Mosaic

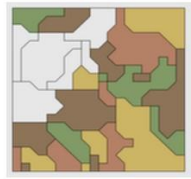

Net


Netslide

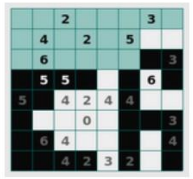

Palisade

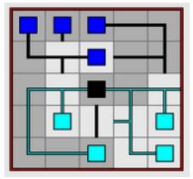

Pattern

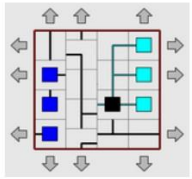

Pearl


Pegs


Range


Rectangles


Same Game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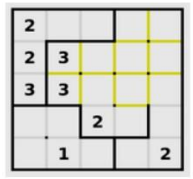
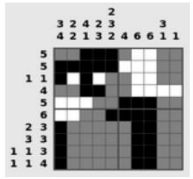

Singles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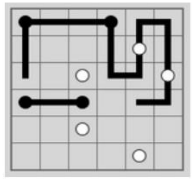
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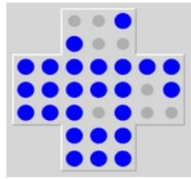
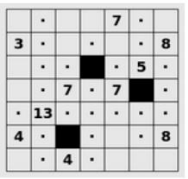

Solo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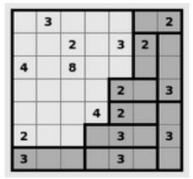
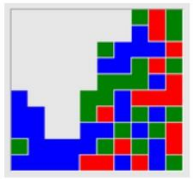
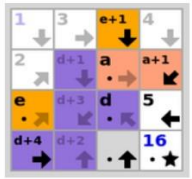
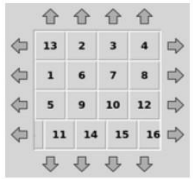
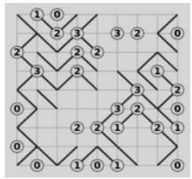

Towers


Tracks


Twiddle

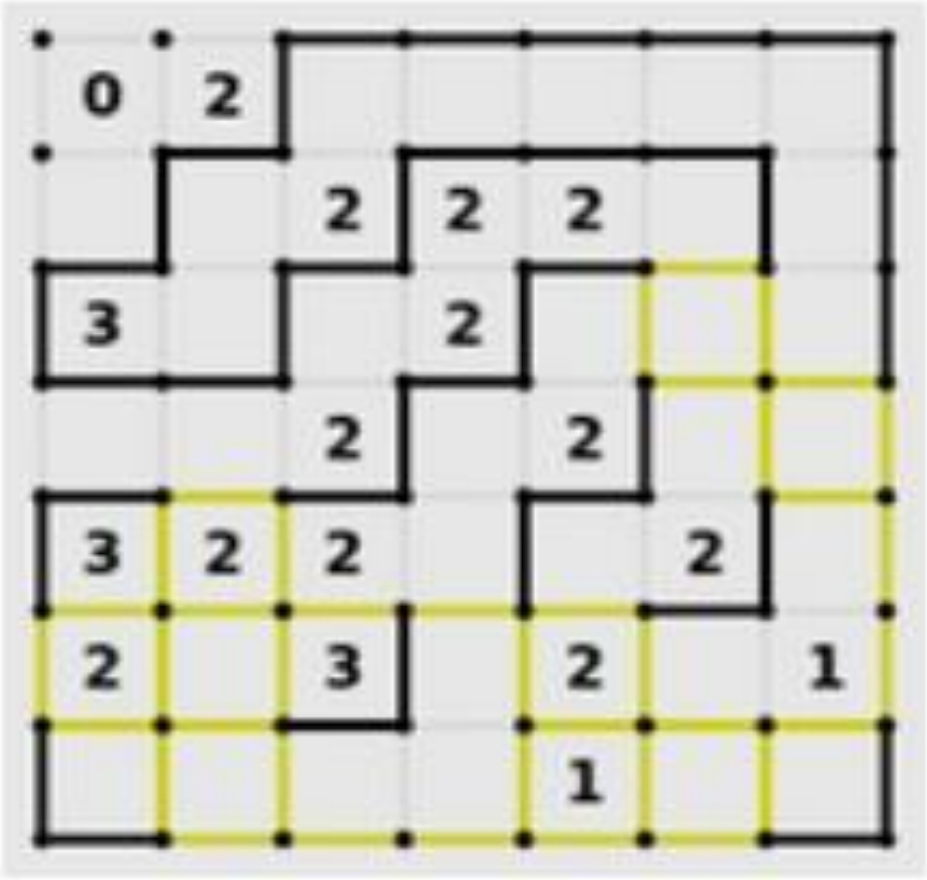

Undead


Unequal

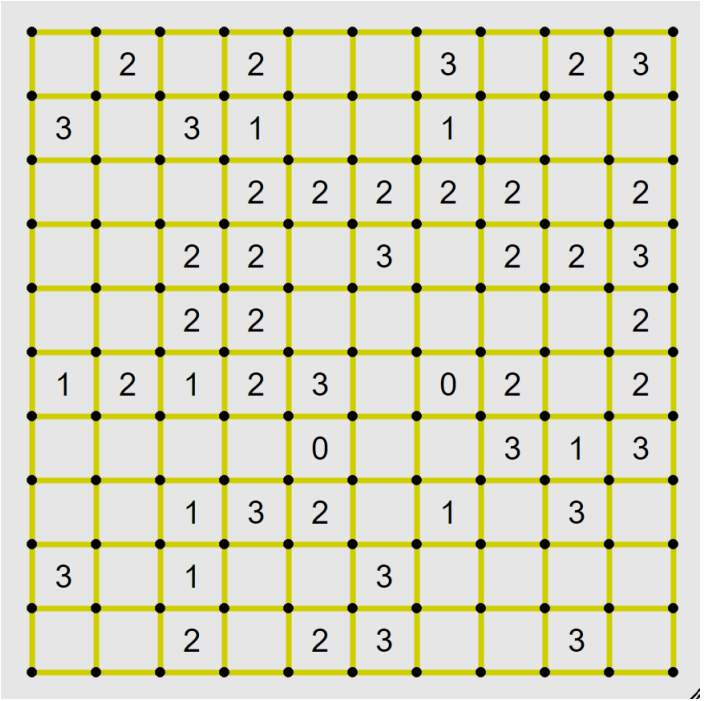

Unruly

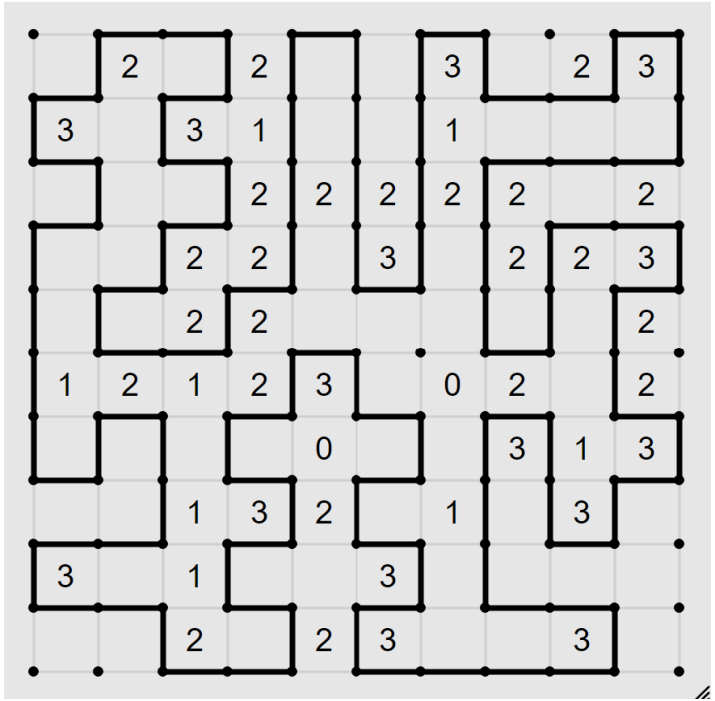

Untangle

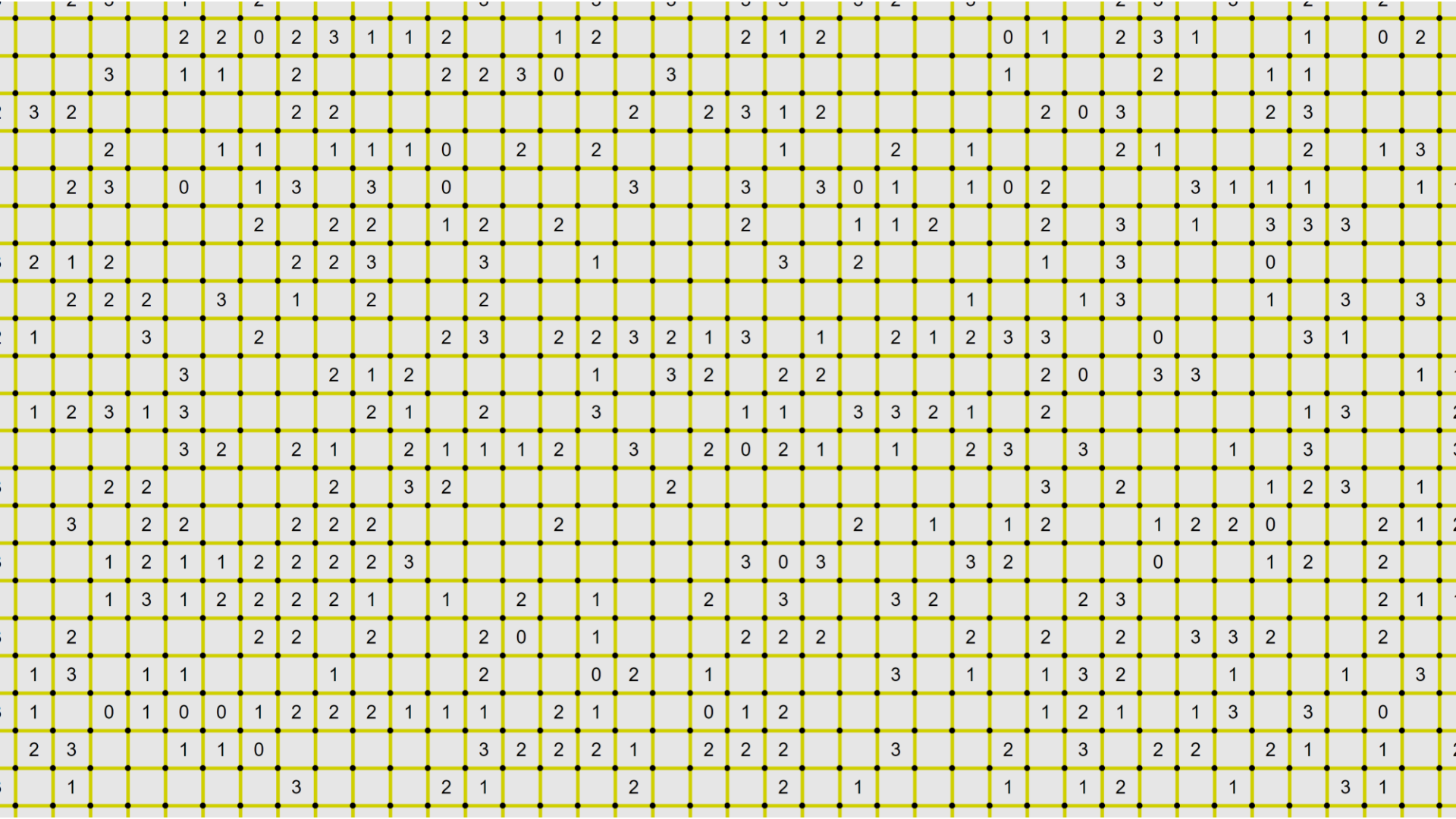
Loopy

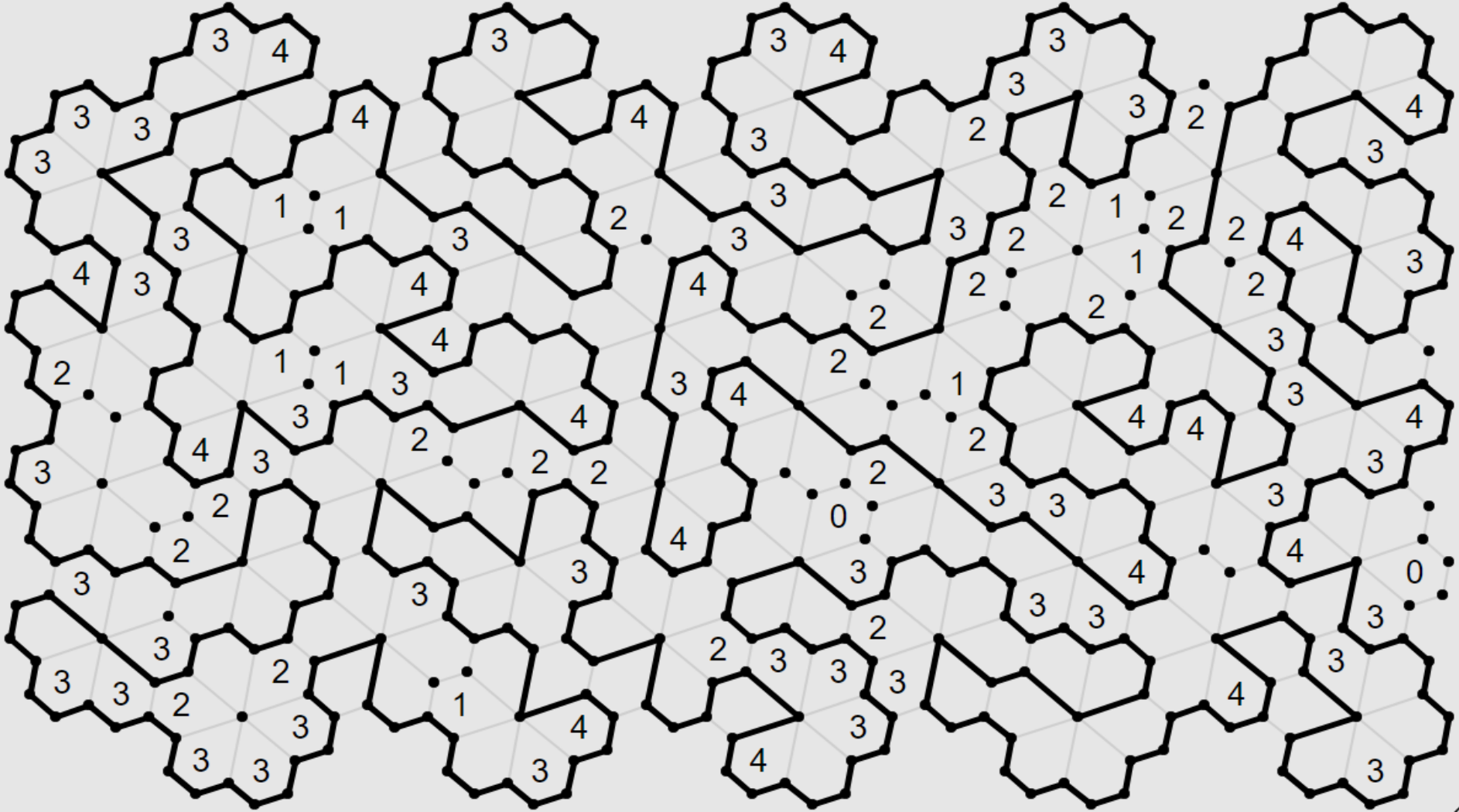


Loopy (Takegaki, Slitherlink, Ouroboros, Suriza, …)

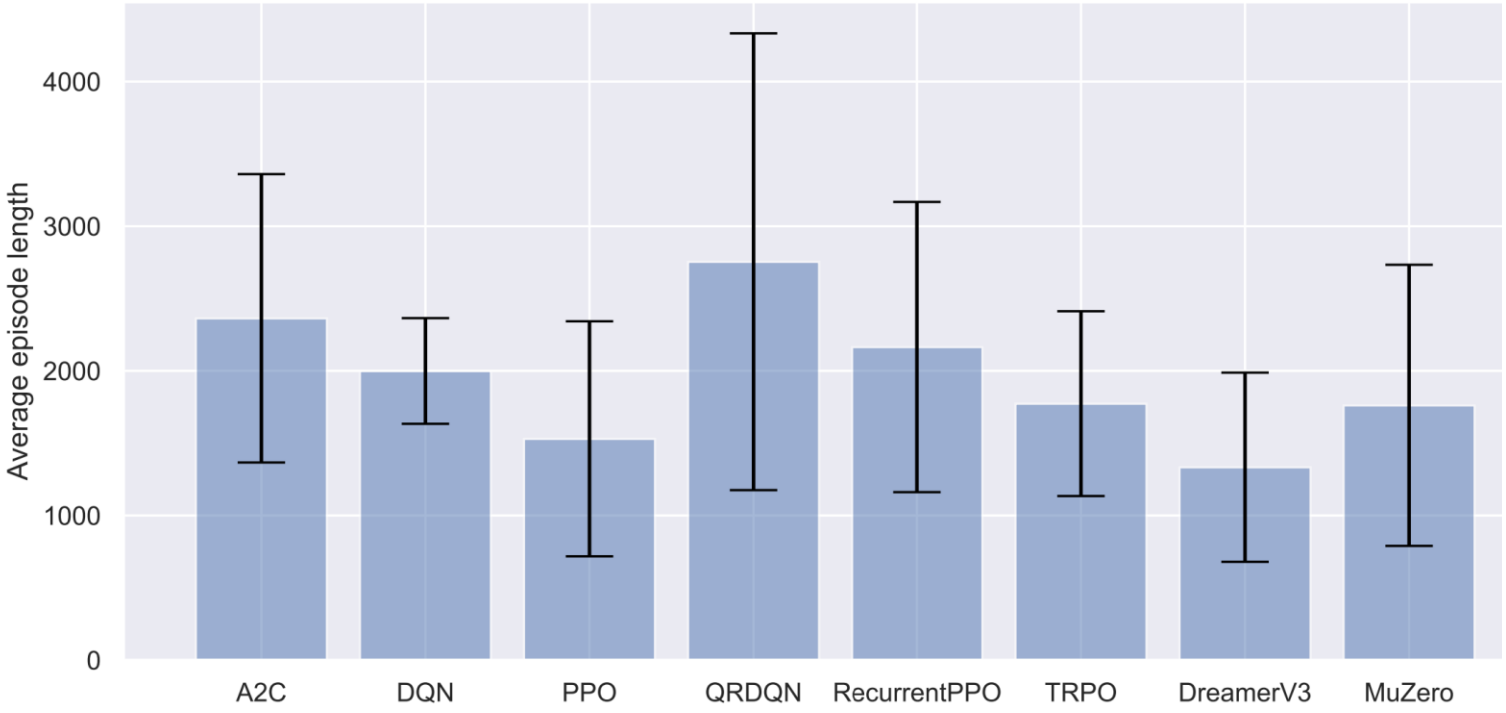




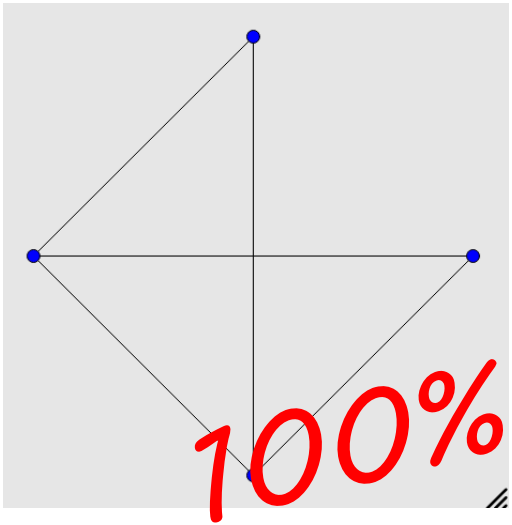


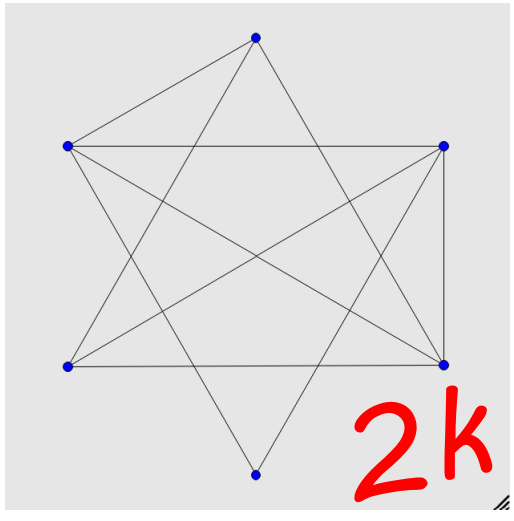


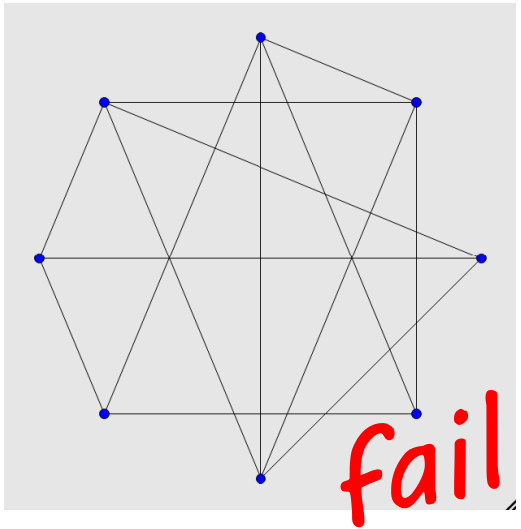
RLP: A Reinforcement Learning Benchmark for Neural Algorithmic Reasoning



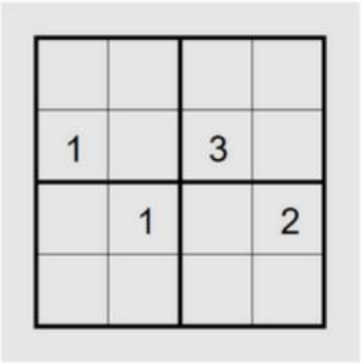
Puzzle	Parameters	PPO	DreamerV3
Netslide	`2x3b1`	35.3 ± 0.7 (100.0%)	12.0 ± 0.4 (100.0%)
	`3x3b1`	4742.1 ± 2960.1 (9.2%)	3586.5 ± 676.9 (22.4%)
Same Game	`2x3c3s2`	11.5 ± 0.1 (100.0%)	7.3 ± 0.2 (100.0%)
	`5x5c3s2`	1009.3 ± 1089.4 (30.5%)	527.0 ± 162.0 (30.2%)
Untangle	`4`	34.9 ± 10.8 (100.0%)	6.3 ± 0.4 (100.0%)
	`6`	2294.7 ± 2121.2 (96.2%)	1683.3 ± 73.7 (82.0%)



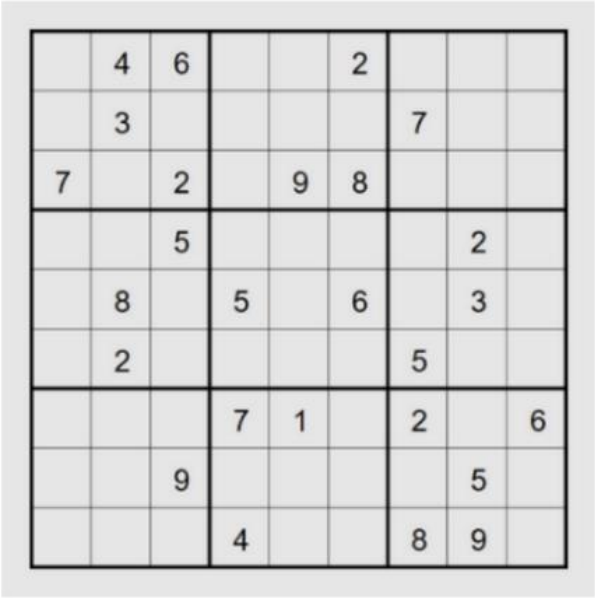




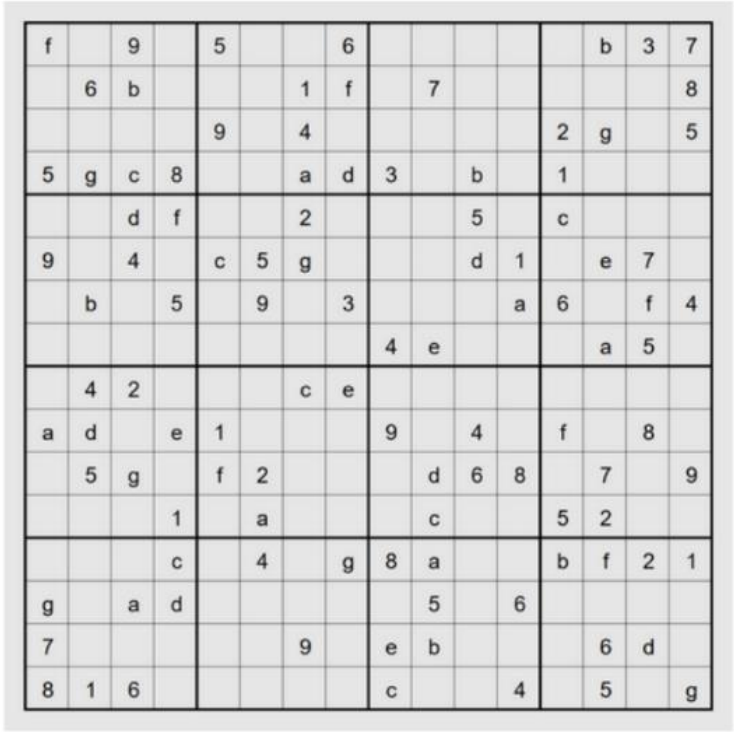
Sudoku



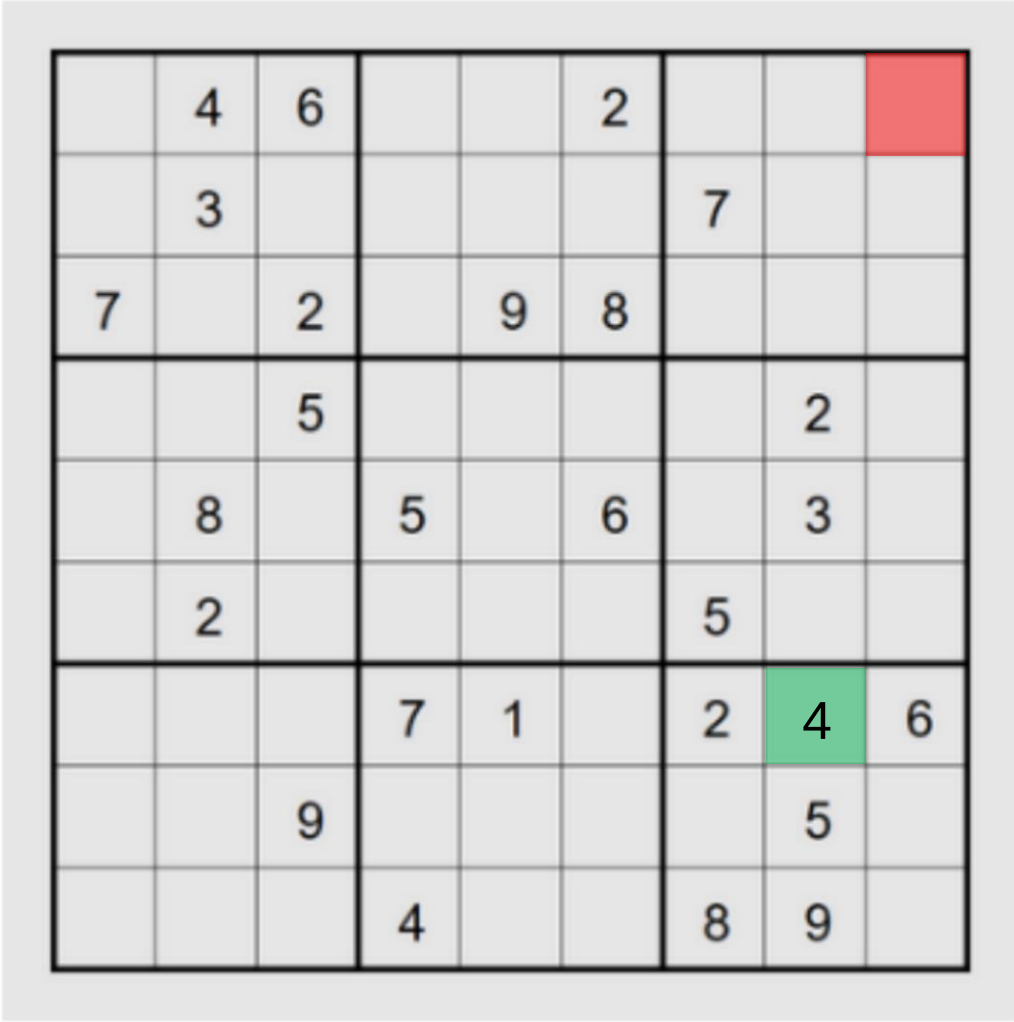
1		3	
	1		2

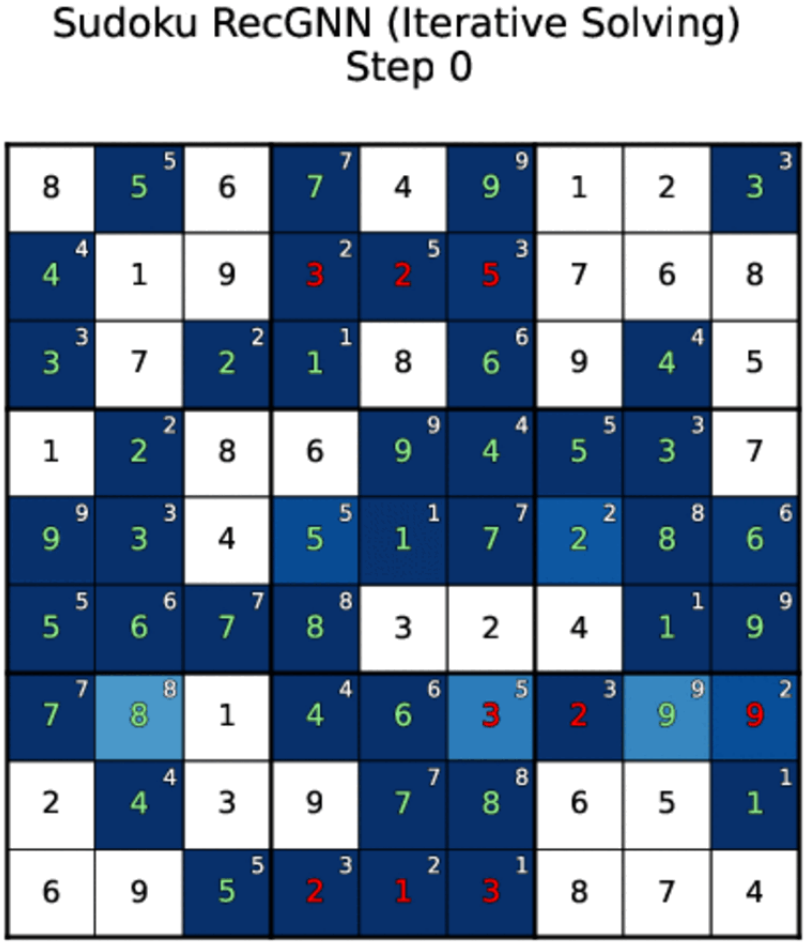


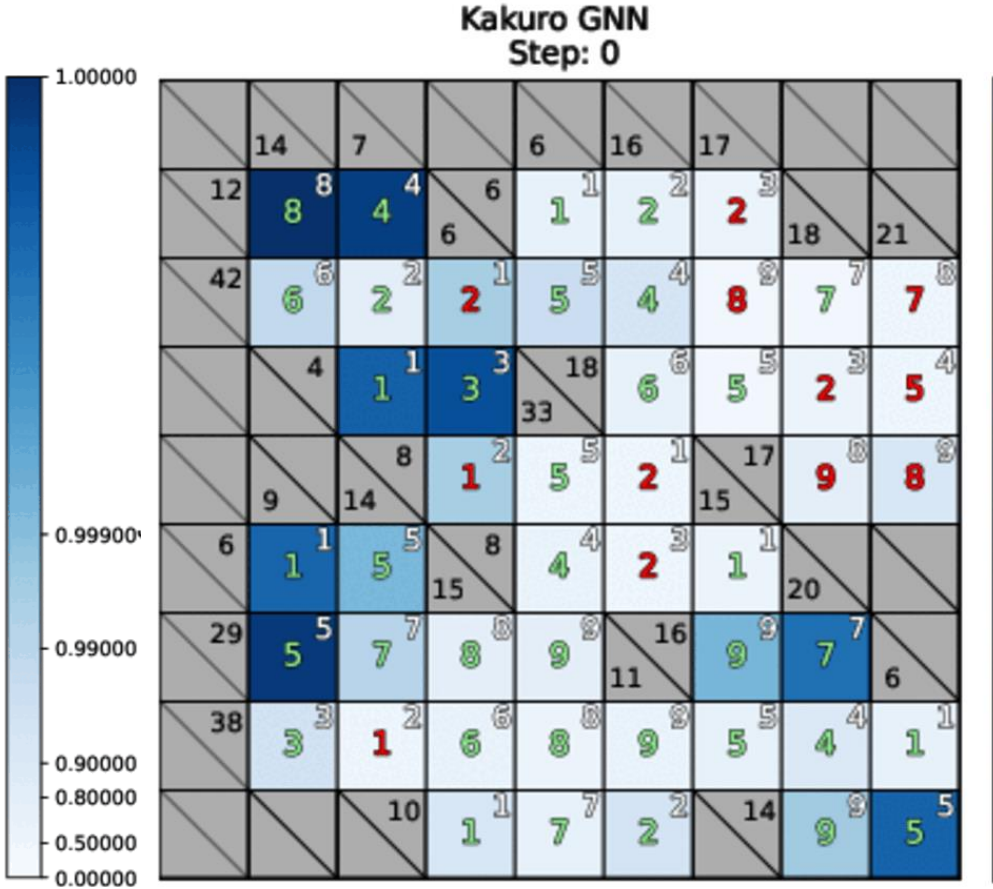
	4	6			2			
	3					7		
7		2		9	8			
		5					2	
	8		5		6		3	
	2					5		
			7	1		2		6
		9					5	
			4			8	9	



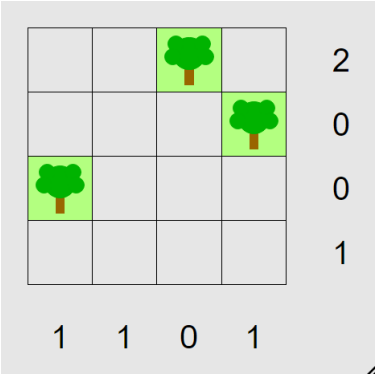
f		9		5			6						b	3	7
	6	b				1	f		7						8
				9		4						2	g		5
5	g	c	8			a	d	3		b		1			
		d	f			2				5		c			
9		4		c	5	g				d	1		e	7	
	b		5		9		3				a	6		f	4
								4	e				a	5	
	4	2				c	e								
a	d		e	1				9		4		f		8	
	5	g		f	2				d	6	8		7		9
			1		a				c			5	2		
			c		4		g	8	a			b	f	2	1
g		a	d						5		6				
7						9		e	b				6	d	
8	1	6						c			4		5		g

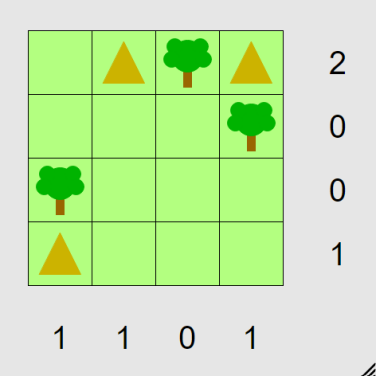


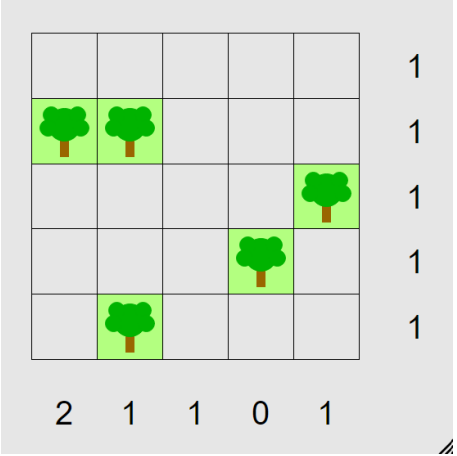

Sudoku RecGNN (Iterative Solving)
Step 0


Kakuro GNN
Step: 0
1.00000
0.99900
0.99000
0.90000
0.80000
0.50000
0.00000

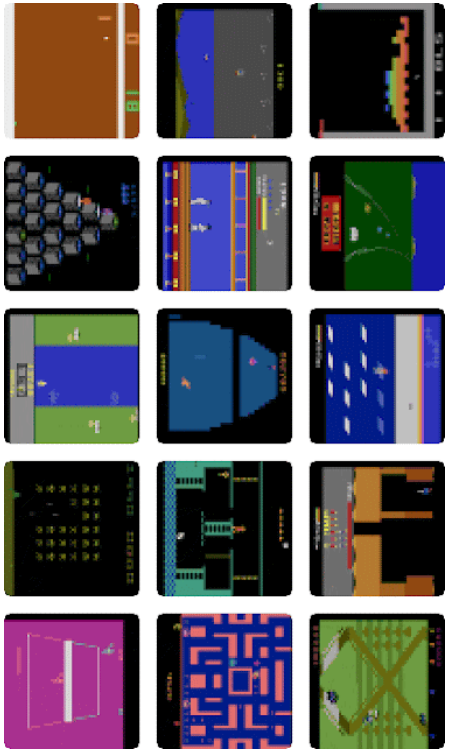
Tents

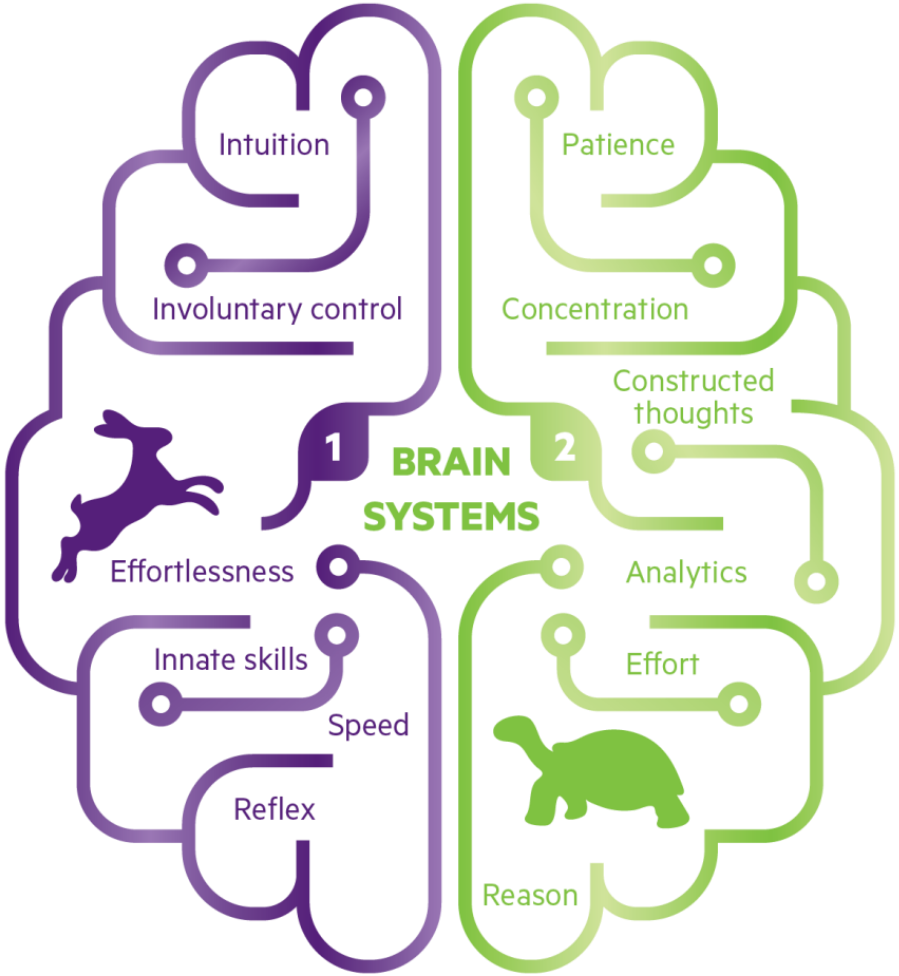




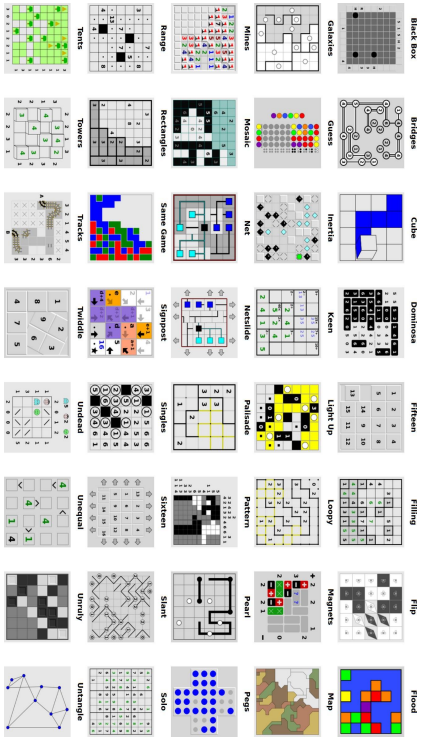


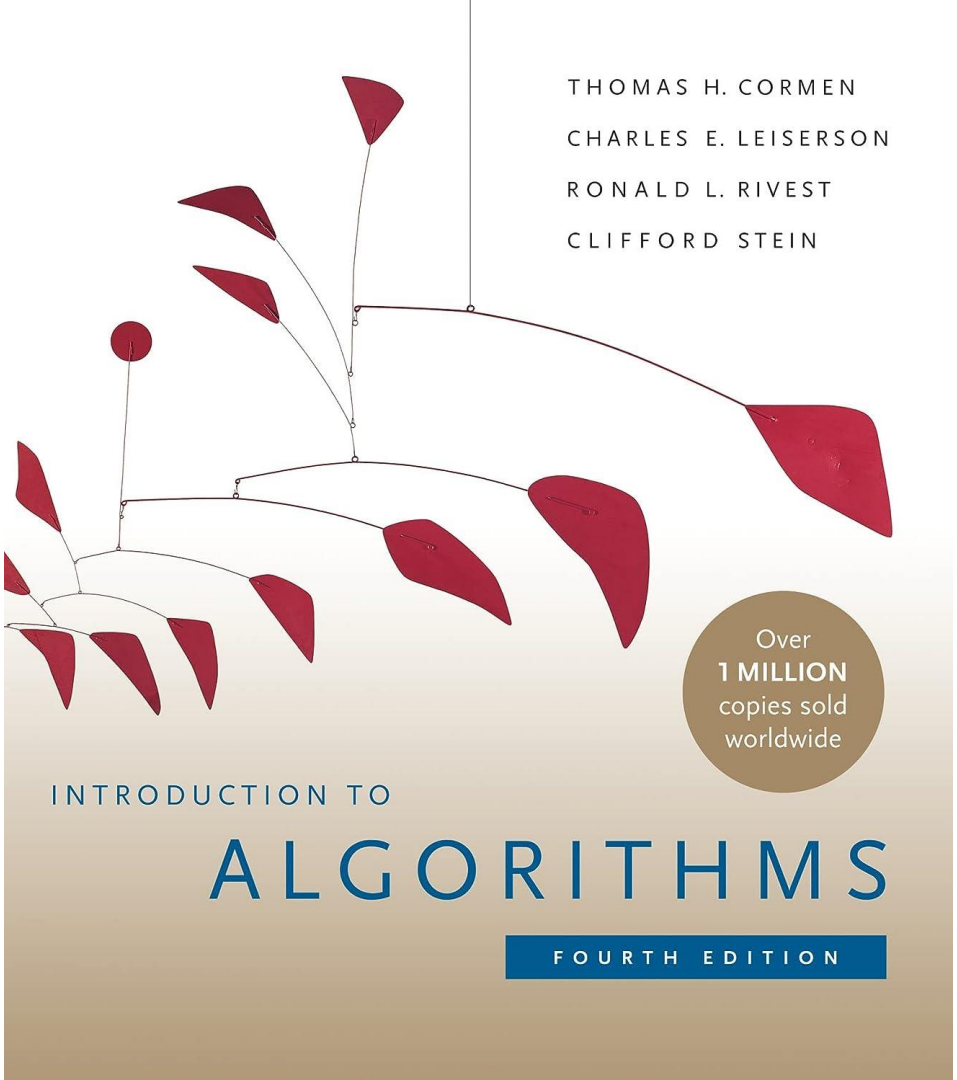
Atari Games





Puzzle Coll.

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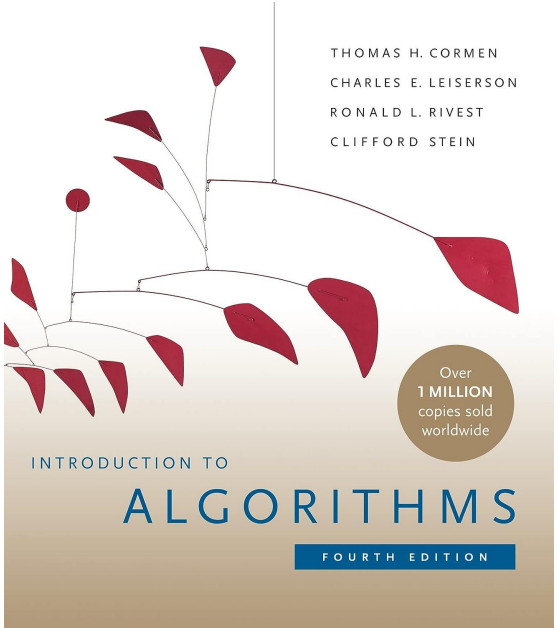
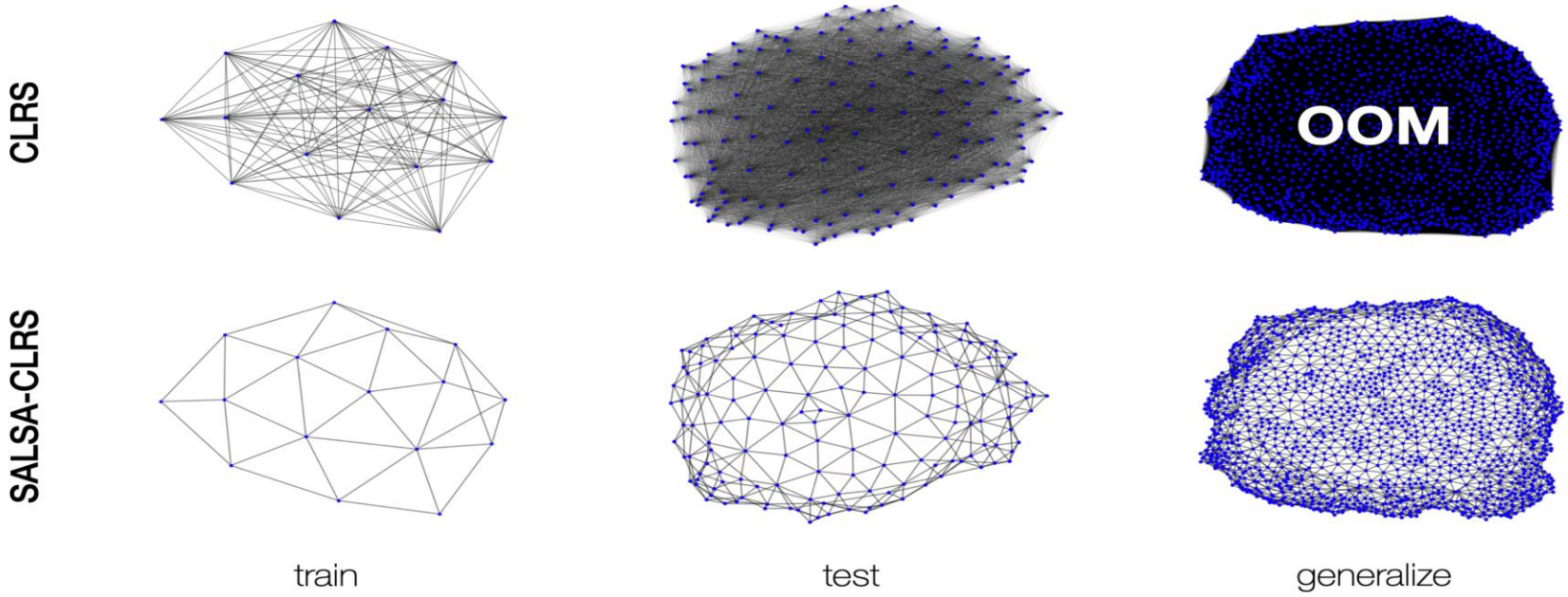
The CLRS Algorithmic Reasoning Benchmark

Petar Veličković [1] **Adrià Puigdomènech Badia** [1] **David Budden** [1]
Razvan Pascanu [1] **Andrea Banino** [1] **Misha Dashevskiy** [1] **Raia Hadsell** [1] **Charles Blundell** [1]



SALSA-CLRS: A Sparse and Scalable Benchmark for Algorithmic Reasoning

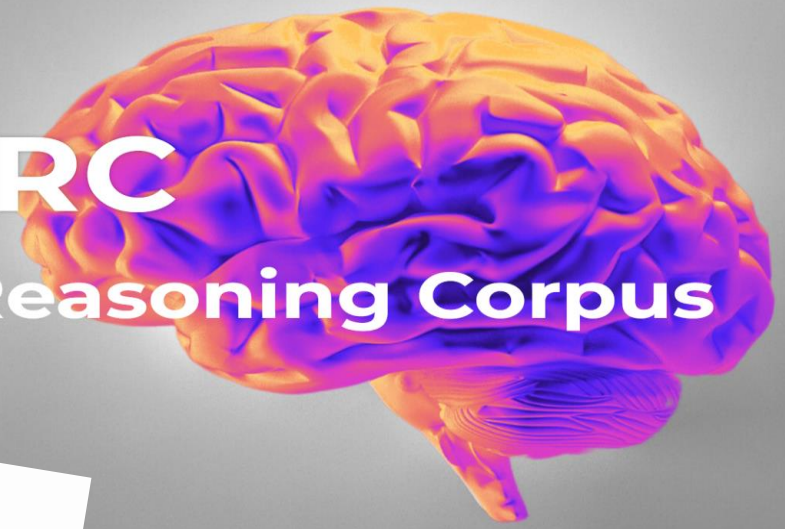
Julian Minder **Florian Grötschla*** **Joël Mathys*** **Roger Wattenhofer**



LAB 42

ARC

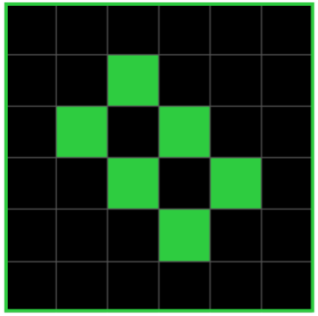
Abstraction & Reasoning Corpus



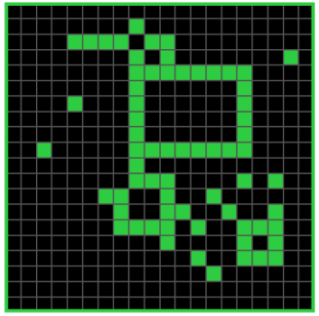
On the Measure of Intelligence

François Chollet *
Google

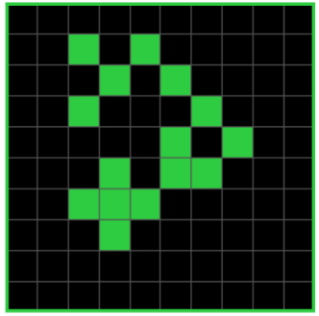
Example 1: Input



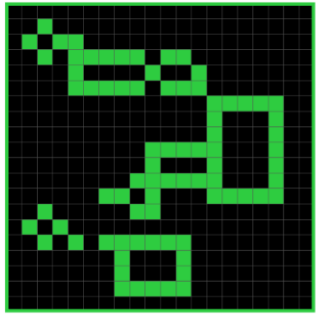
Example 2: Input



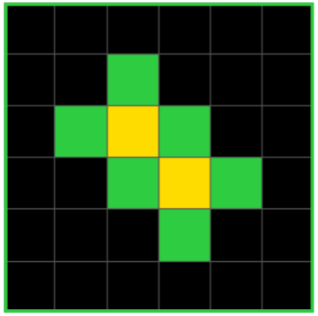
Example 3: Input



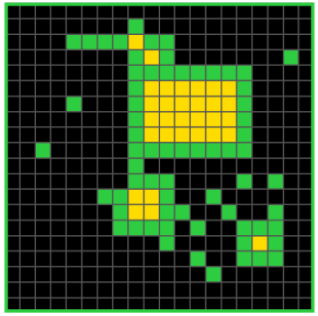
Test: Input



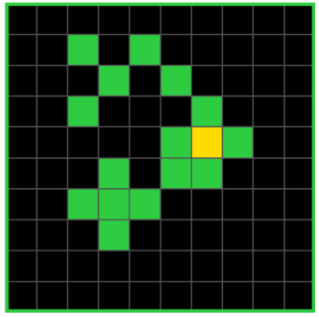
Example 1: Output



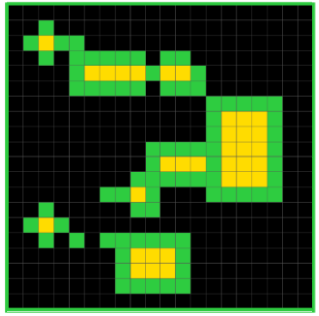
Example 2: Output



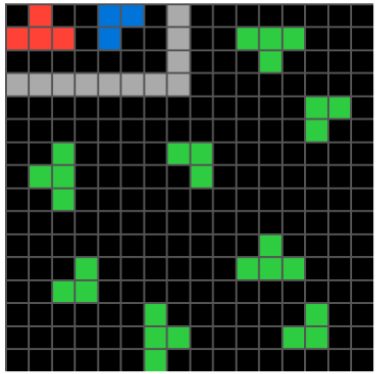
Example 3: Output



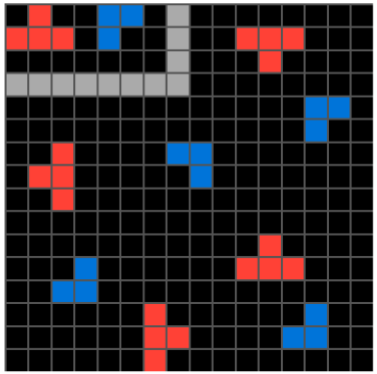
Test: Output



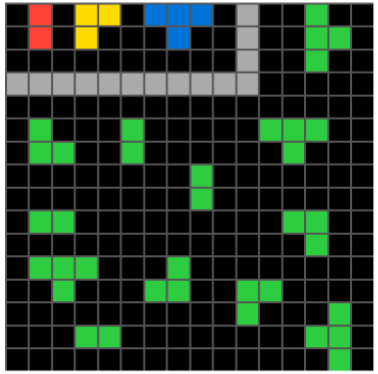
Input 1



Output 1



Input 2



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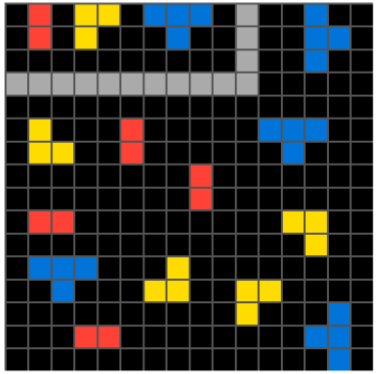

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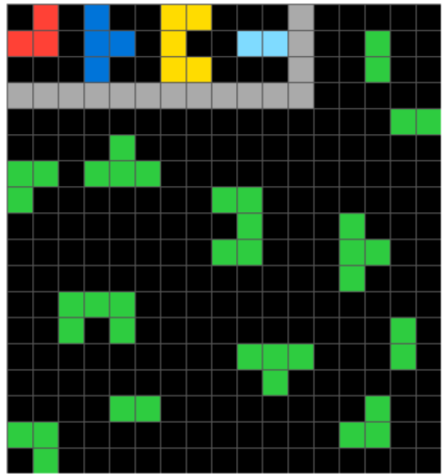

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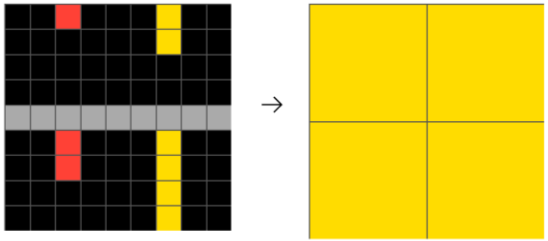
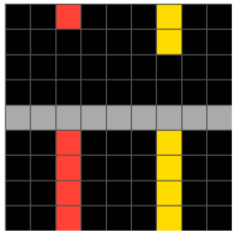
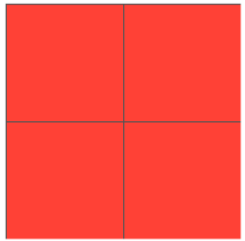


Output 1

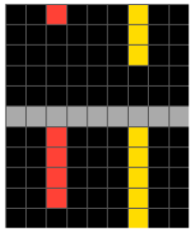
Input 2



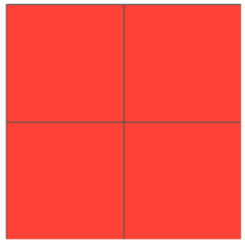
Output 2



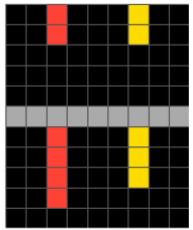
Input 3



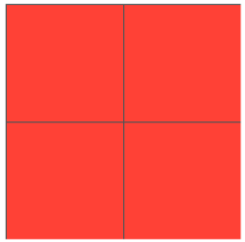
Output 3



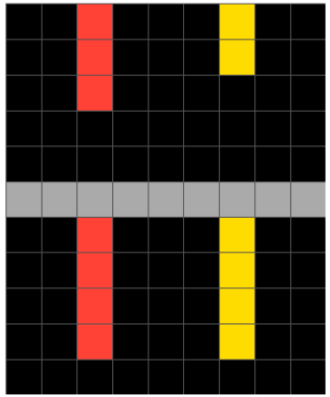
Input 4



Output 4



Input



Abstract Visual Reasoning Enabled by Language

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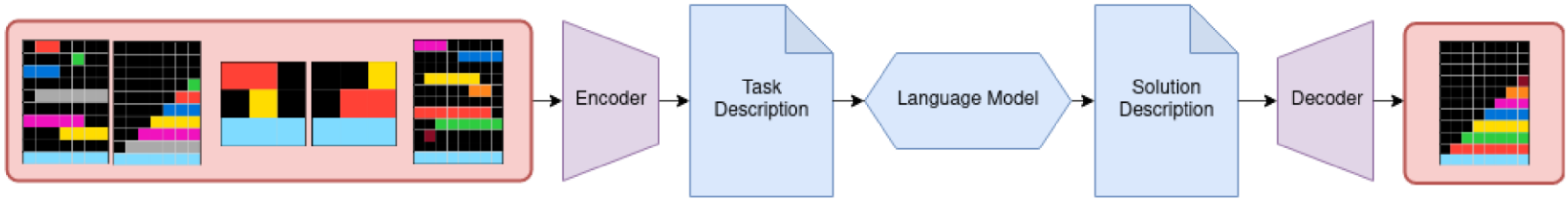
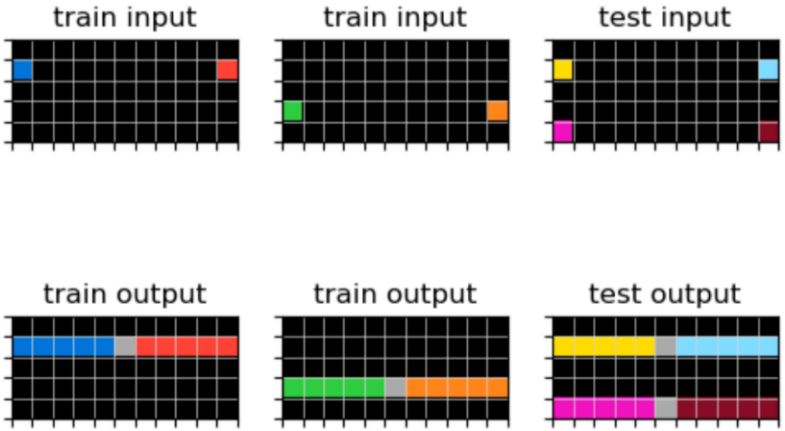
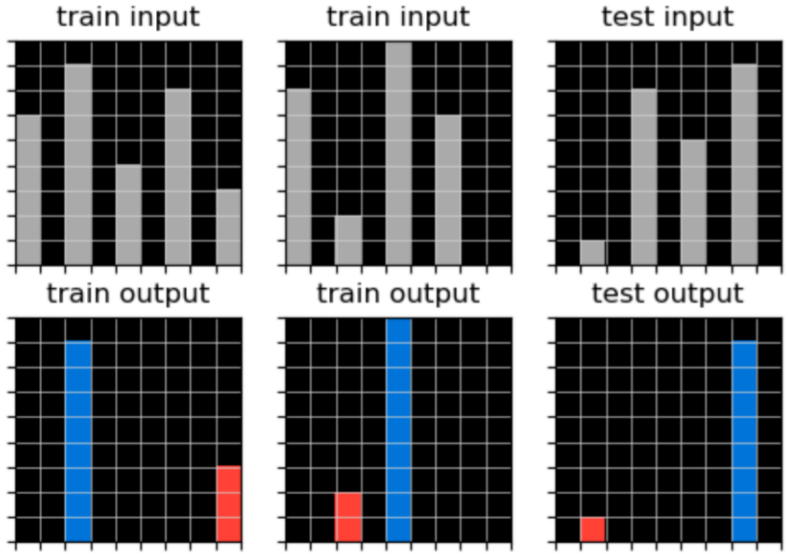
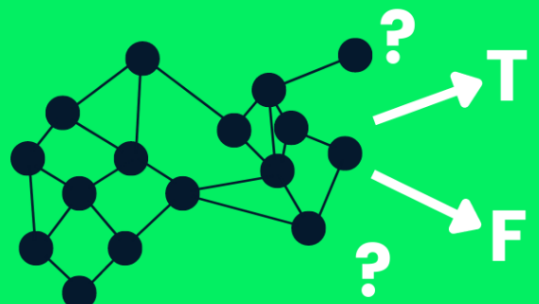


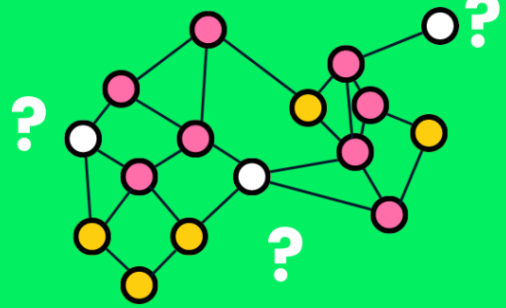

train input
train input
test input
train output
train output
test output


train input
train input
test input
train output
train output
test output

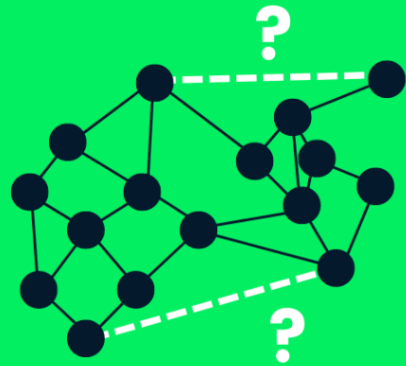
Graph Classification



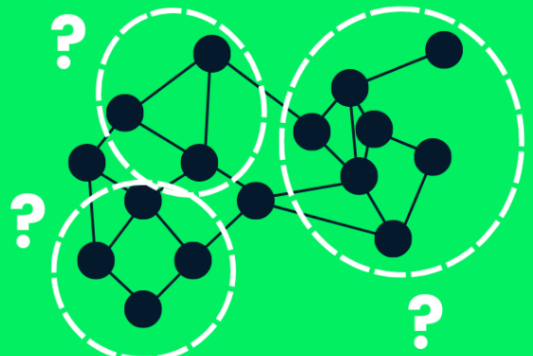
Node Classification



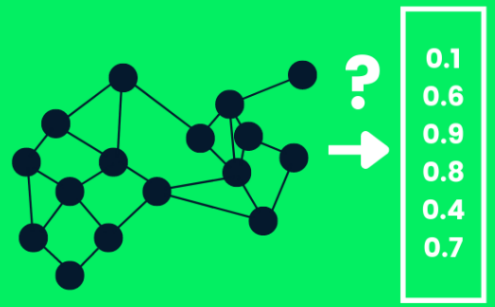
Link Prediction



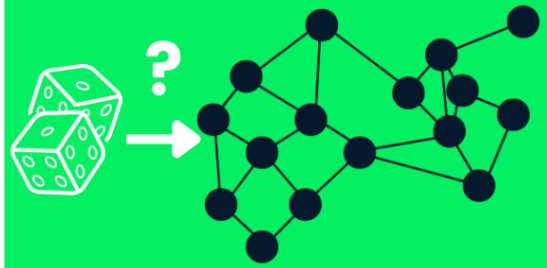
Community Detection



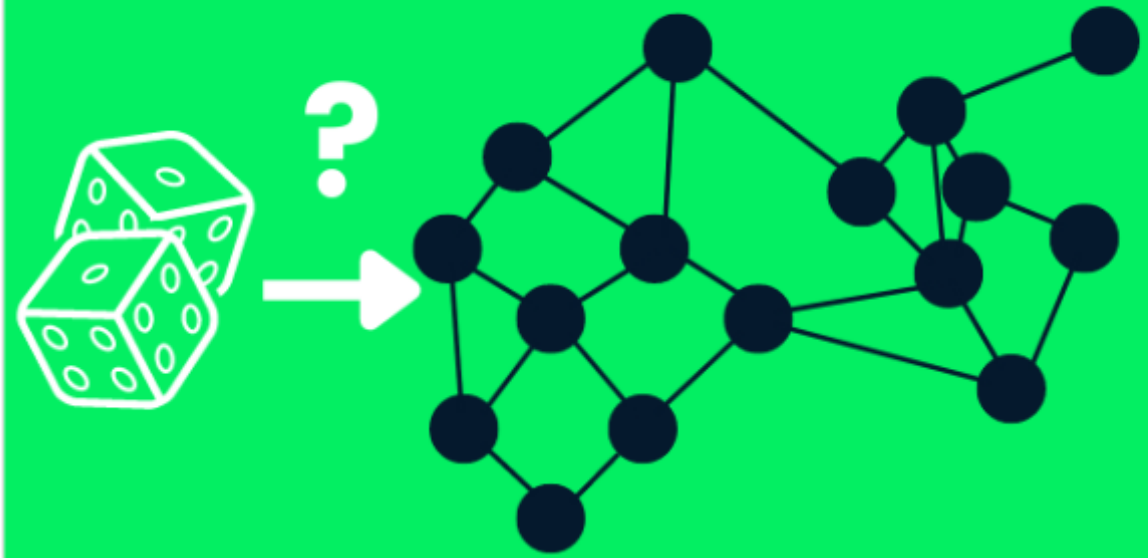
Graph Embedding



Graph Generation

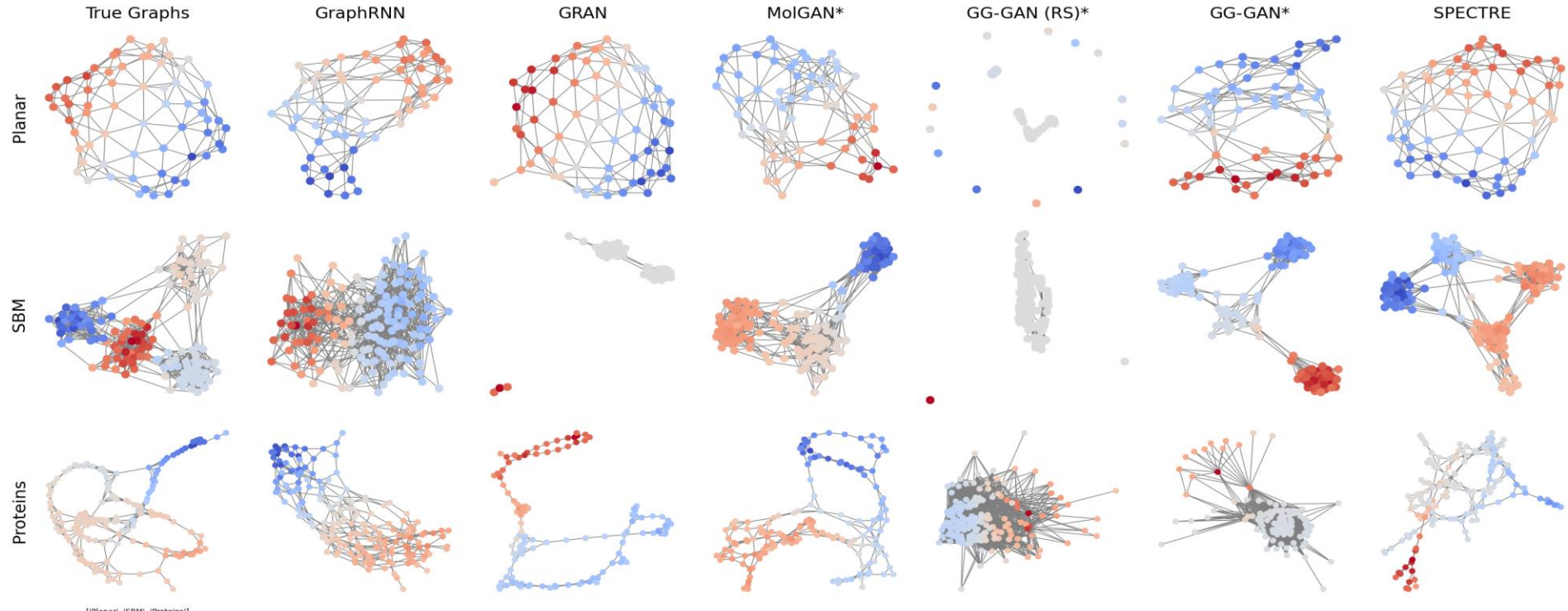


Graph Generation



SPECTRE : Spectral Conditioning Helps to Overcome the Expressivity Limits of One-shot Graph Generators

Karolis Martinkus [1] **Andreas Loukas** [*2] **Nathanaël Perraudin** [*3] **Roger Wattenhofer** [1]



Discovering Graph Generation Algorithms

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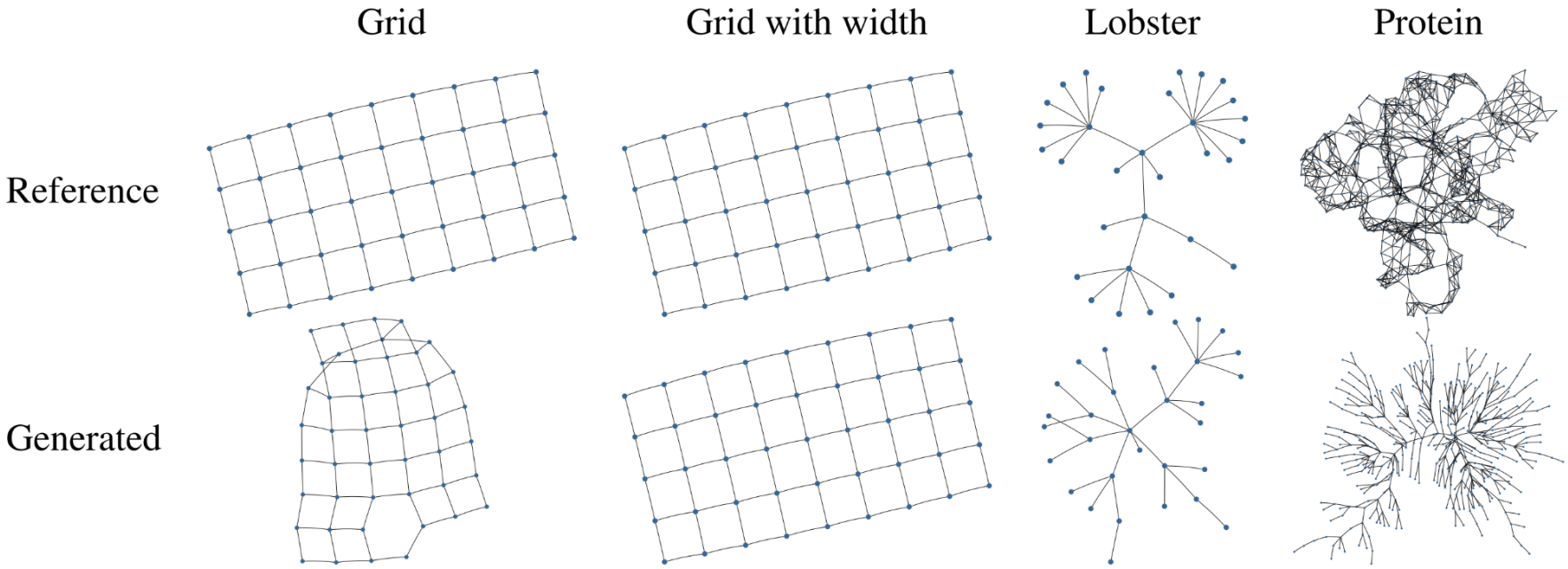
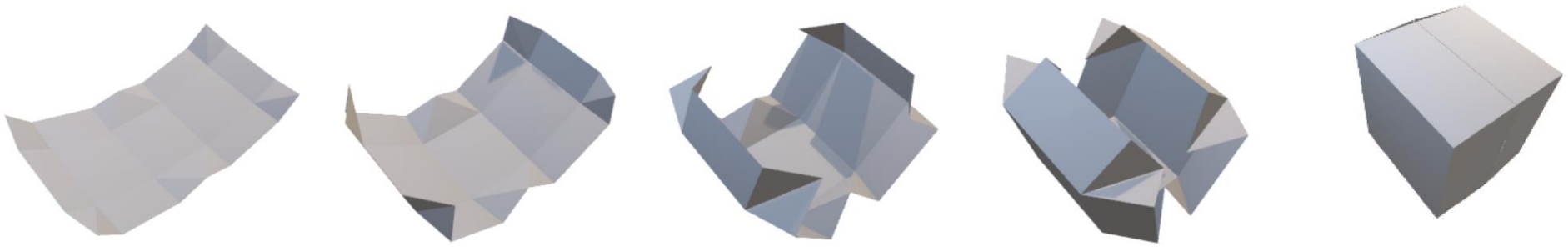
```
def outer_loop():
    for i in range(N):
        inner_loop()

def inner_loop():
    for j in range(i):
        float00 = random(0, 1)
        bool00 = float00 < 0.4
        if bool00:
            add_edge(i, j)

outer_loop()
```

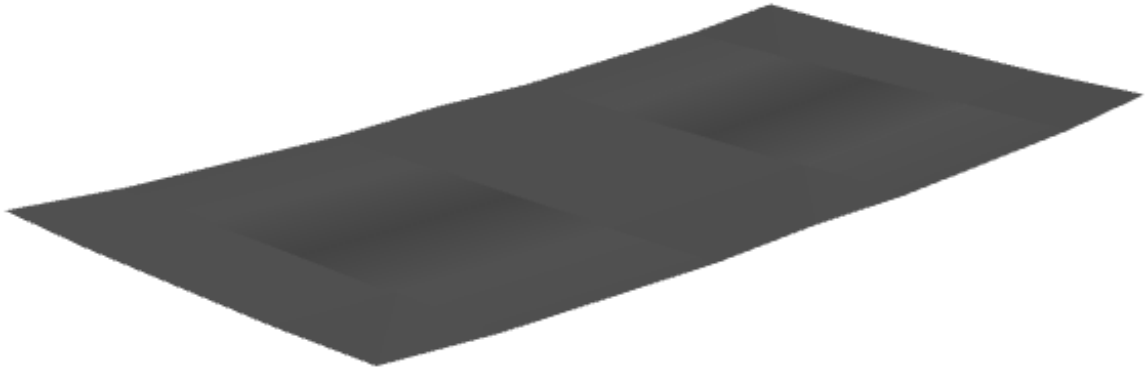
```
def outer_loop():
    for i in range(N):
        int00 = i + n
        add_edge(i, int00)

        int01 = i % n
        bool00 = int01 == 0
        if not bool00:
            int01 = i + 1
            add_edge(i, int01)

outer_loop()
```

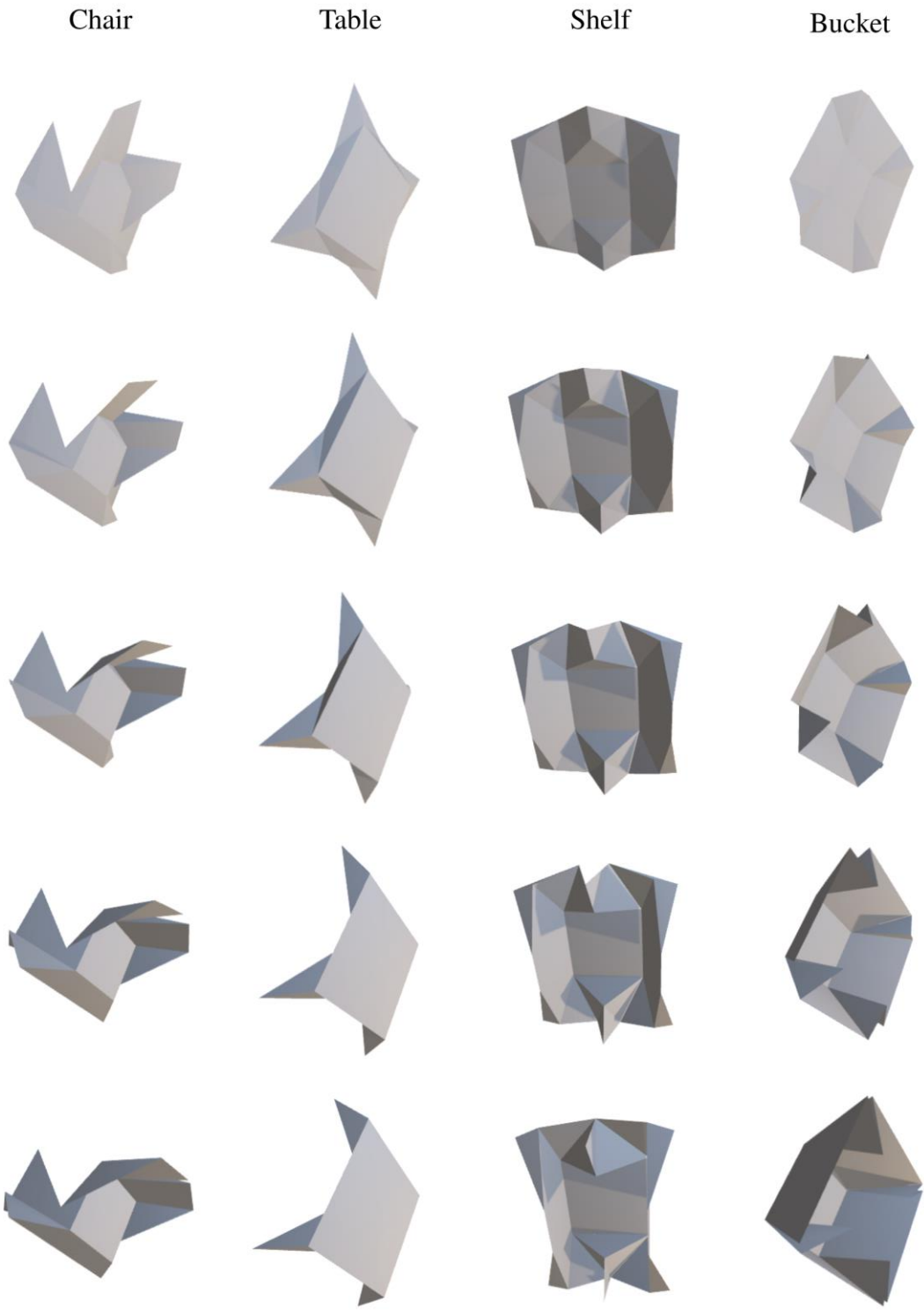

Grid
Grid with width
Lobster
Protein
Reference
Generated

Automating Rigid Origami Design

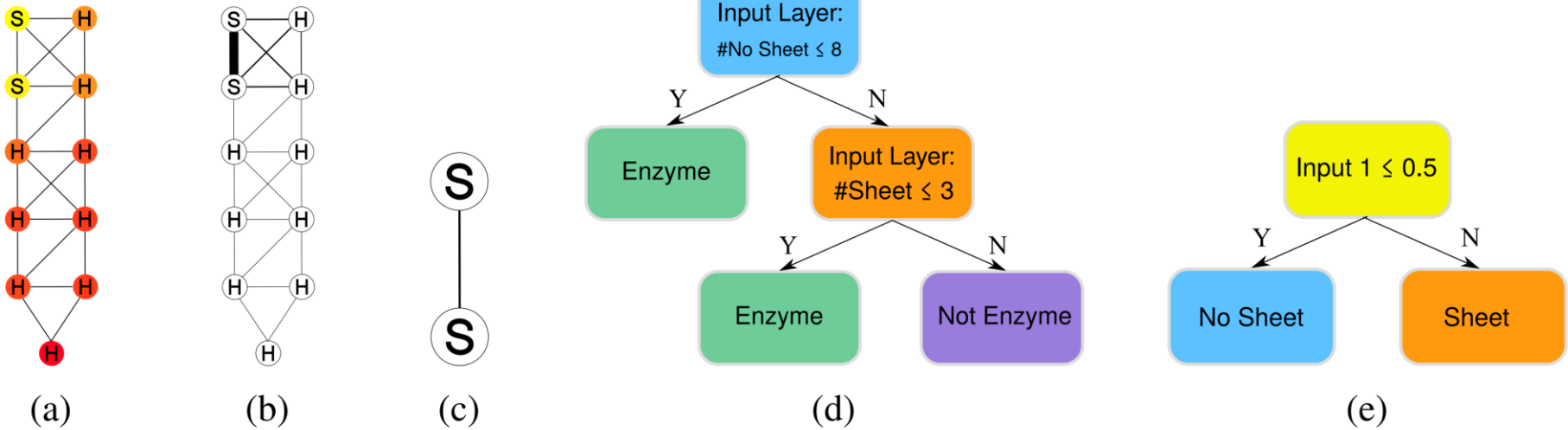
Jeremia Geiger, Karolis Martinkus, Oliver Richter, Roger Wattenhofer





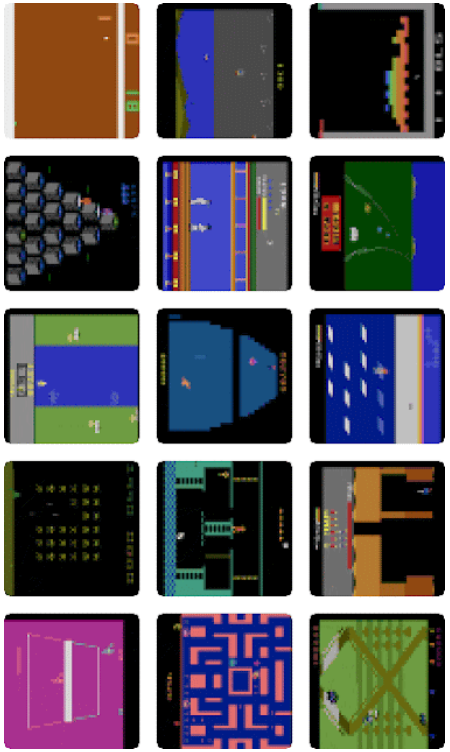

Chair
Table
Shelf
Bucket

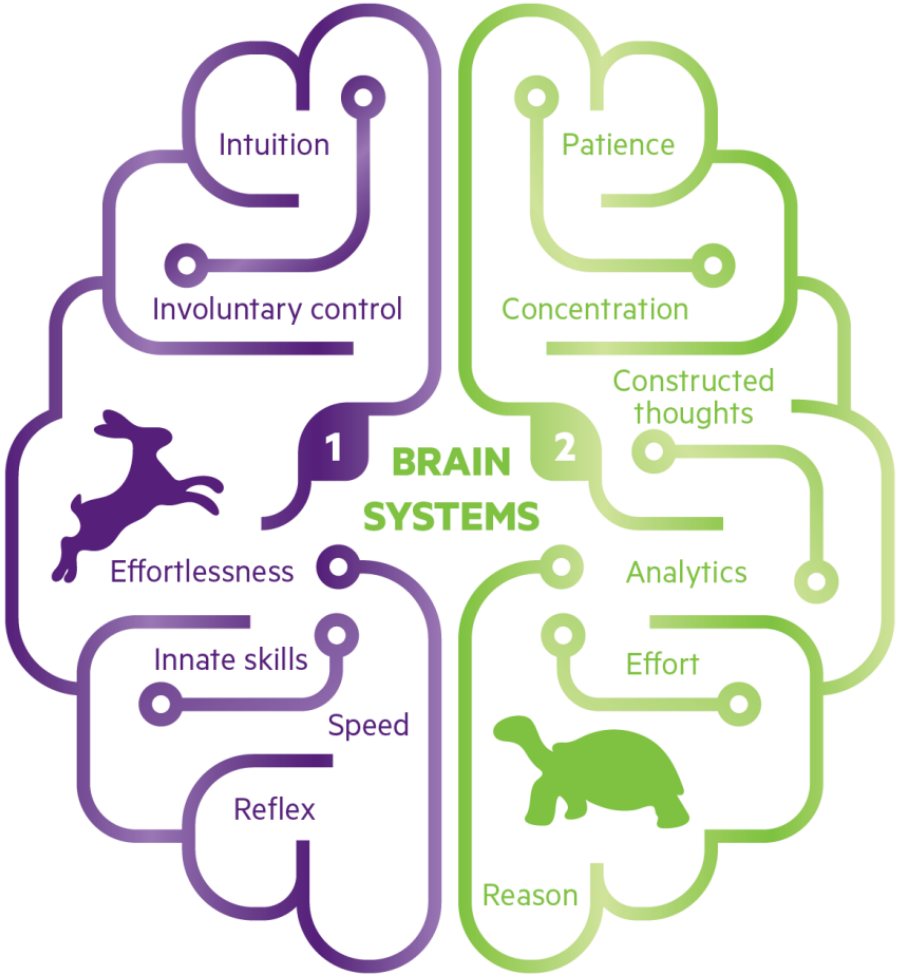
GraphChef: Learning the Recipe of Your Dataset



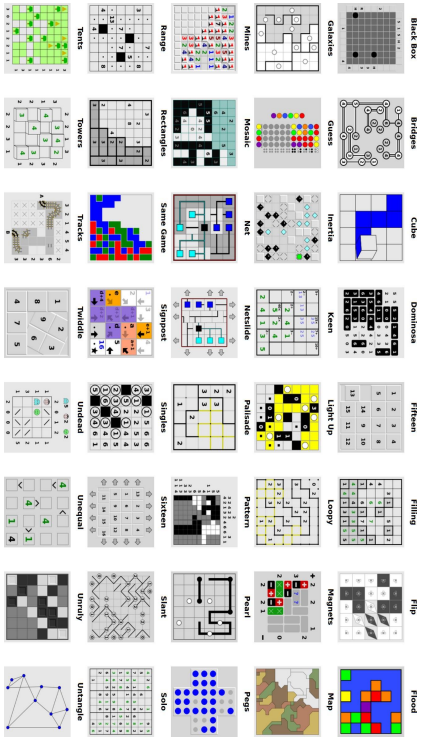

120"
100"
85"
The Bigger Picture

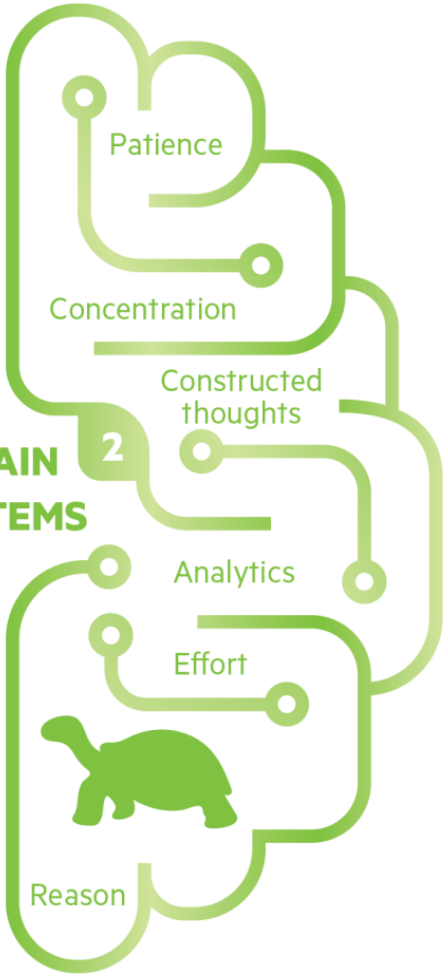
Atari Games





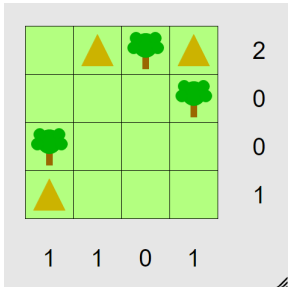
Puzzle Coll.

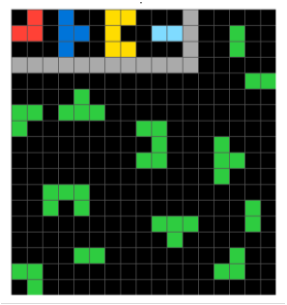


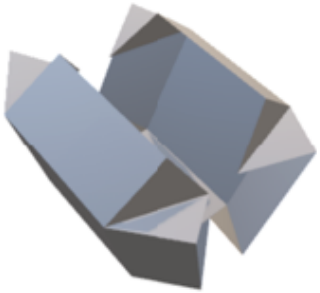


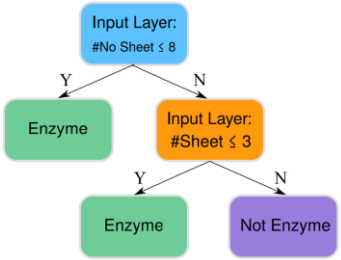
Problem = Graph

Solution = Dist. Learning









Thank You!

Any questions or comments?

Thanks to co-authors: Peter Belcák, Benjamin Estermann, Lukas Faber, Florian Grötschla, Ard Kastrati, Luca Lanzendörfer, Karolis Martinkus, Joël Mathys, etc.

Roger Wattenhofer