

Evolving Cognitive Models of Verbal Learning

Noman Javed

Dmitry Bennett

Genetically Evolving Models in Science

- **GEMS Project:** <https://gems-science.netlify.app/>



CPNSS@



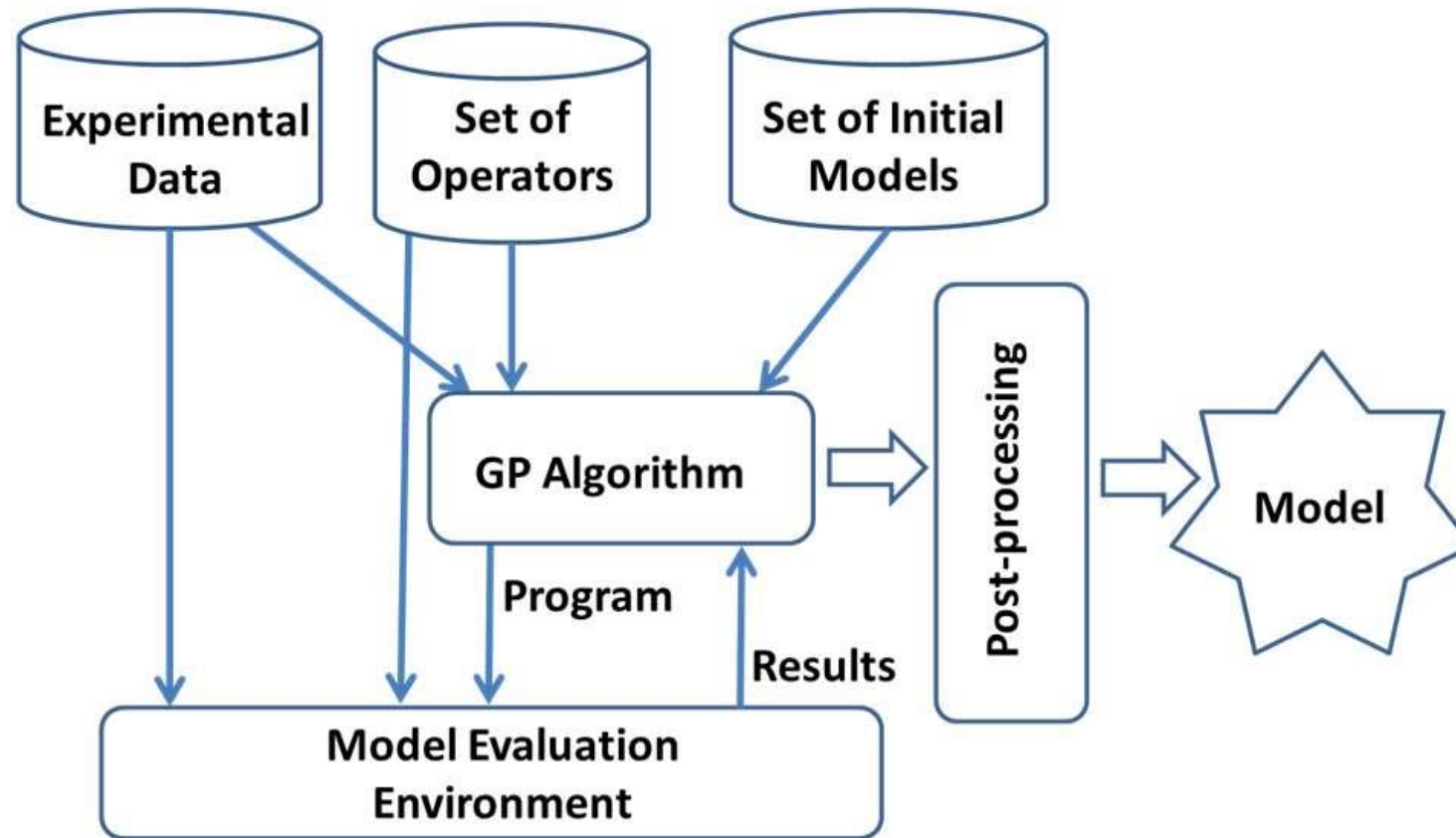
THE LONDON SCHOOL
OF ECONOMICS AND
POLITICAL SCIENCE ■

- This project has received funding from the European Research Council (ERC) under the European Union's Horizon 2020 research and innovation programme (Grant agreement No. ERC-ADG-835002).

Plan

- GEMS
- GEMS Architecture
- CHREST
- Verbal learning
- GEMS interfacing with CHREST
- Experiment 1: Constant Learning Time
- Experiment 2: Intralist Similarity

Genetically Evolving Models of Science (GEMS)

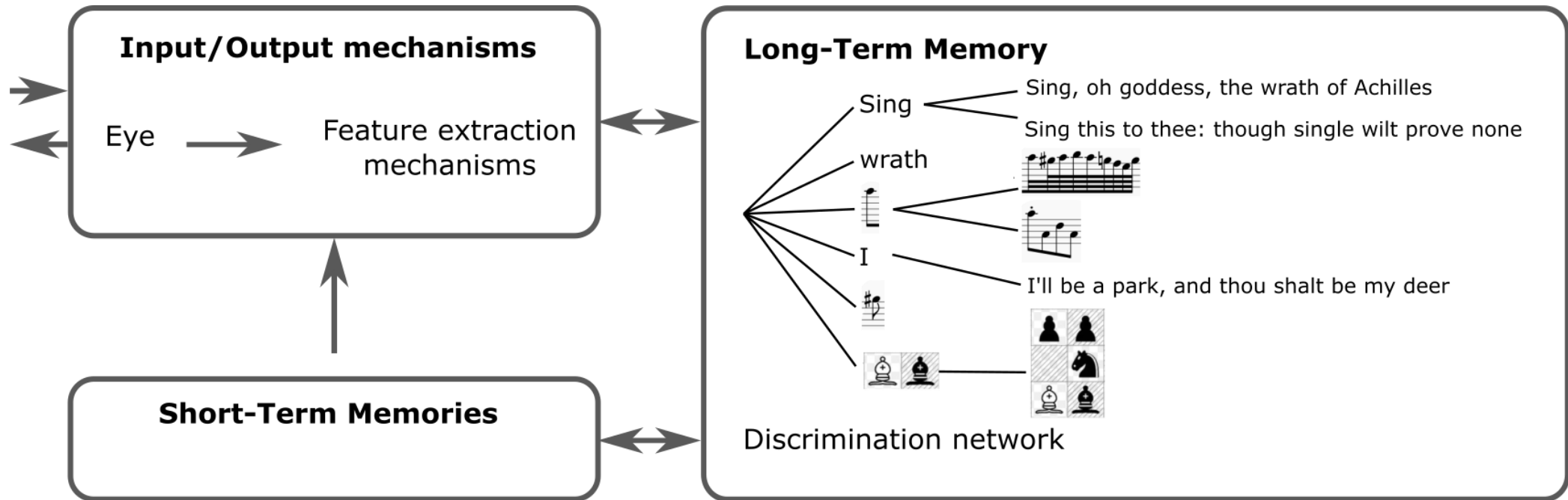


CHREST and Verbal Learning

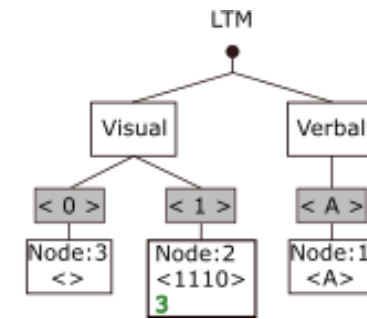
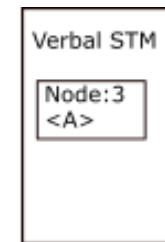
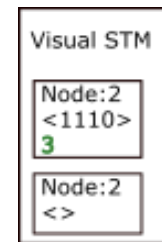
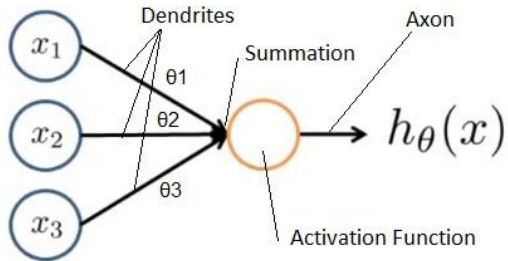
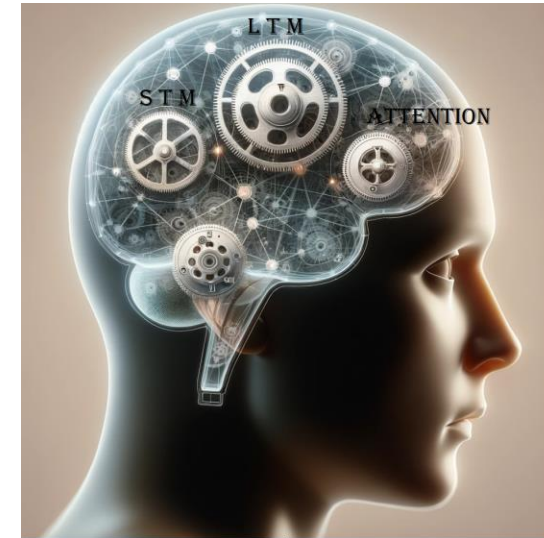
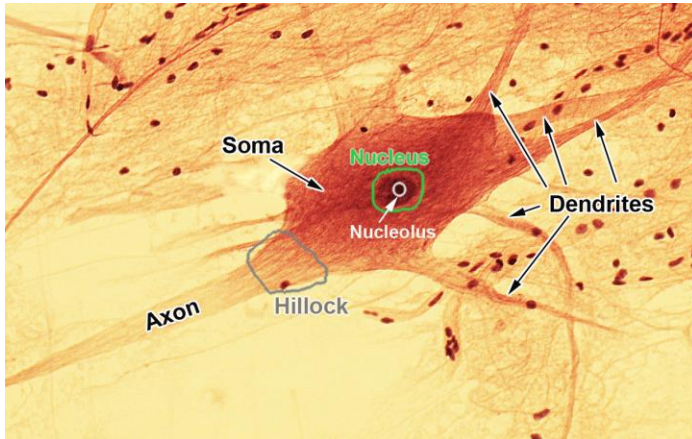
Outline:

- What is CHREST
- Why should we care
- How it works
- Verbal learning simulation

CHREST (Chunking Hierarchy REtrieval Structures)



Why CHREST?



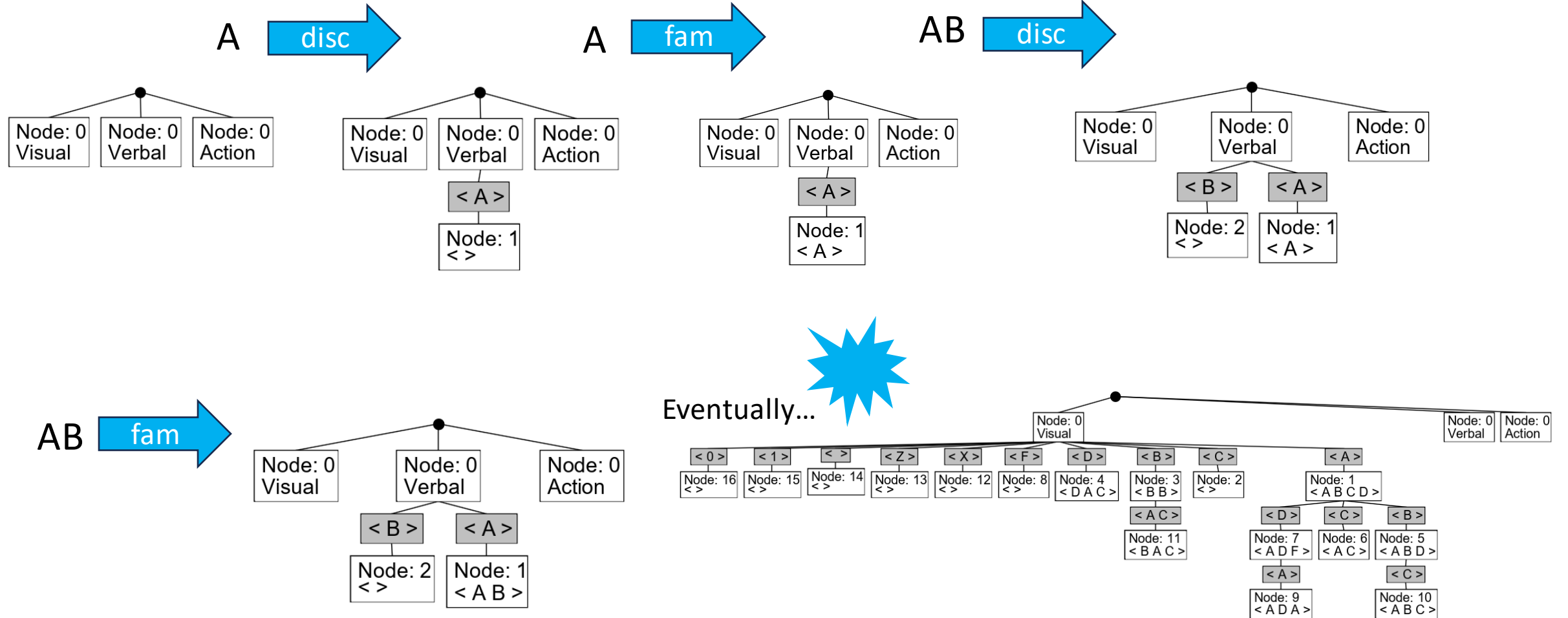
Chunking

827608214703

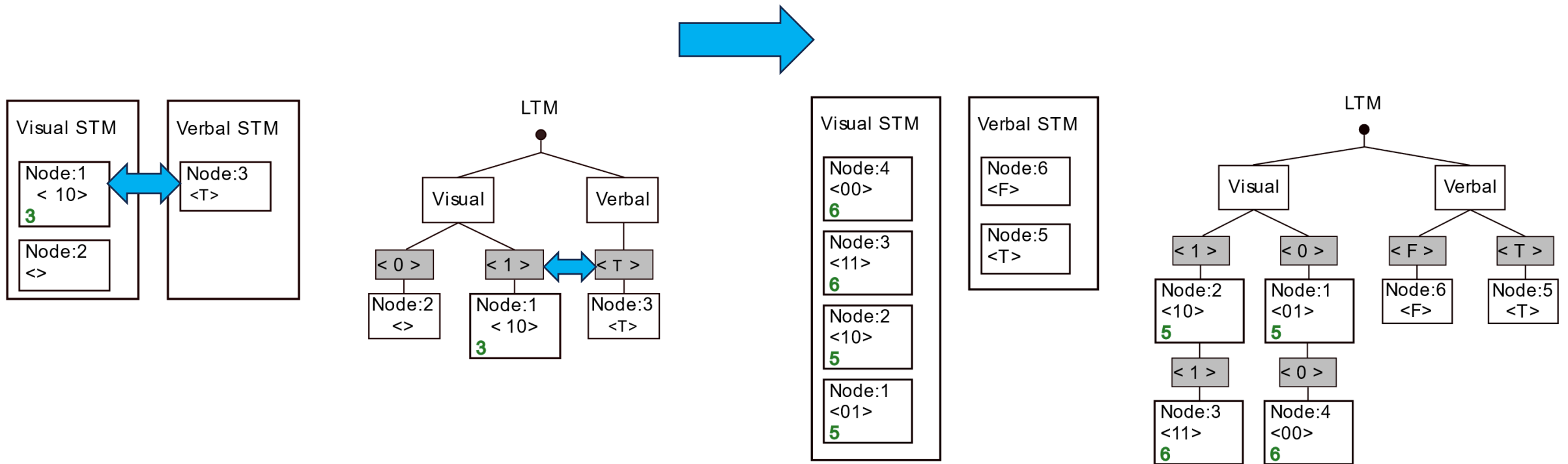
121519562023



Unsupervised chunking: automatic discrimination and familiarisation



Supervised association learning: XOR

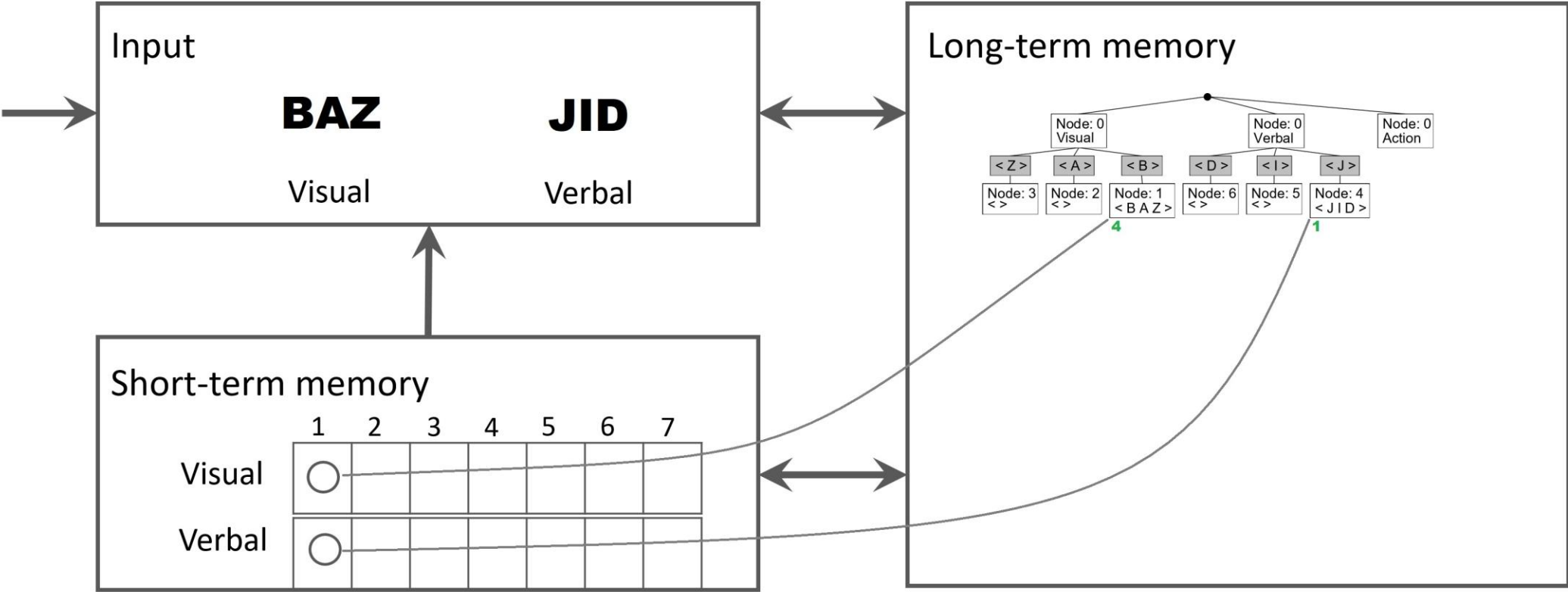


Verbal learning experiments

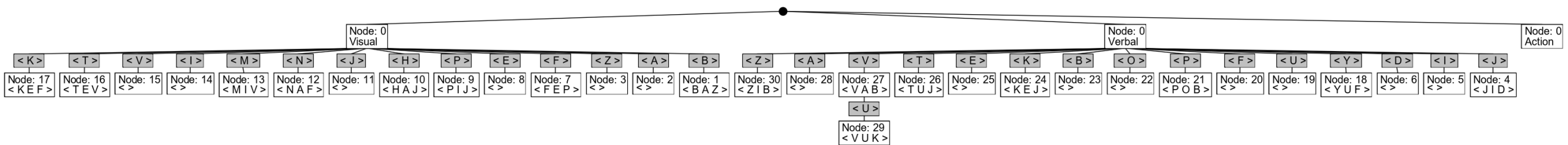
- Nonsense syllable learning
- Participants presented with CVC nonsense syllables
- Trials repeated until learning occurs
- Uncover laws of memory and learning
- Refine computer models through simulations



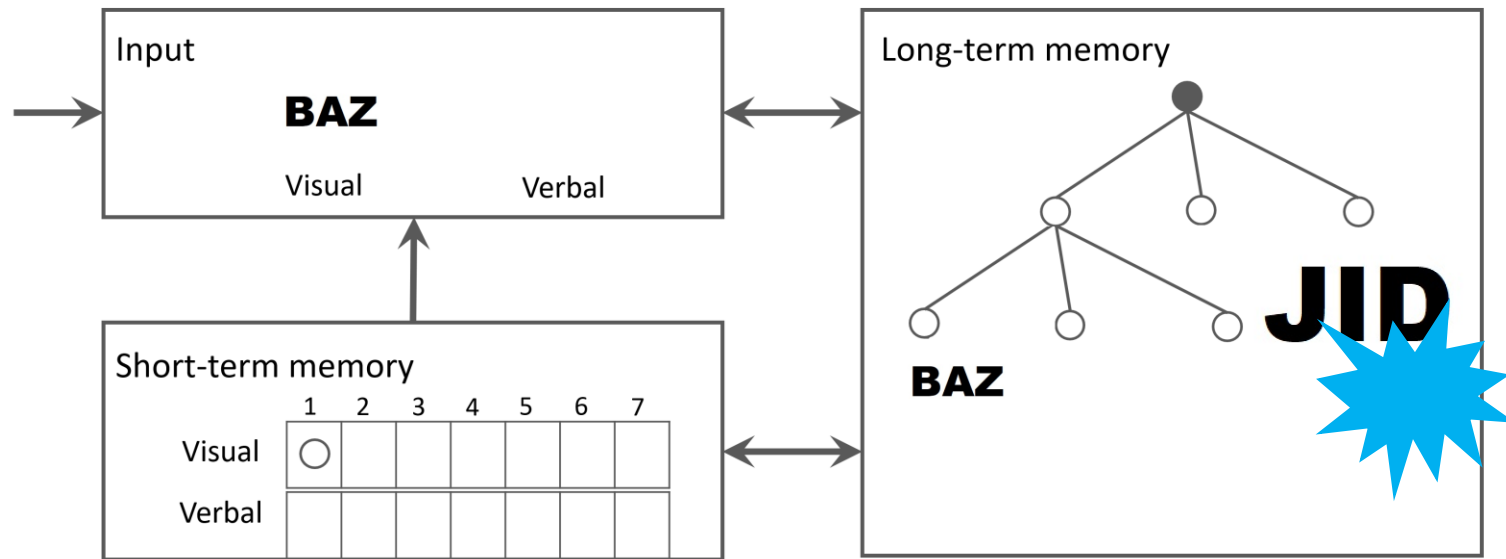
BAZ-JID

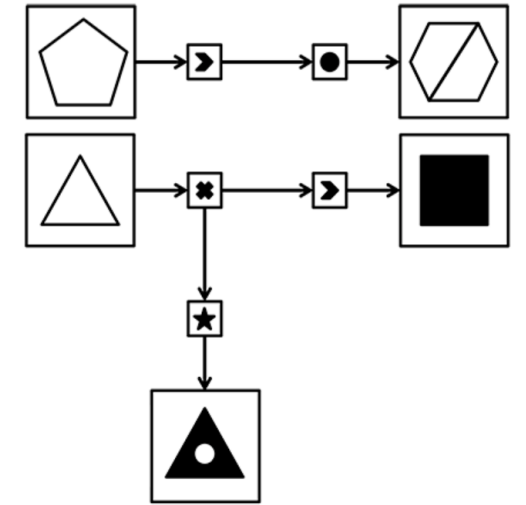
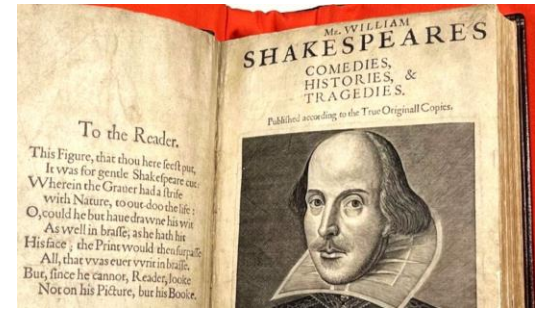
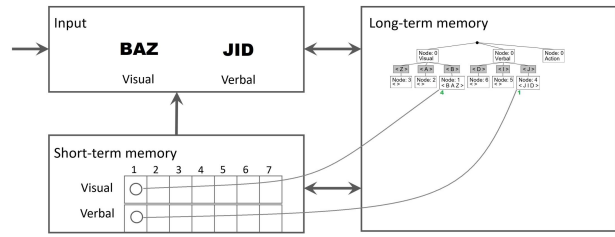
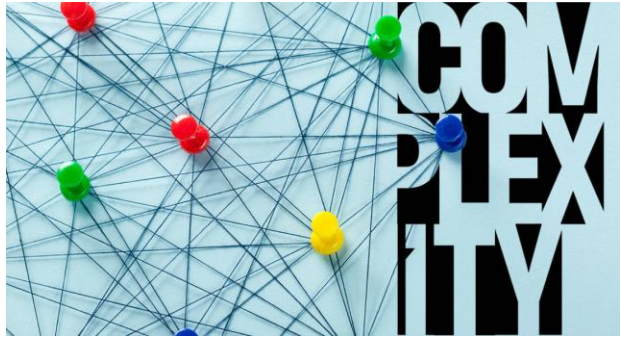
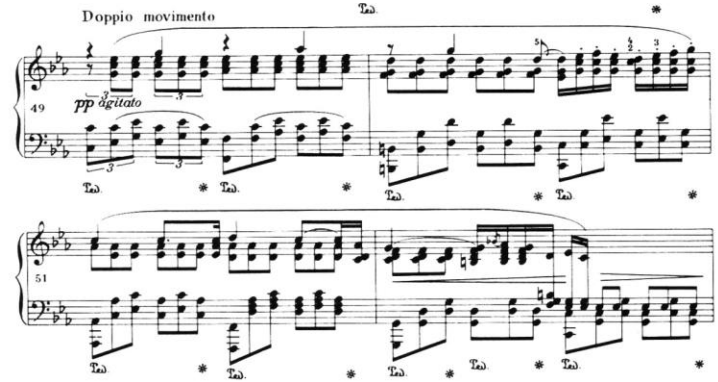
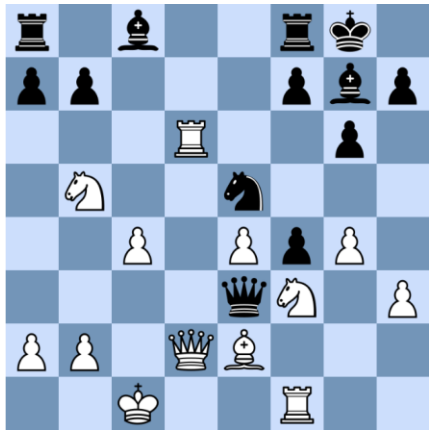


FEP - YUF
 HAJ - POB
 NAF - KEJ
 MIV - TUJ
 TEV - VAB
 KEF - VUK
 PIJ - BIF



BAZ - ?



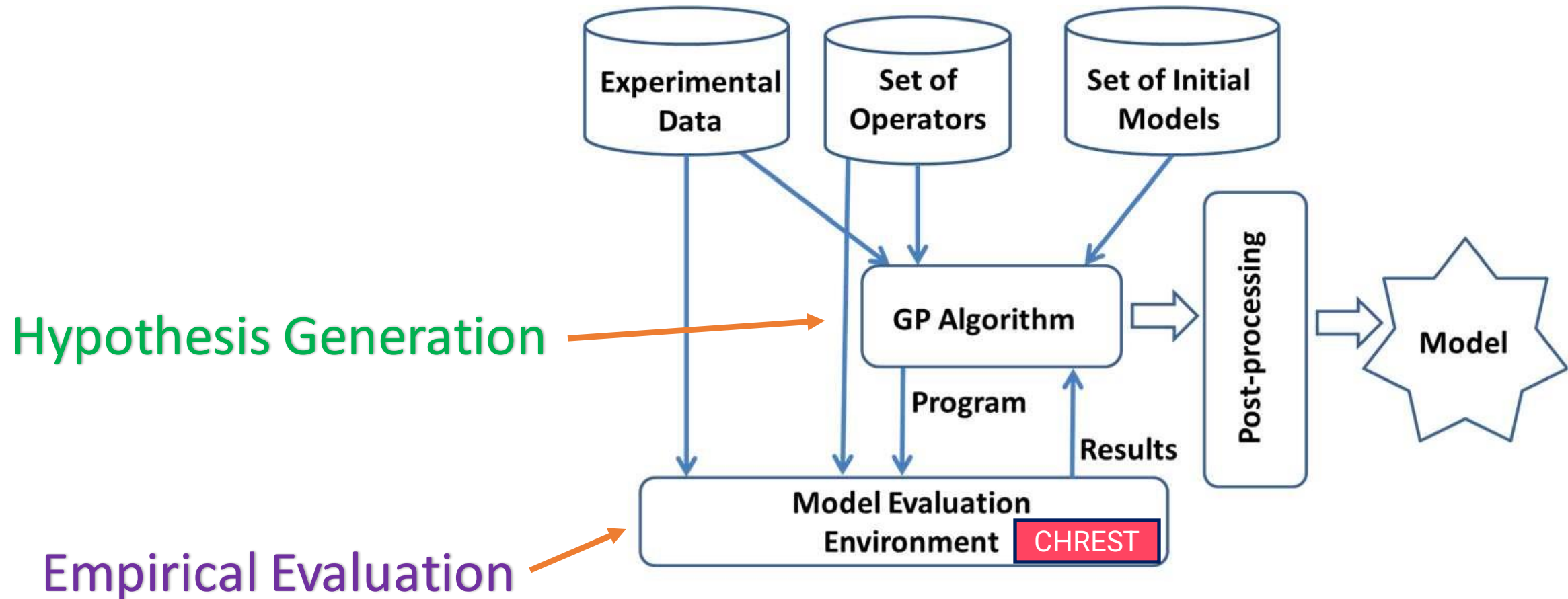


Summary

- CHREST is a general cognitive architecture
- Learns by chunking
- Has *general purpose learning* mechanisms

- Verbal learning – cognitive experiment paradigm
- Uncovers laws of human memory and learning
- Provides building blocks for designing formal cognitive architectures

Genetically Evolving Models of Science (GEMS)



GEMS Operators

Attention

Attend-Stimulus
Attend-Response
Wait

Short-term-memory

Put attended item in STM

LTM (CHREST)

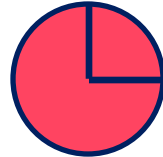
Recognise-and-learn
Learn-and-link

GEMS: Model Evaluation Environment

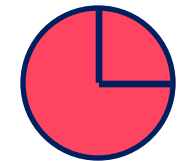
Short-term-
memory



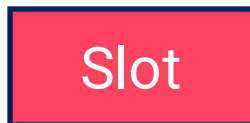
Clock



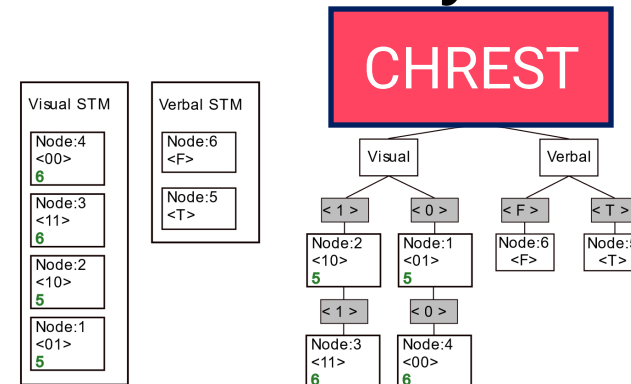
Learning
Clock



Current



Long-term
memory



Bugelski's Experiment: Constant Learning Time

K A R

W E H

Bugelski's Experiment: Constant Learning Time

C E Z

M U N

Bugelski's Experiment: Constant Learning Time

- Total time required to learn a paired-associate list is not affected by presentation rate
- Number of cycles/trials is inversely proportional to the presentation time

Bugelski's Experiment: Constant Learning Time

- List of non-sense syllables are presented until learned completely
- Stimulus presented for 2 seconds
- Response presented for n seconds
- 2 seconds wait before next stimulus appear
- No gap between trials

Constant Learning Time: Results

Presentation Time (sec)	People (trials)	GEMS (trials)
6	10.2	10
8	8.8	9
10	5.8	6
12	4.7	5
19	3.3	3

Constant Learning Time: Results

- GEMS able to generate models learning in the same number of trials
- GEMS models producing errors in recalling list

Underwood's Experiment: Intralist Similarity of Stimuli and Responses

- intralist similarity of stimuli effects learning rate
- intralist similarity of responses does not

Underwood Results

Intralist Similarity	People (trials)	GEMS (trials)
Low-Low	23.2	23
Low-Medium	22.4	22
Low-High	24.4	24
Medium-Low	25.5	25
High-Low	30.7	31

Summary

- GEMS can produce cognitive models for verbal learning
- CHREST acts as the Long-term memory for GEMS

Need further investigation:

- GEMS models producing a lot of errors. Need to compare it with human errors.
- A theory of short-term memory
- Integration of CHREST and GEMS short-term memories