

how does the brain make language?  
towards biologically plausible intelligence

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Fact: AI is still far behind brains on several fronts...

- Natural language interpretation and communication
- Common sense reasoning and understanding of the world
- Creative problem-solving and idea generation
- Flexibility and adaptability to new situations throughout a lifetime
- Emotional intelligence and understanding of social dynamics
- Ability to process information using wetware rather than hardware

**[chat GPT, 2023]**

# But where does one start?

- ***Natural language interpretation and communication***
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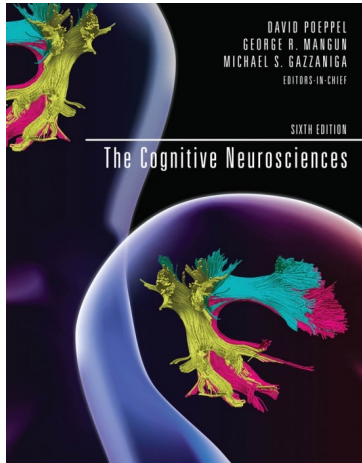
# But where does one start?

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# Where does one start? **My approach**

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# *How does the Brain create the Mind?*



*“...we do not have a **logic** for the transformation of neural activity into thought ... I view discerning [this] **logic** as the most important future direction of neuroscience”*

**Richard Axel, *Neuron*, Sep 2018**



So, what kind of formal computational theory would qualify as Axel's *logic*?





# Our approach to Axel's logic

Cognitive phenomena



implements

***math model of the brain***



models reasonably well

Spiking neurons and synapses

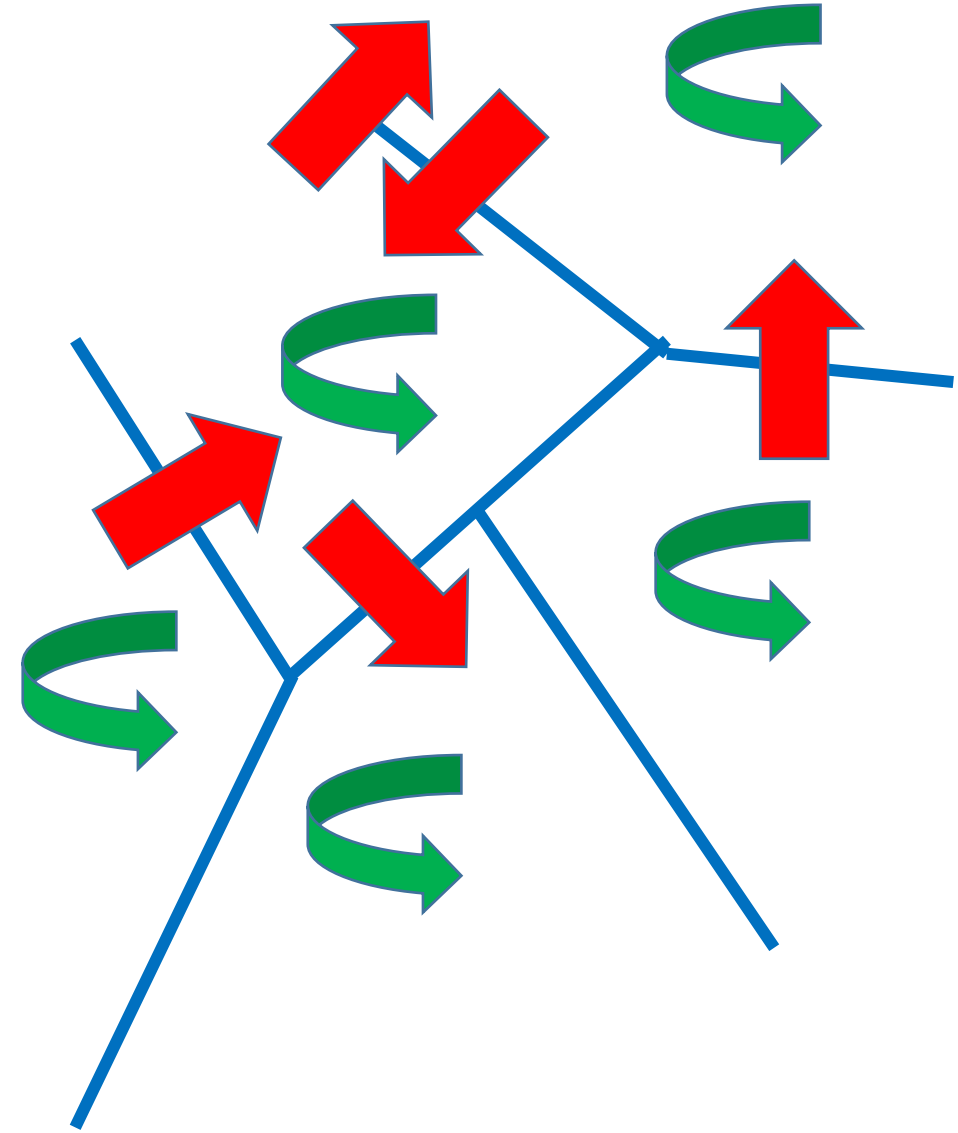
# A math model of the brain

A finite number of brain areas

Each has  $n$  excitatory neurons

Certain pairs of areas are connected  
by fibers of **random** connections

All areas are recurrently connected  
by **random** synapses

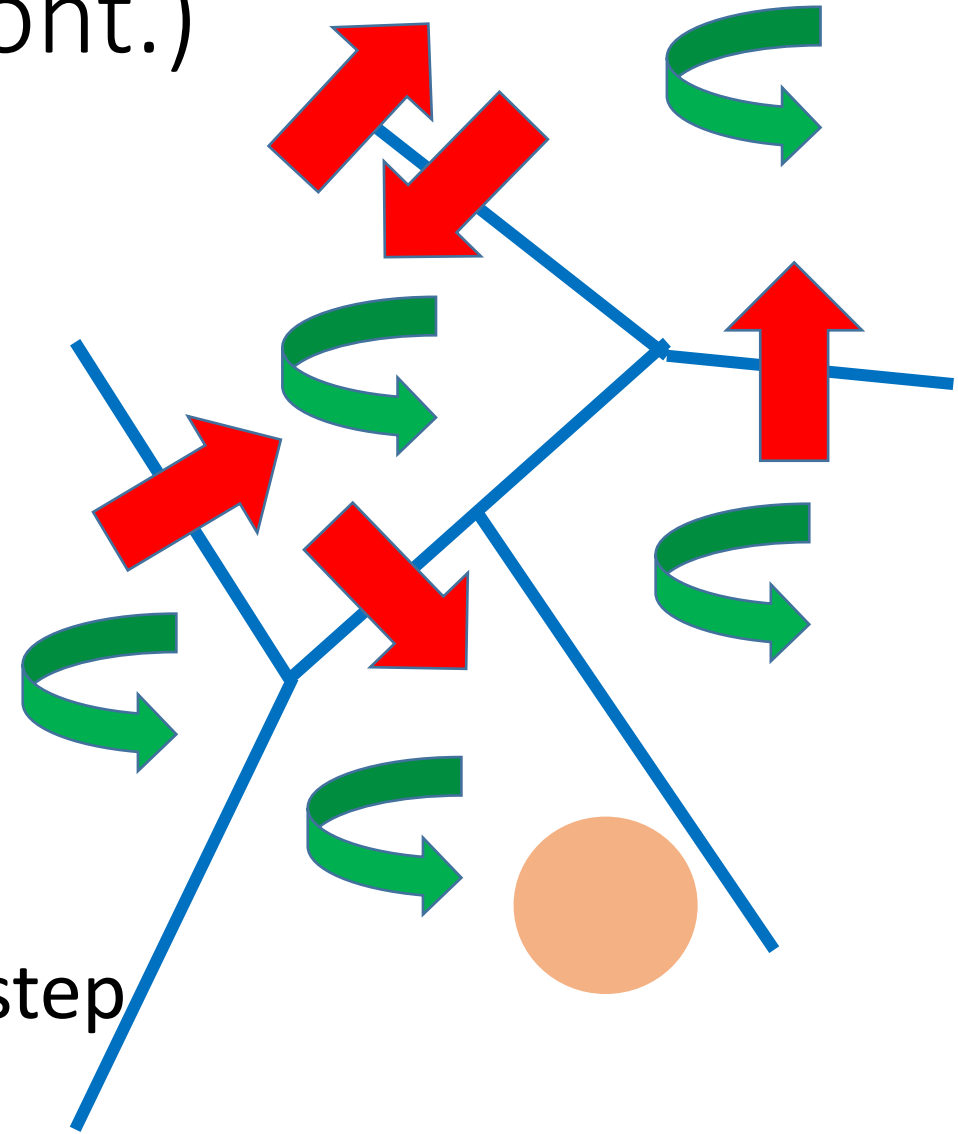


# A math model of the brain (cont.)

Neurons fire in **discrete steps**

At each step, in each area,  
the  $k \ll n$  neurons fire,  
those with the highest synaptic input  
from the previous step  
(This models local inhibition)

Also: an area can be **inhibited** during a step



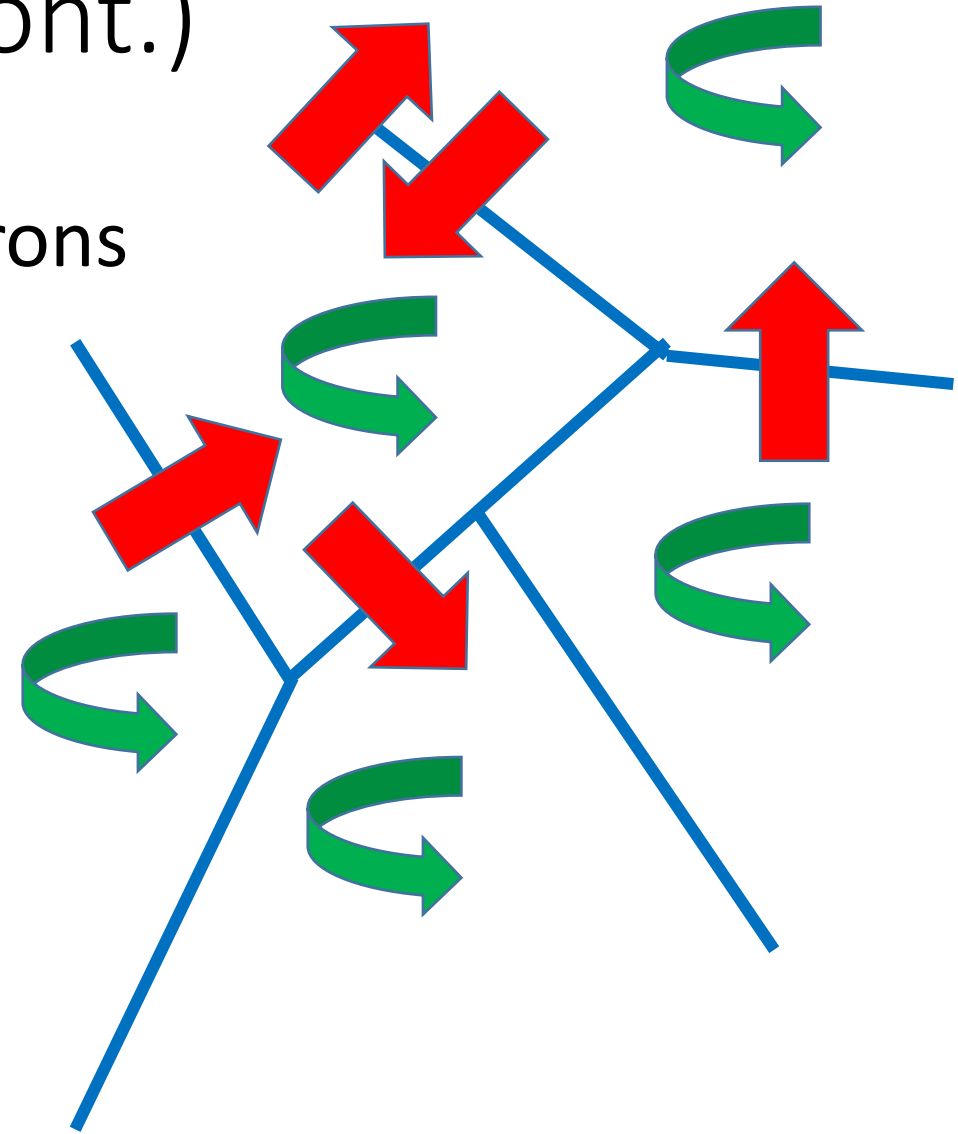
# A math model of the brain (cont.)

Finally, **plasticity**: if two connected neurons “fire together” then the weight of their synapse is multiplied by  $(1 + \beta)$

**Plasticity, Randomness, Selection**

Typical Parameters:

$n \sim 10^7$ ,  $k \sim 10^4$ ,  $p \sim 10^{-2}$ ,  $\beta \sim 5\%$



# A math model of the brain

(Also: homeostasis, forgetting)

This defines a **dynamical system**

**State:** neurons that spiked, synaptic weights, inhibited areas

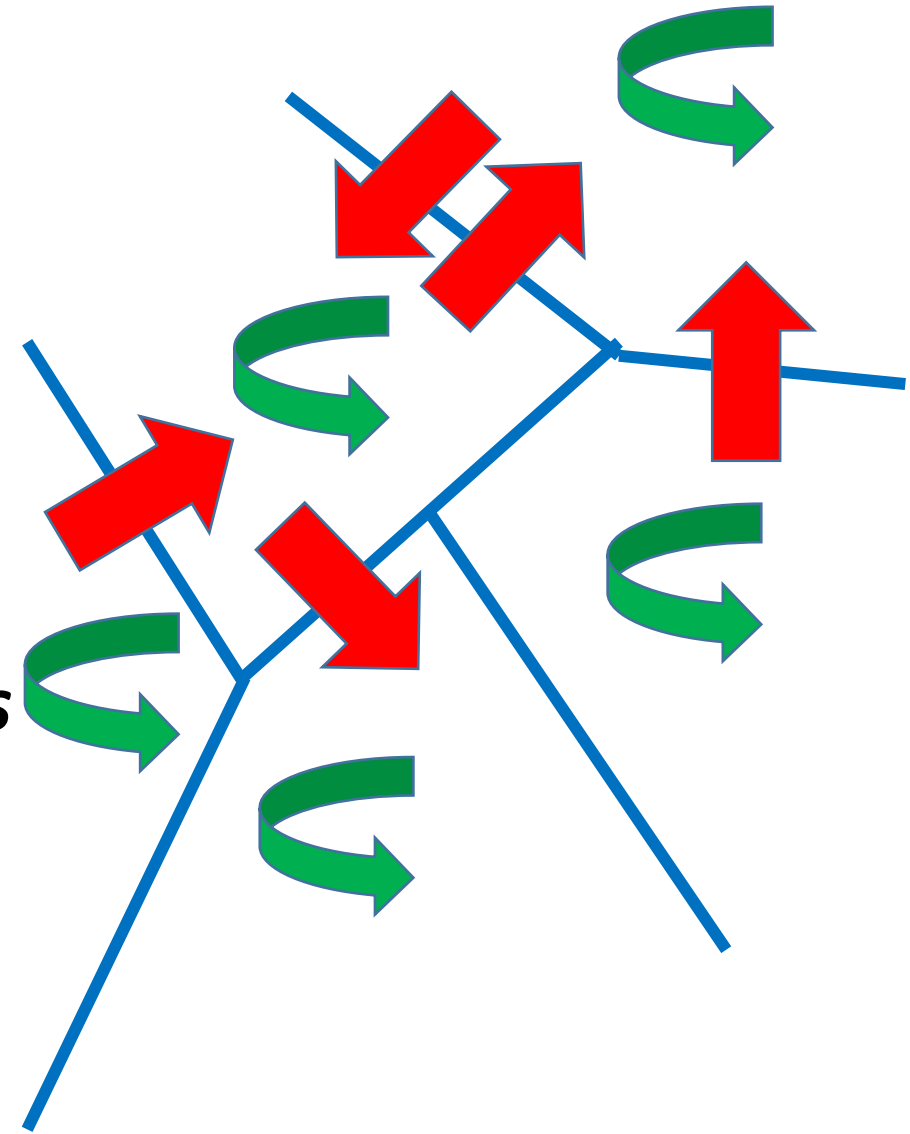
The next step function: fully defined!

Emergent behavior: ***assemblies of neurons***

Simulator available online

Q: But how is computation initiated?

A: Typically, by external stimuli



# Short history of assemblies of neurons

- [Hebb 1949, Harris 2003, 2005; Buzsaki 2008, 2010, Yuste 2019,...]
- Assembly: A large *and densely interconnected* set of excitatory neurons in a brain area whose near simultaneous firing is tantamount to the subject's thinking of a particular memory, concept, person, name, word, episode, etc.
- ***G. Buzsaki 2020: "assemblies are the alphabet of the brain"***

# Assembly operations

- Projection
- Reciprocal projection
- Association
- Pattern completion
- Merge
- Sequence recall
- Few shot learning of simple classification tasks

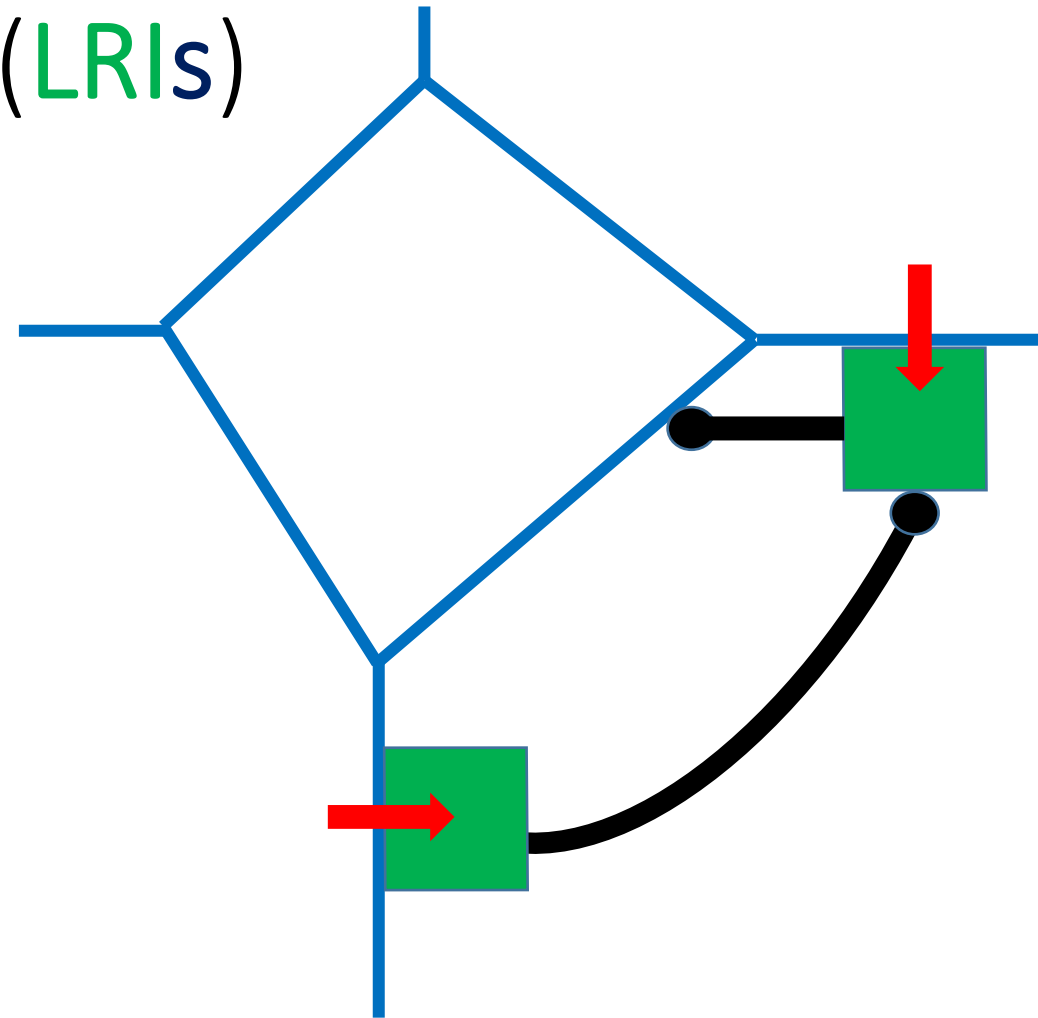
***Theorems in the math model state that these behaviors function as specified “with high probability”***

***Plus, simulations***

Q: How are areas inhibited/disinhibited?

A: **long range interneurons (LRIs)**

- Populations of inhibitory neurons with long axons
- They can **inhibit** remote brain areas...
- ...or other LRIs
- They can be **recruited** by the assemblies of an area
- **LRI**s seem to be necessary for brain computation [Roux and Buzsaki 2015]
- ***They constitute the program of the computation!***





assemblies + LRIs = computation

A hardware language that can do  
arbitrary  $\sqrt{n/k}$ -bit computation

# The math model as software-implemented neuromorphic computation

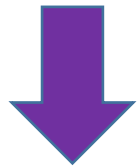
- To simulate  $T$  steps of the model, one needs  **$O(pn^2T)$**  computer time
- (the number of areas is hidden in the  $O$ -notation)
- $T \approx \text{\#seconds} \times 20$
- A ***lazy simulation*** technique reduces this to  **$O(pk^2T^2)$**
- $10^6$  speedup, allows us to simulate ***a few seconds*** of brain time at 3 GHz
- Enough for some cool cognitive phenomena

# Recall: Computation in the brain

- Cognitive phenomena



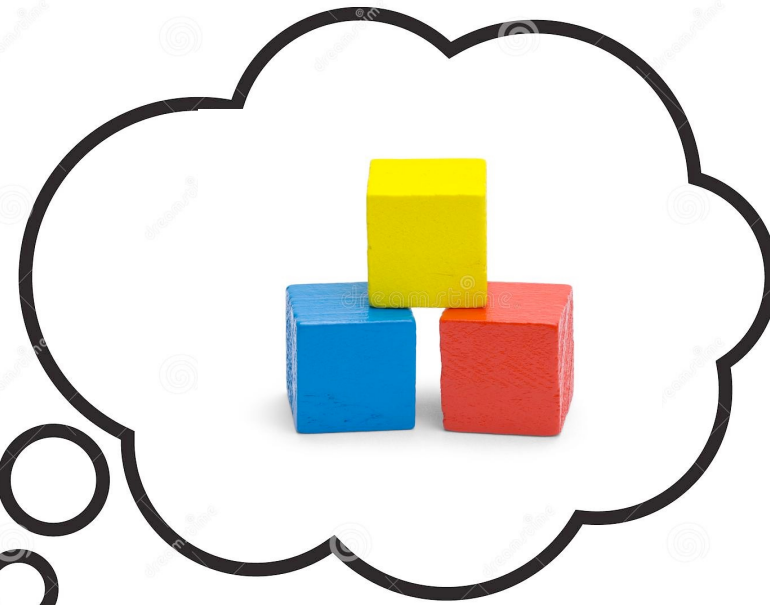
***math model***



**models reasonably well**

spiking neurons and synapses

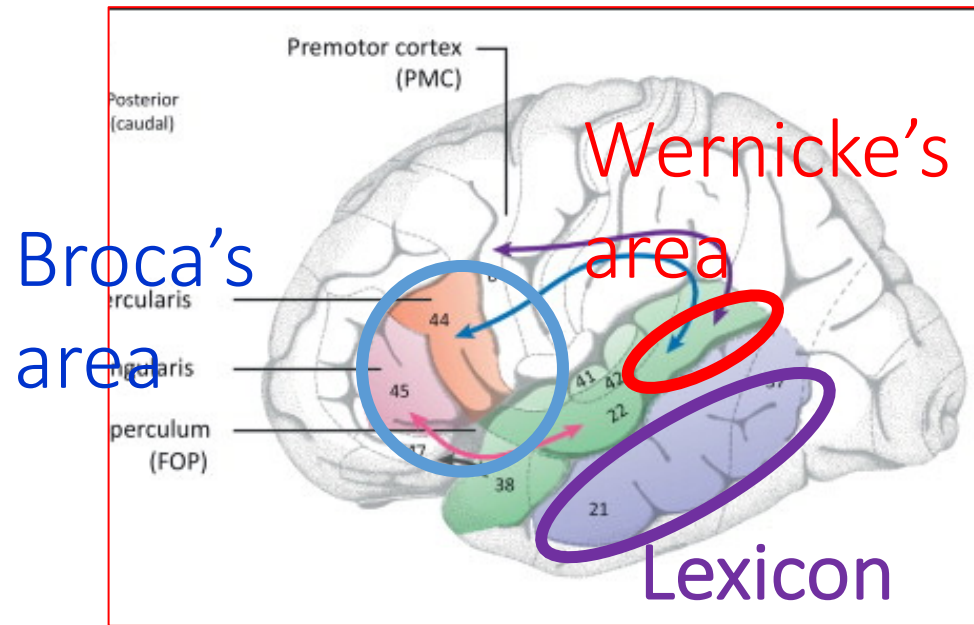
# Planning in the Blocks World (AAAI 2021)



Output:

**put the red block on the table  
next to the blue block;  
then put the yellow  
on top of both**

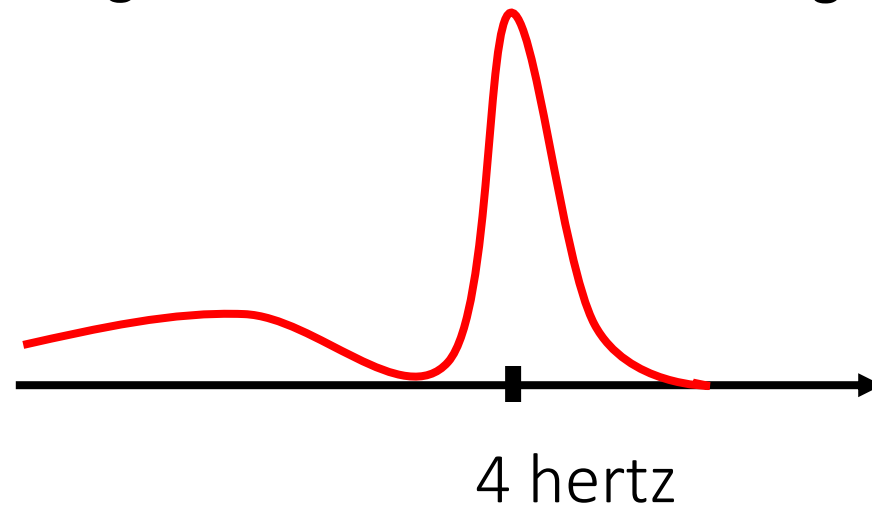
# Language!



It is the hardest thing any brain has done, and so it must hold the key  
But how does it happen in the Brain?

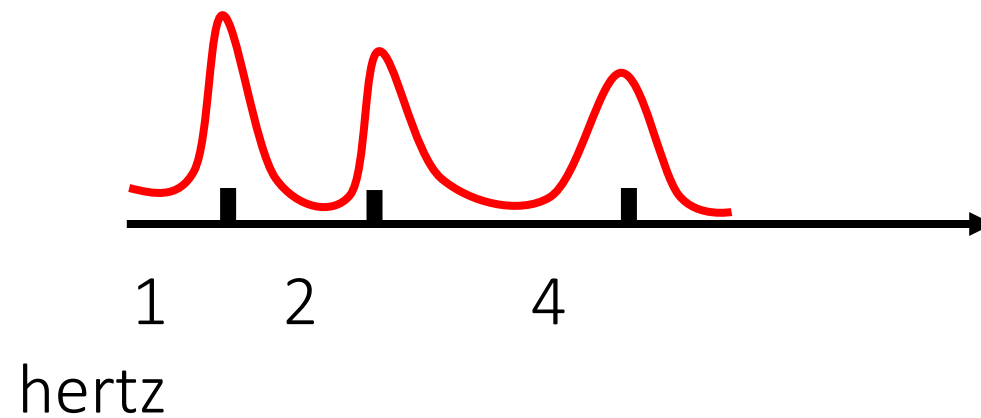
# The [Poeppel 2016] experiment

fret ship hill give true melt fans blue guess hits then cats

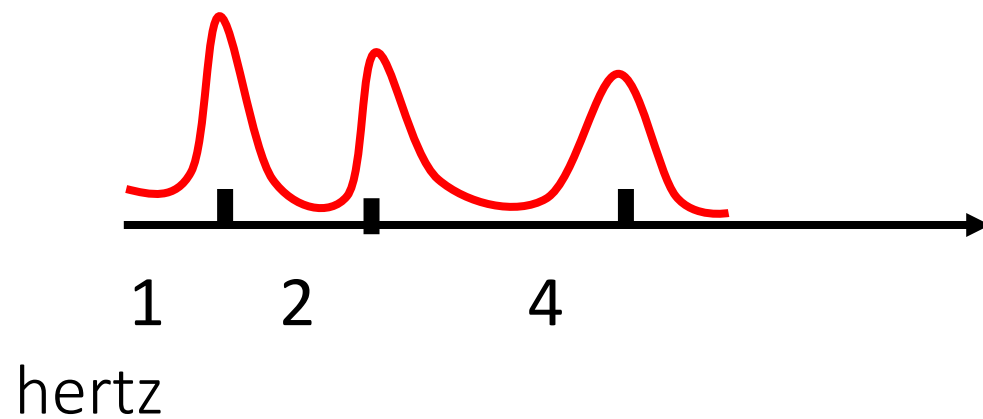
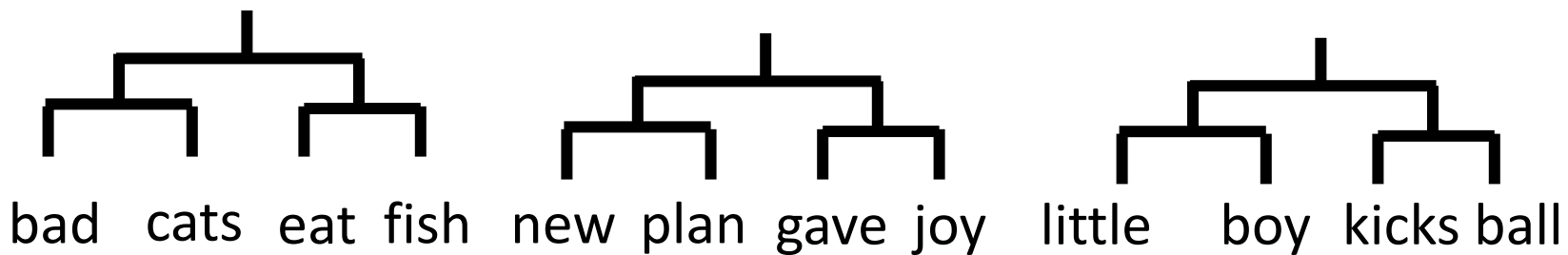


# The [Poeppeel 2016] experiment, stage II

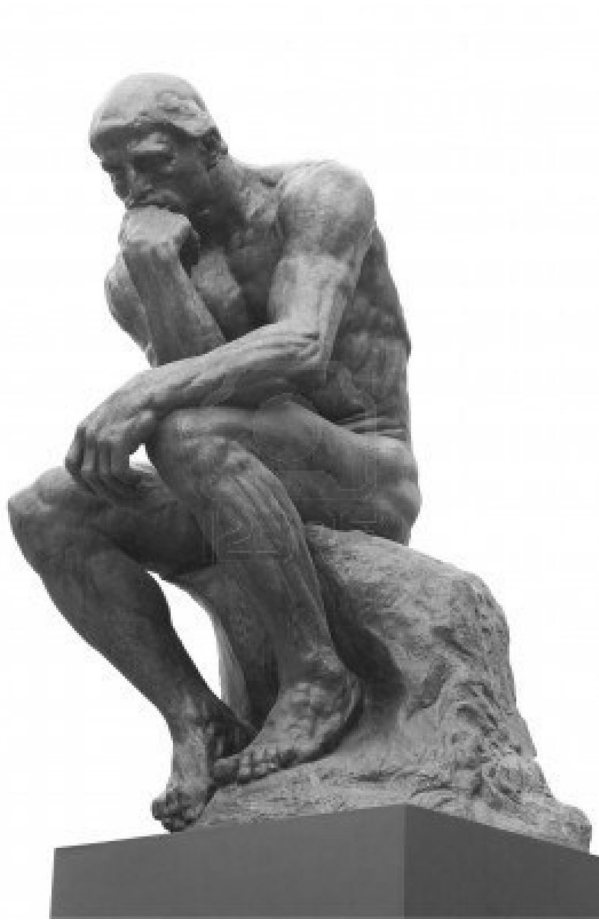
bad cats eat fish new plan gave joy little boy kicks ball



# My interpretation

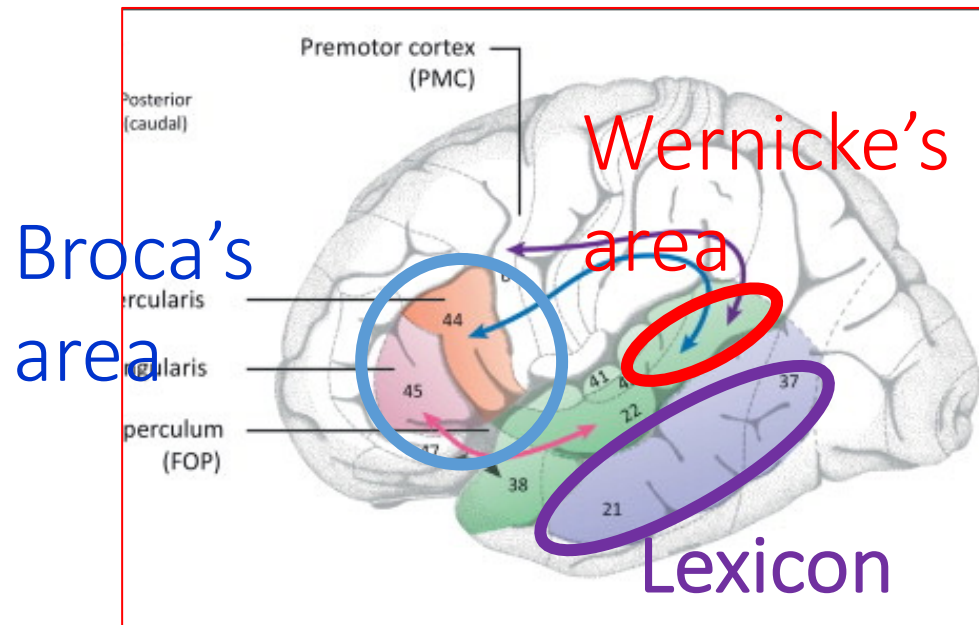






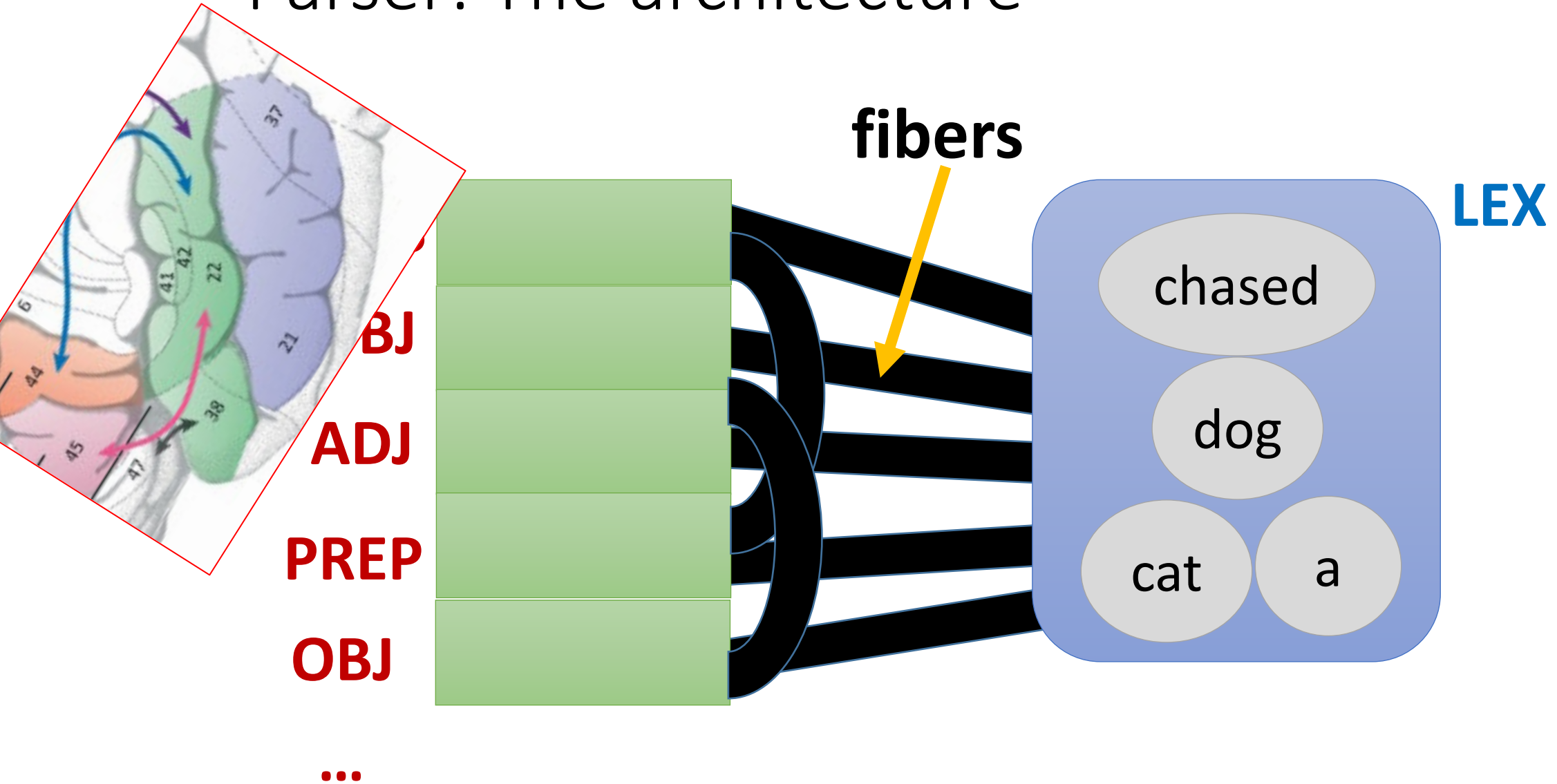
- Hmmm.... how are these tree representations created?  
in about a dozen spikes per step?
- (Why a dozen?  
neuron spikes  $\sim 50\text{Hz}$  :  
language  $\sim 4\text{Hz}$ )
- Can we simulate parsing?

Next:



a parser of English  
implemented by  
spiking neurons

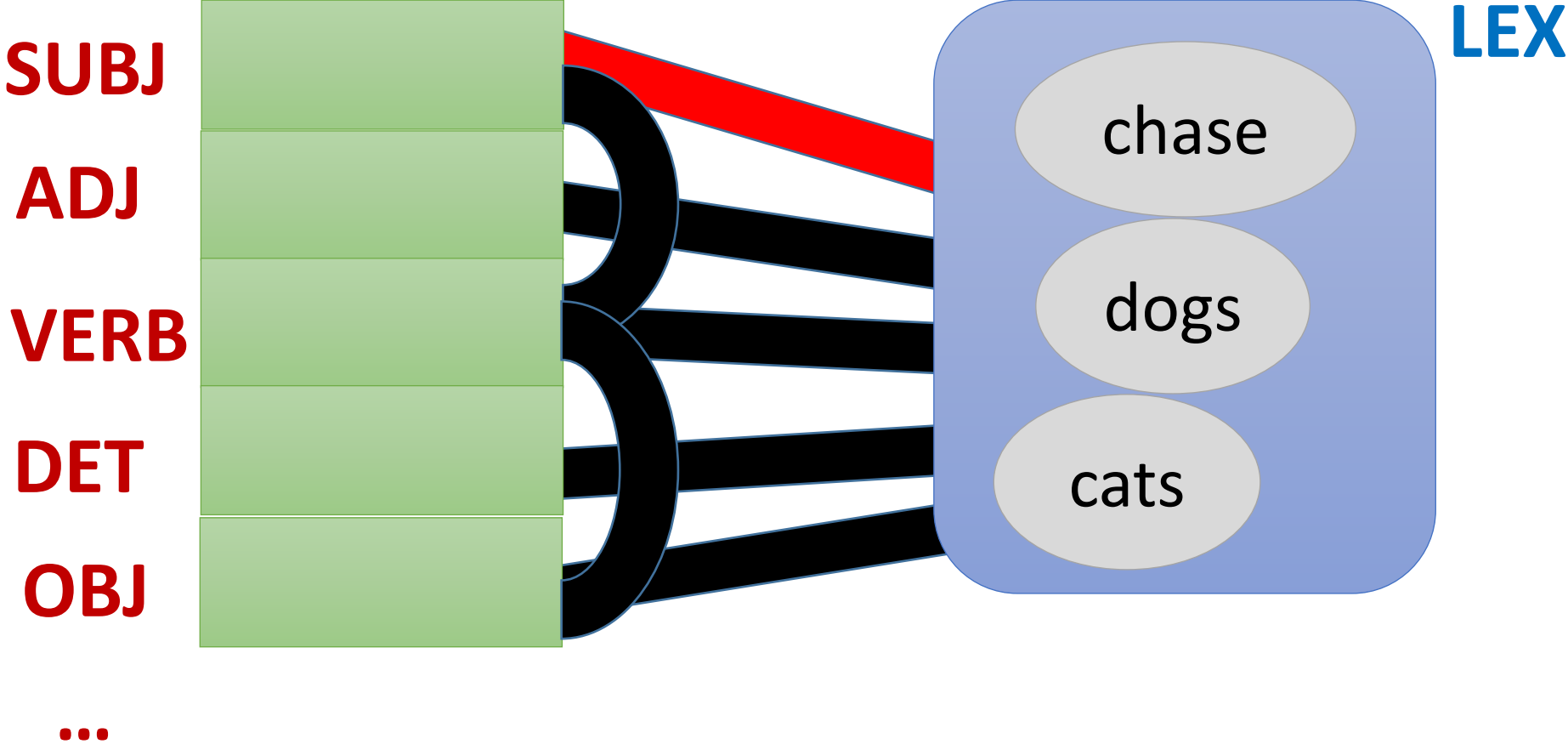
# Parser: The architecture



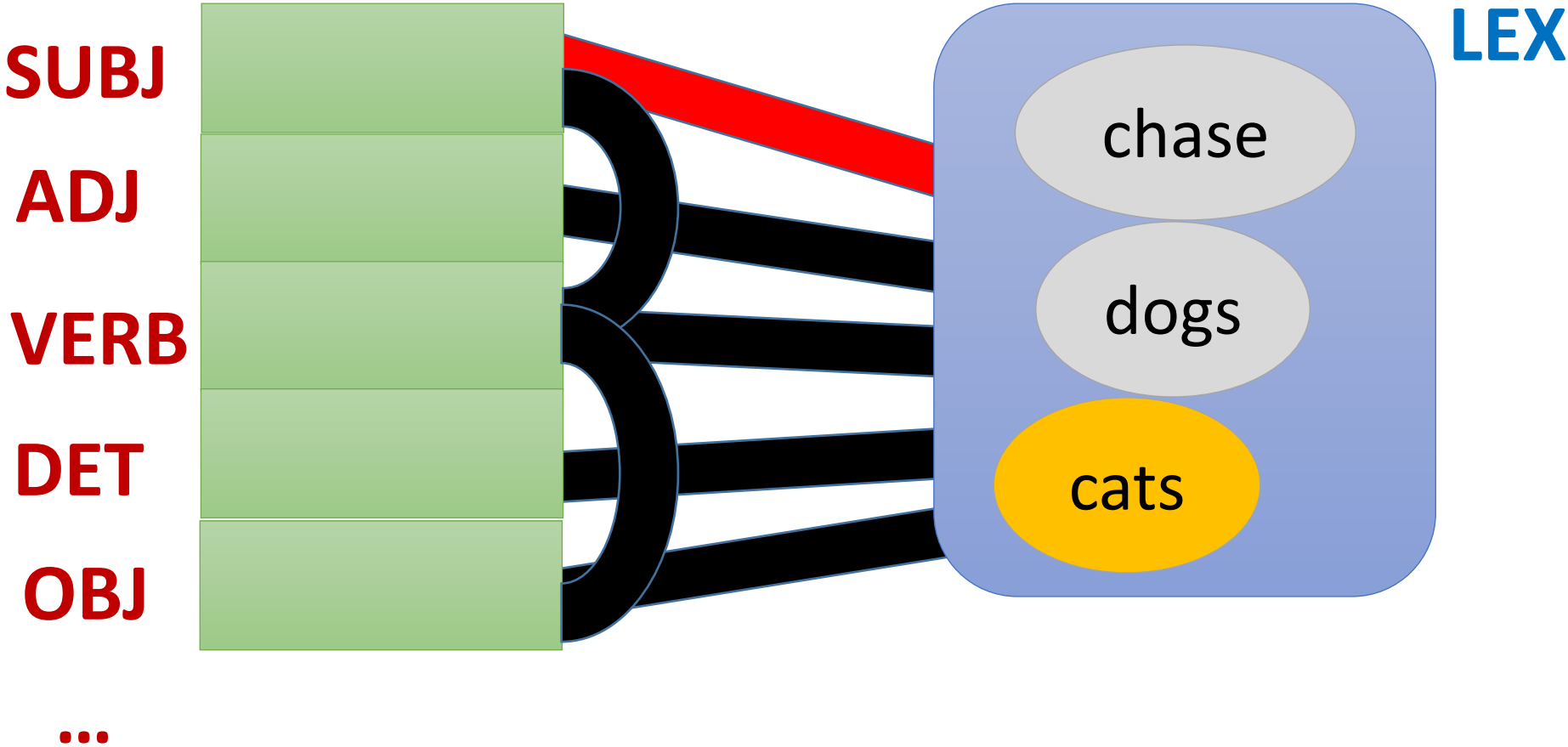
# The Parser

- Input: a sequence of excitations of word assemblies
- (We assume that **phonetics** *has been solved*)
- Each word assembly has an **action set**  
(inhibit/disinhibit actions implemented by **LRIs**)
- Encodes its ***part of speech***, its syntactic role
- When word assembly fires, its action set is ***executed***
- The sum total of the word action sets  $\approx$  ***the grammar***

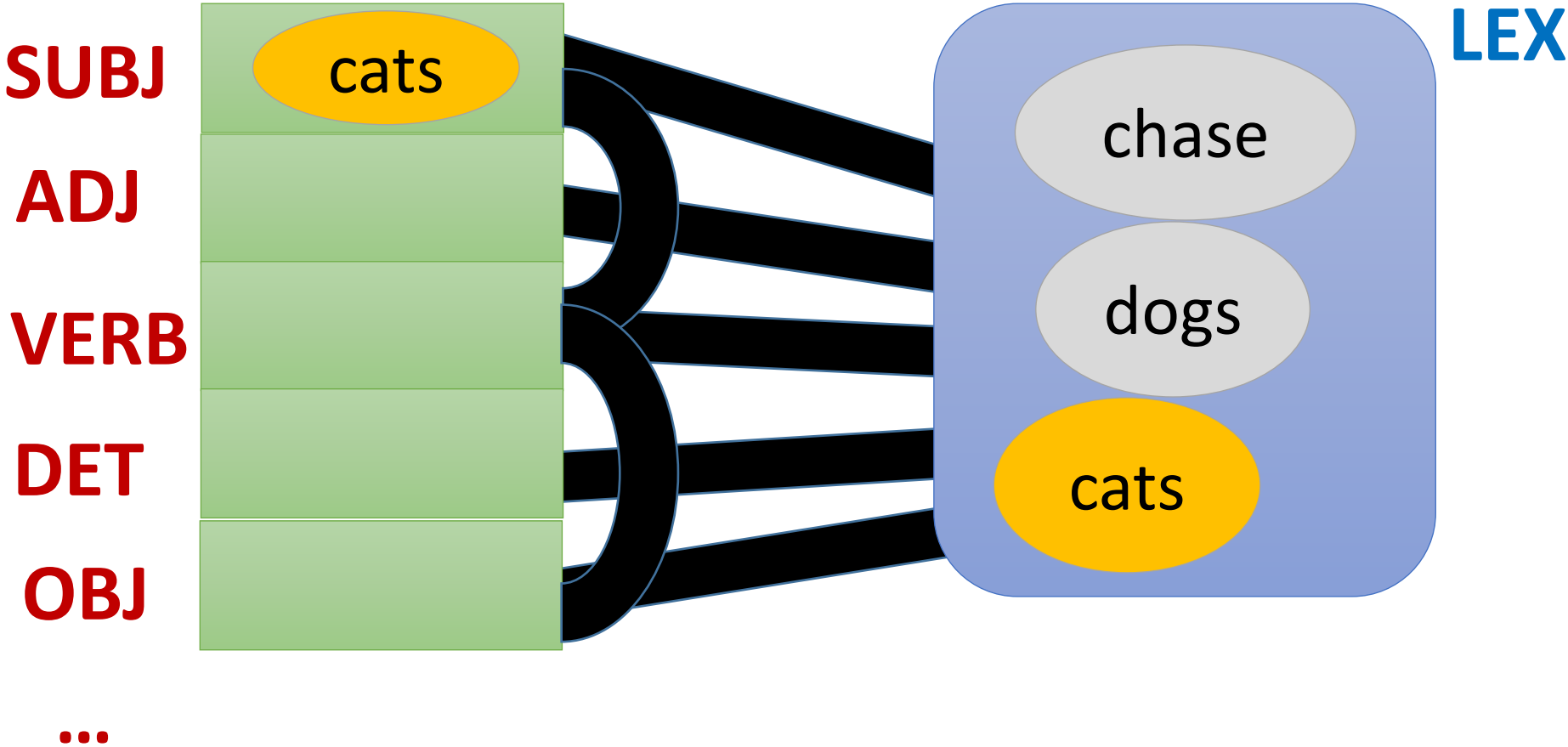
On input cats chase dogs



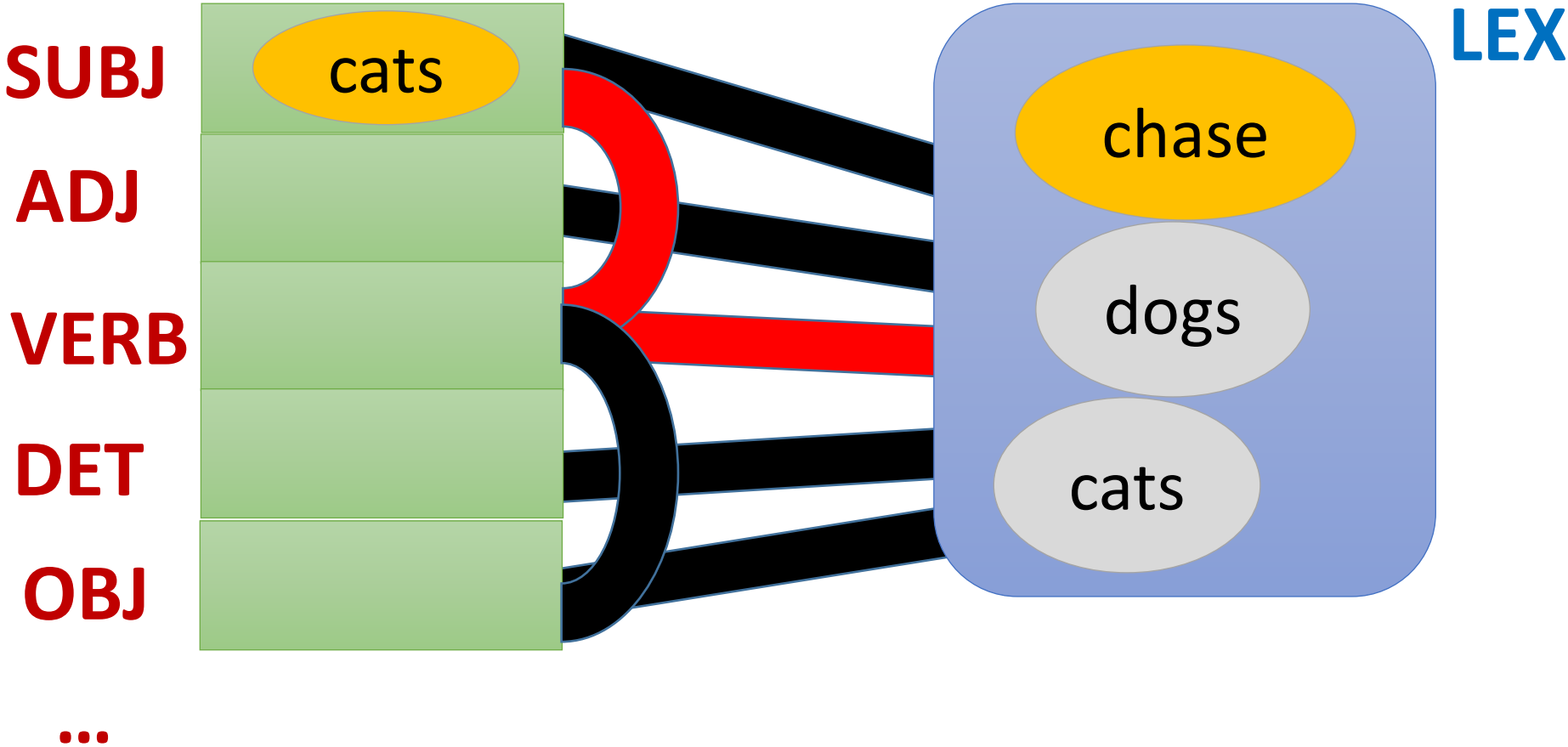
On input **cats** chase dogs



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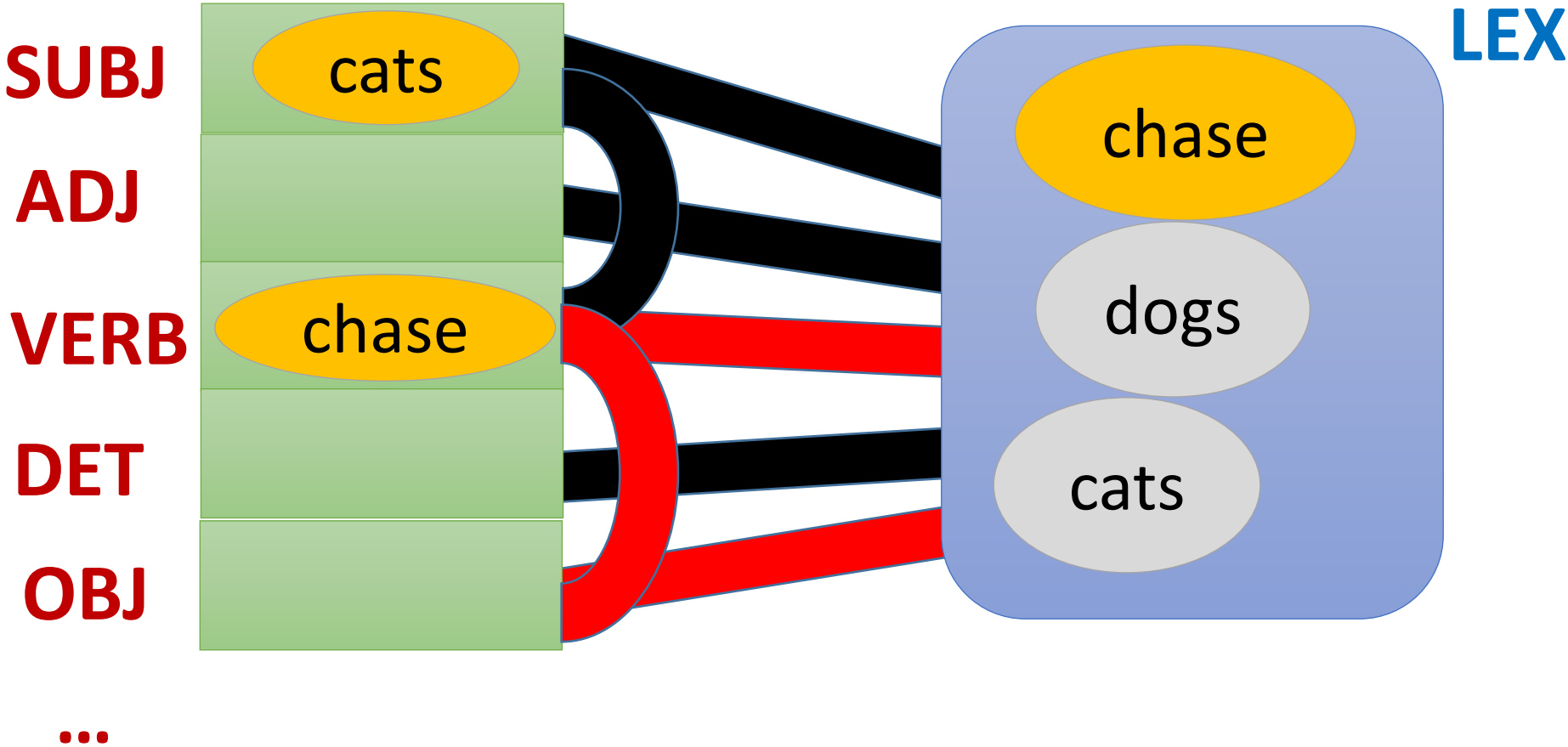


On input cats chase dogs

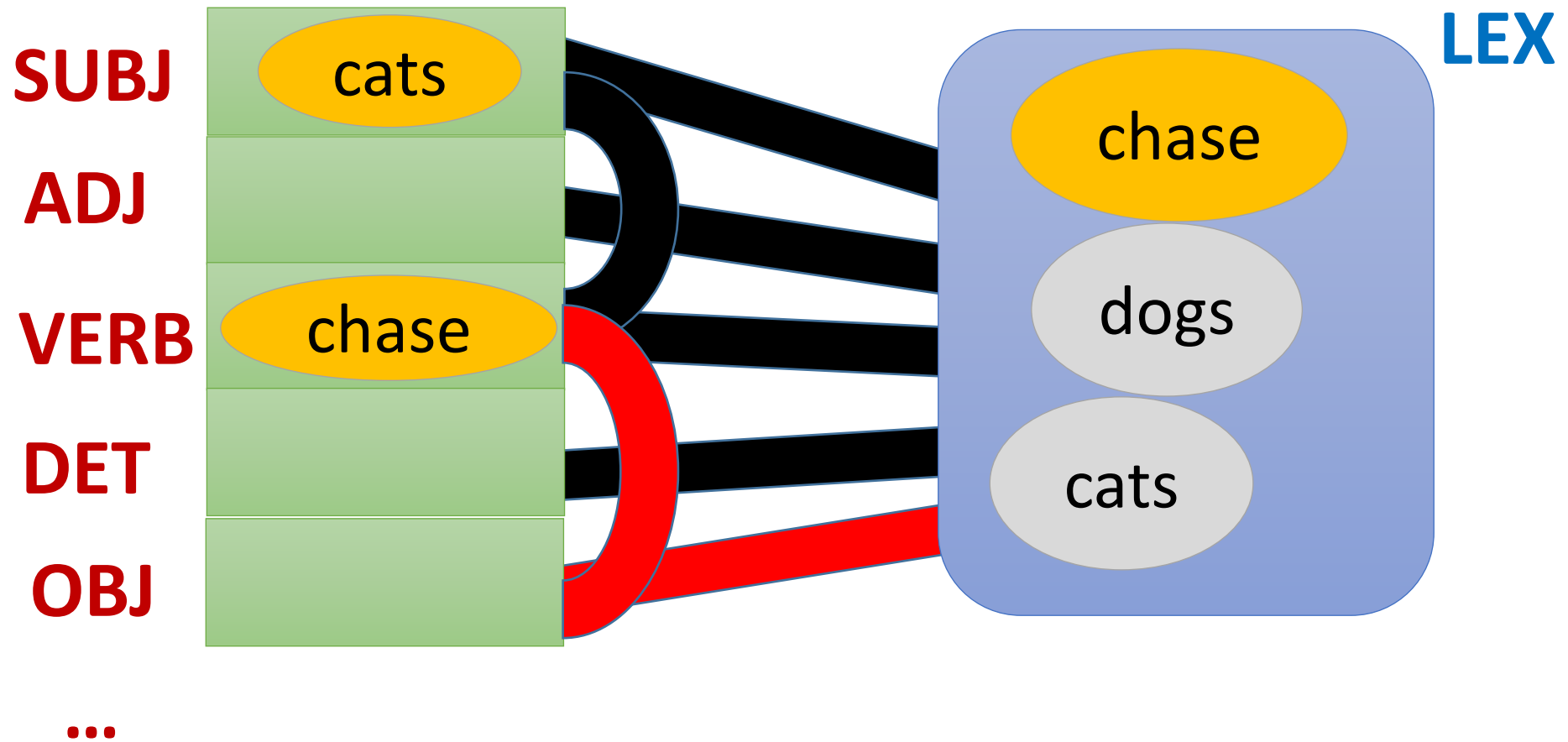




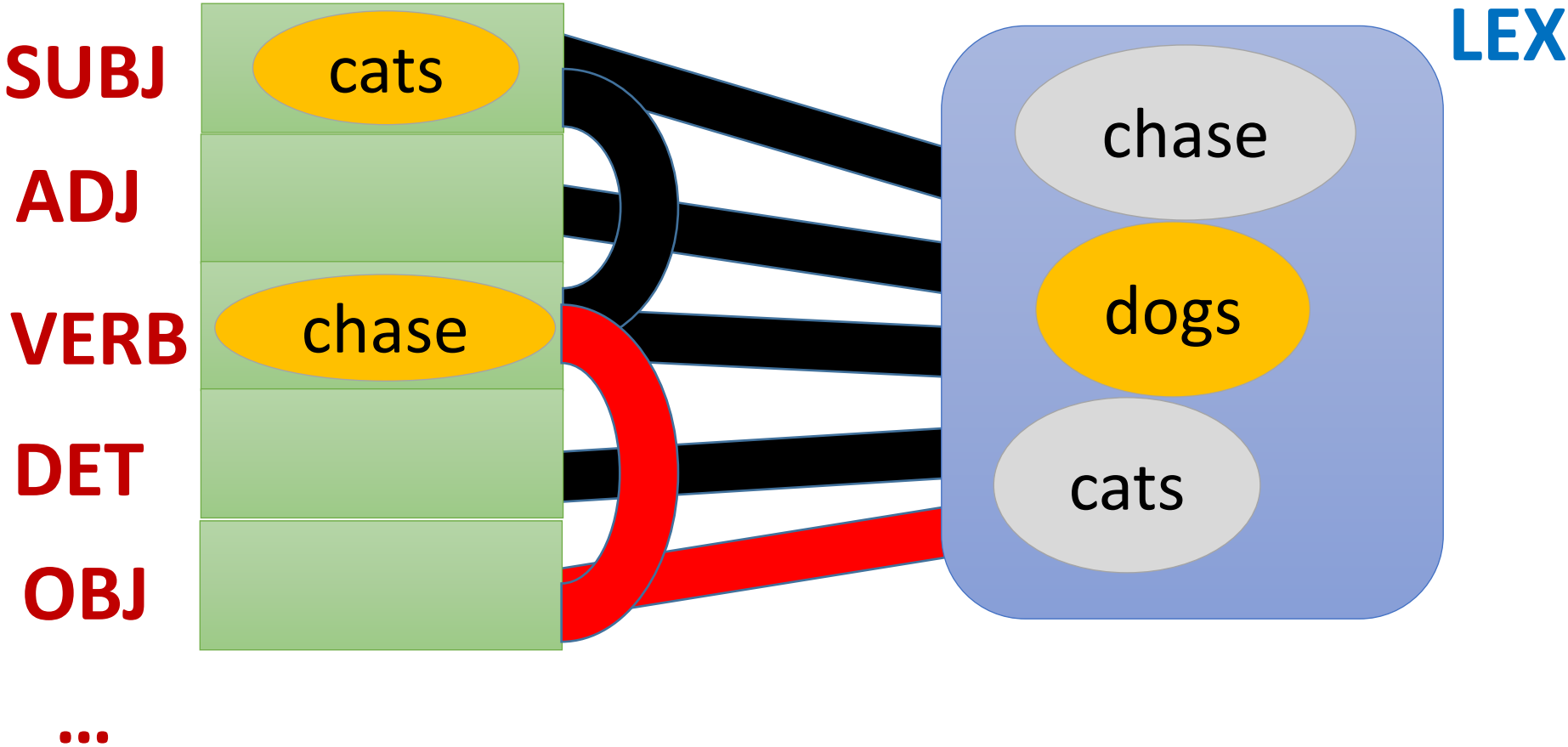
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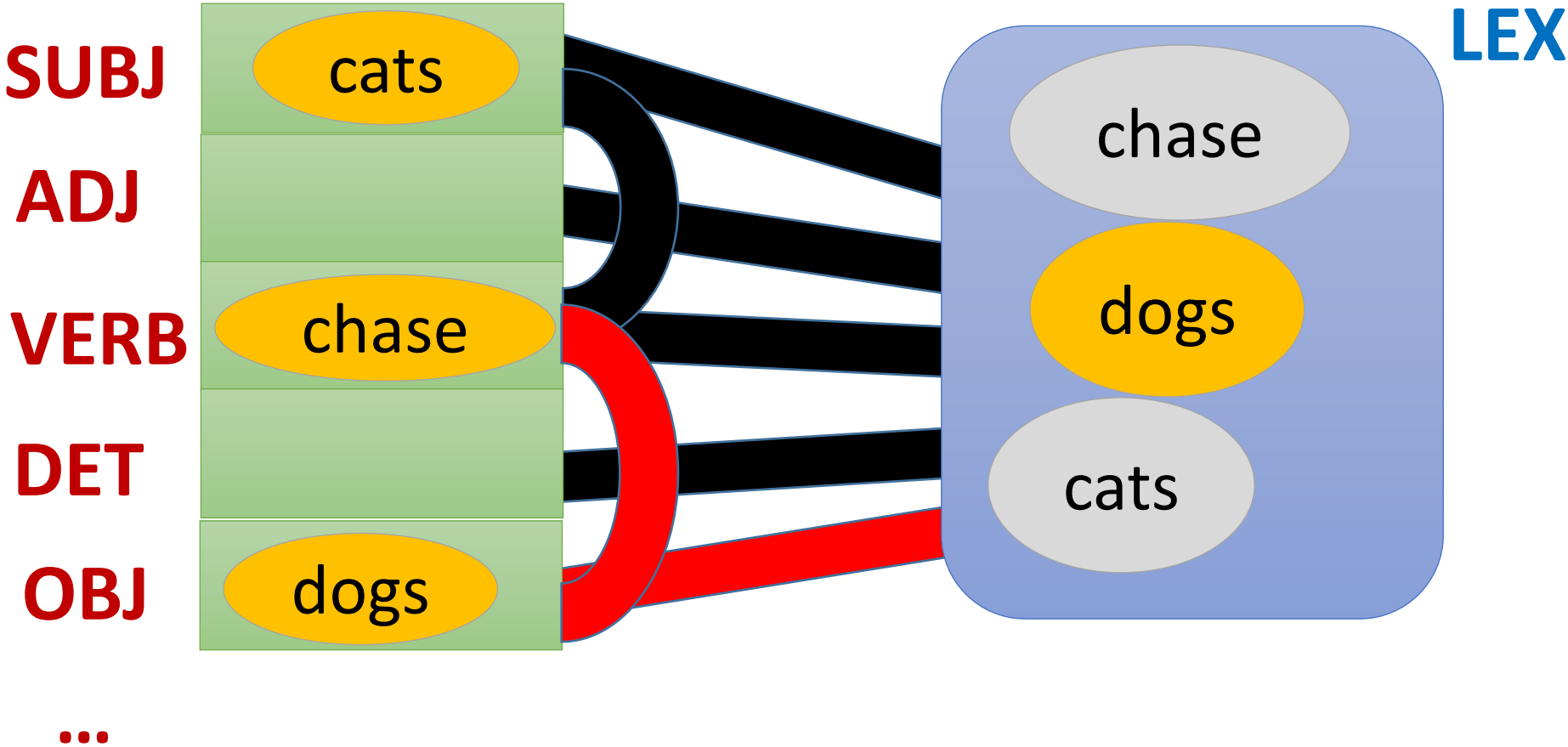
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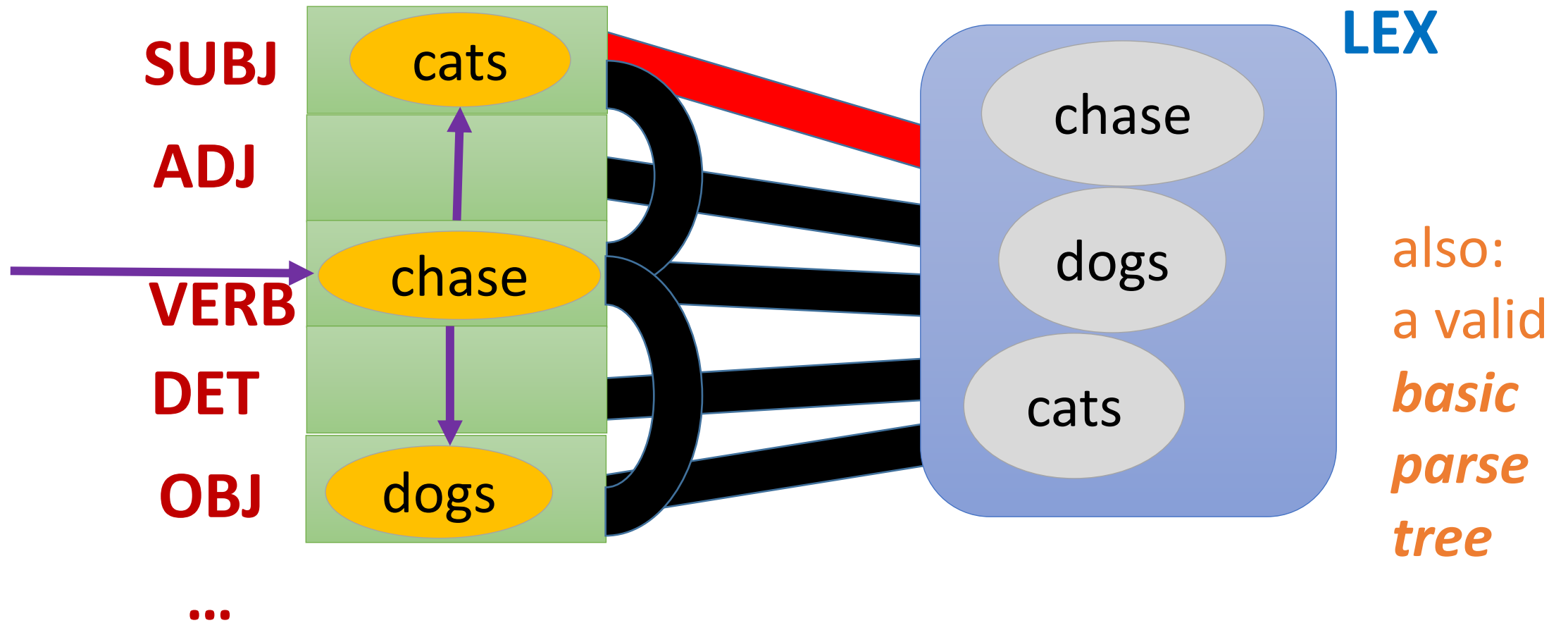


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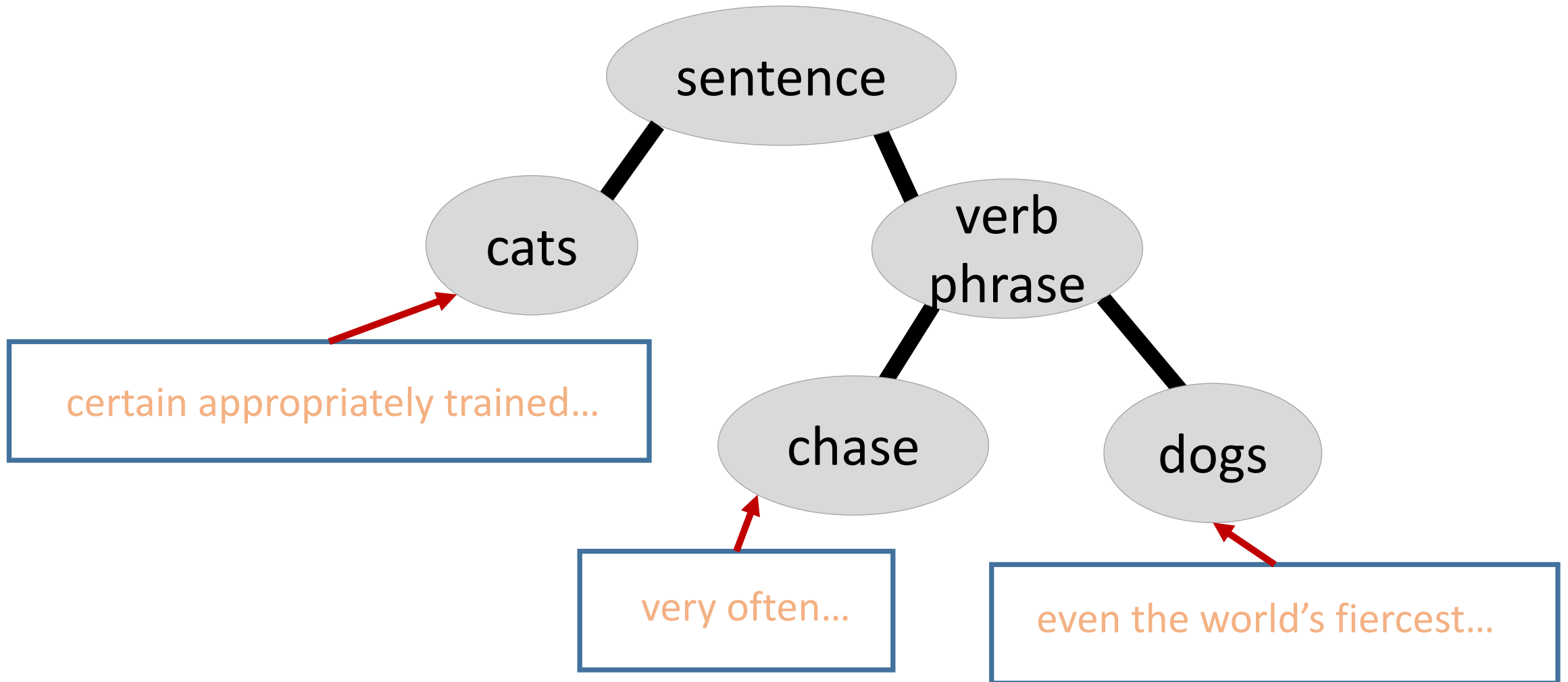


Q: But what does it mean for this device to correctly *parse* a sentence?

After input “cats chase dogs .” high synaptic weights form a *valid dependency tree*



# Basic parse tree



# The Parser as neuromorphic computation

- ***Implemented exclusively through the spikes of stylized neurons***
- $10^{7+}$  neurons, trillions of synapses (cf Intel's Loihi2)



# The Parser

- It parses simple sentences such as:

*“the young couple in the next house saw the old little white car of the main suspect quite clearly”*

- Speed: about **20-25 spikes** (0.4-0.5 sec of brain time) *per word*
- Also running: simple parsers of **Russian, Japanese, Hungarian, Chinese**
- Paper in ACL 2022, code available online

# The Parser

- It parses simple sentences such as:

*“the young couple **who live in the next house** saw the old little white car of the main suspect quite clearly”*

- Speed: about **20-25 spikes** (0.4-0.5 sec of brain time) *per word*
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(A parenthesis on center embedding)

**“cats, when they are fearless, chase dogs”**

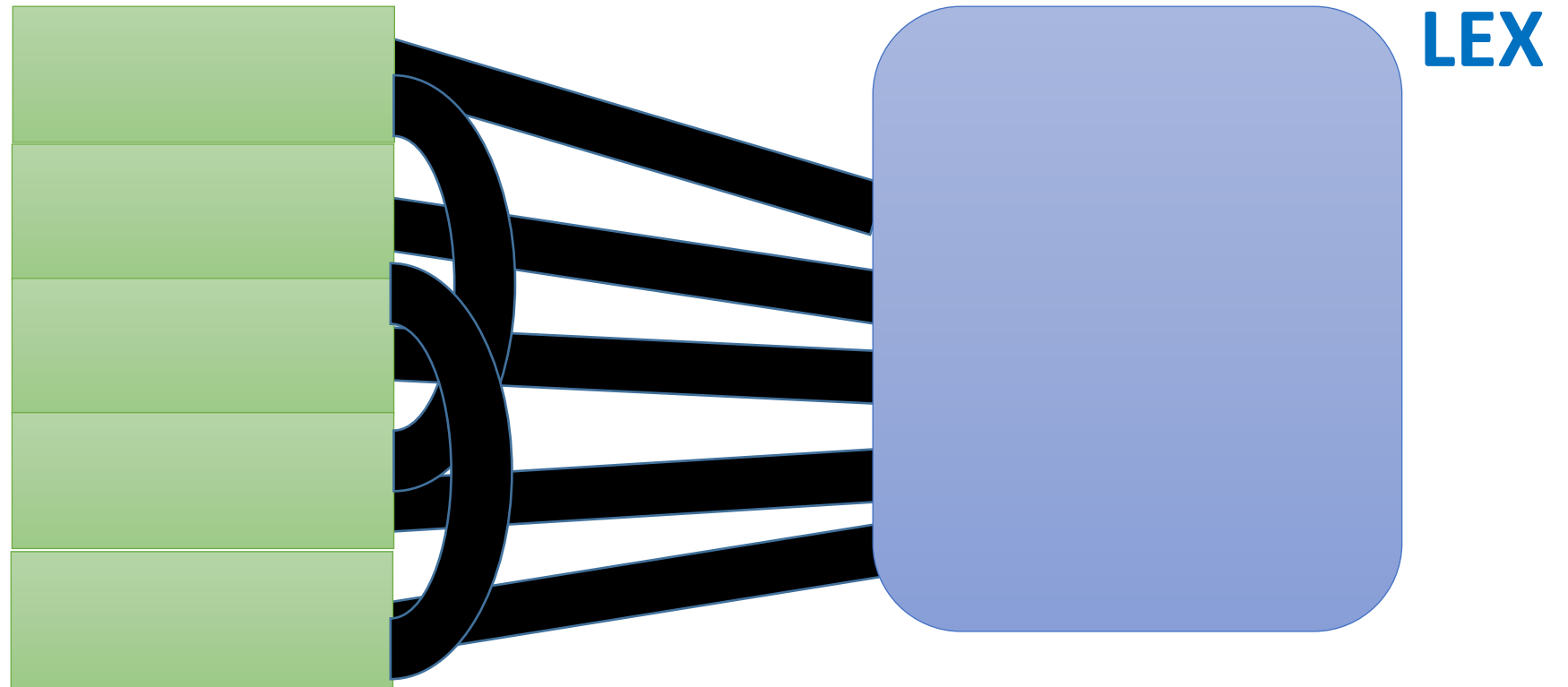
- *What is needed for center recursion?*
- Idea: parse on and return
- Note that, so far, the Parser was a ***finite state machine***
- State = the graph of disinhibited fibers
- Fallback automaton: at any point, it can ***return*** to the most recent ***marked symbol***, unmark it, then skip the ***already seen*** part, and continue parsing (NB: needs sequence recall)

**Theorem:** FBA = CFL [NALOMA workshop 2022]

(On methodology: What are the desiderata for a biologically plausible implementation?)

- That it works, with reasonable parameters and within realistic **brain** time
- That it is in rough agreement with the neurolinguistic **consensus** on the function, if there is one
- Should be plausibly created by **development**
- It is a proof of existence, not a hypothesis or argument that the function happens this **exact** way in the brain
- Multiple **alternative** implementations are a plus

*Q: so, is this the neural basis of language?*





A: not really, because  
*the hard part is filling the blanks!*

**VERB**

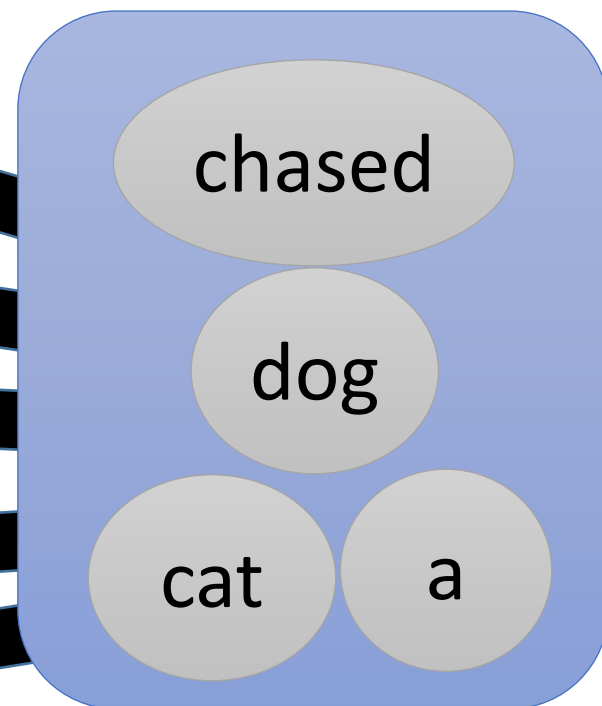
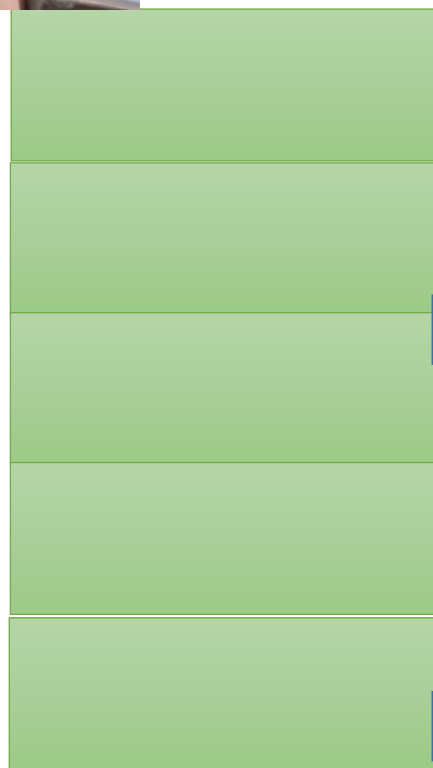
**SUBJ**

**ADJ**

**PREP**

**OBJ**

...



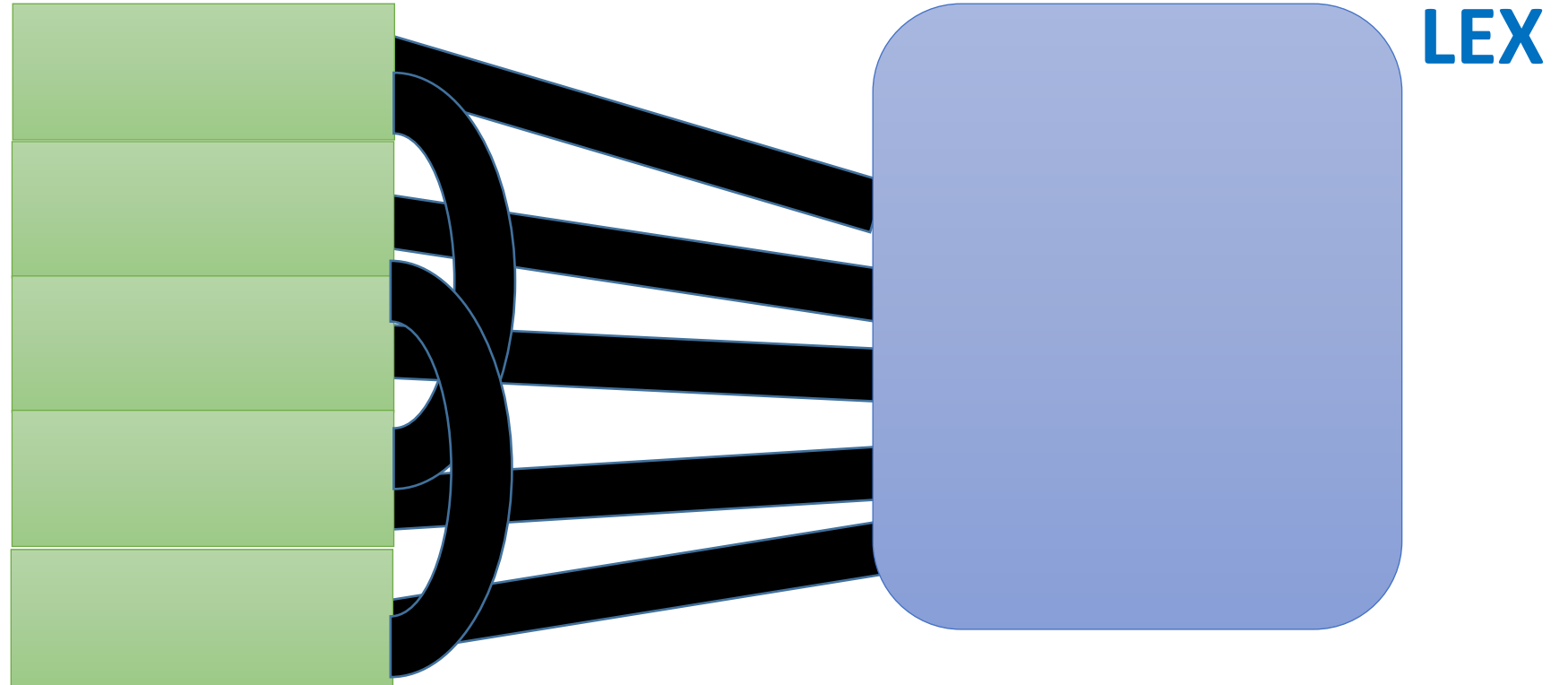
**LEX**

And how about *semantics*  
and *generation*?

# The neural basis of language?

This,  
plus hardware for  
*semantics* and  
*generation...*

*...plus the ability  
to learn all these  
things...*



Current work: Biologically plausible language acquisition

**Hardware:** a tabula rasa of a couple of dozen brain areas, fibers, and populations of LRIs

**Input:** a modest amount of *grounded, shared-attention* language

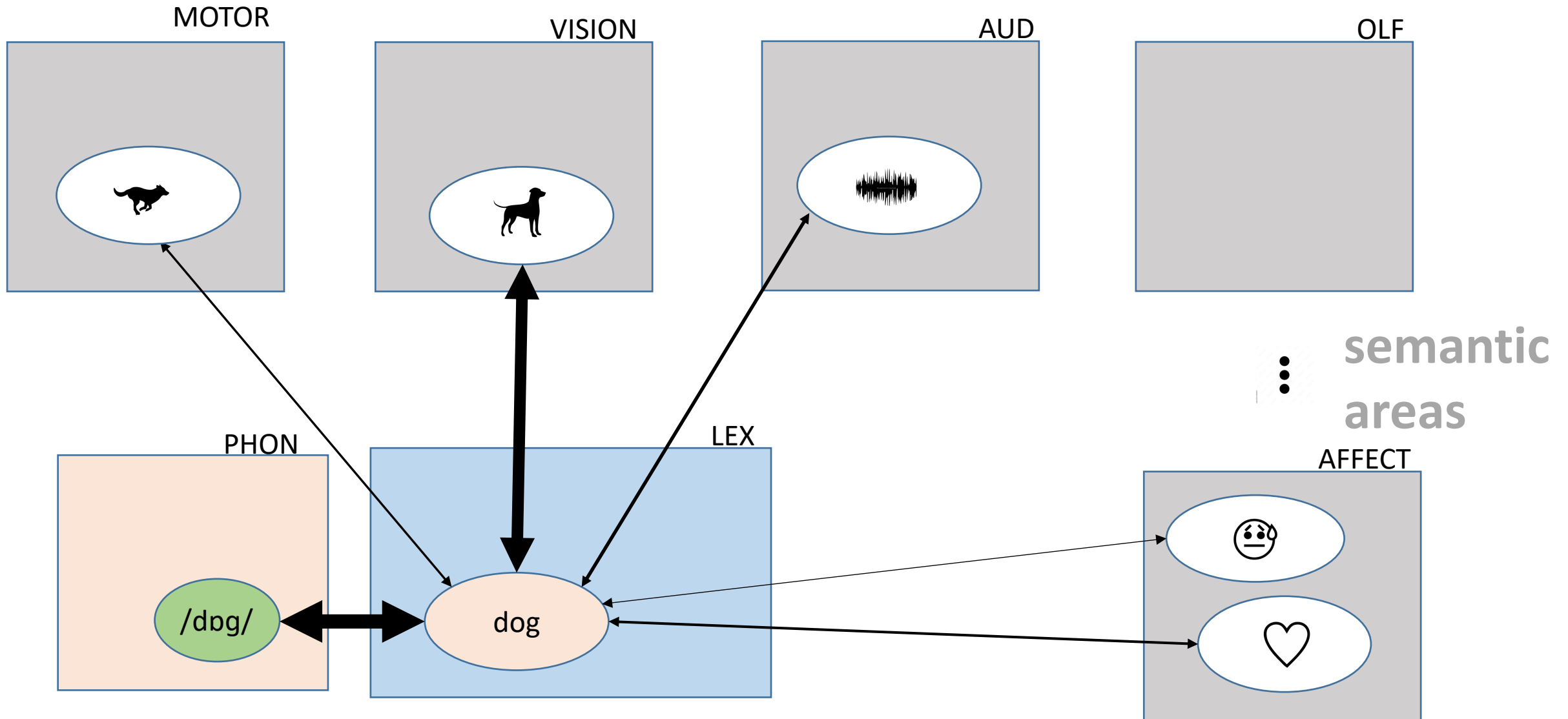
**Output:** a mature language organ, complete with a parser and a generator



# Biologically plausible language acquisition: what we have implemented and tested so far

- Learning of nouns
- (We assume, as with the Parser, that *phonetics* has been solved)
- Representation of noun semantics
- Representation of verbs and their semantics
- Learning to generate two-word sentences
- Learning the word order of the language (two alternative ways)
- The **hard case**: learning syntactic roles of words from overheard language – still grounded and with shared attention.

# Representation of nouns and their semantics



# Representation of word semantics (cont.)

- Implements the “hub and spokes” semantics theory in neurolinguistics
- Each word representation consists of a *star of assemblies*
- *Not unlike word2vec*
- The assemblies, their overlap, and the strengths of their connections are all **dynamic**
- They reflect cooccurrences of words, statistical regularities, order statistics, and changing world contexts

# Representation of word semantics (cont.)

- In fact, *“man is to woman as king is to what?”* is answered ***automatically***
- “man” in LEX is connected with assemblies “human” and “male” in the semantic areas, “queen” is connected with “royal” and “female”
- *“I saw the tree with my binoculars”* will be disambiguated correctly
- “see” and “binoculars” will have strong overlap in the semantic areas, and as a result the prepositional phrase will synapse preferentially with the verb

...and of course,  
no matter how clever you are,  
you could be doing  
the wrong thing...



Soooooo...

- The study of the brain is fascinating and bottomless
- Language is the hardest thing any brain has done, and so it must hold the key
- Implementing language in the brain entails close encounters with profound questions
- Progress on biologically plausible acquisition
- **A promising way forward in AI?**
- **Are assemblies and long-range interneurons the seat of Axel's "logic"?**

# my collaborators!



Santosh  
Vempala  
GaTech



Max Dabagia  
GaTech



Mike Collins  
Google



Pilu Crescenzi  
Gran Sasso



Dan  
Mitropolsky  
Columbia

Merci!

PASAJE GALVEZ

