# USING NENGO.AI FOR DEVELOPING MODELS OF SPEECH PRODUCTION AND SPEECH PERCEPTION

Prof. Dr. Bernd J. Kröger Neurophonetics Group, Phoniatrics Department, Medical Faculty RWTH-Aachen University, Aachen, Germany <u>www.speechtrainer.eu</u>

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#### Introduction

- How can experimental sciences benefit from models?
- The NENGO.ai framework
- Our speech processing model (production and perception)
  - architecture of the modeled brain
  - simulated tasks -> simulated behavior (modeled test person)
- Concluding remarks
  - On the realism of brain models
  - On the benefit of brain models for medical research

# Motivation for models

- Models do not necessarily lead to new (detailed) knowledge
  - -> ... can not answer too specific (quantitative) questions, like: How strong is a specific effect?
- But: Models help to understand the basics
  - Functional organization of the neural processes of speech production and speech perception
  - Interactions between different processing stages (modules or levels of a model)
  - ... can answer the question: an effect occurs because of ...

#### Motivation for models

- Models can generate new knowledge:
  - e.g.,: relationship between neural structure and behavior:
  - i.e.,: how does a neural dysfunction change behavior (-> speech)?
  - Thus: Neural models may help to refine medical screening methods or to develop new screening methods

### Motivation for models

- Neural models
  - define structure (neurons and connections)
  - define neural functioning (spiking, forwarding and spreading of neural activity) -> neural processing
  - are able to simulate macroscopic behavior (in case of large-scale models = a avatar)
  - give clear definitions for neural dysfunctions (resulting in specific speech disorders)

# The NENGO.ai framework

- "Neural ENGineering Objects" framework:
- <u>www.nengo.ai</u>
- Nengo is free -> open source, but professionally developed!
- Python based: -> runs on: Windows, Mac OS X and Linux
- Embedded in Python Development Environments (e.g., Anaconda)
- Python Math-Libraries, plotting libraries (-> comparable to MatLab) can be used
- Professional scientific background:
  - Group of Prof. Chris Eliasmith, Leader of Centre of Theoretical Neurosciences, University of Waterloo, Waterloo, Canada
  - Eliasmith, C. (2013). How to build a brain: A neural architecture for biological cognition. Oxford University Press.

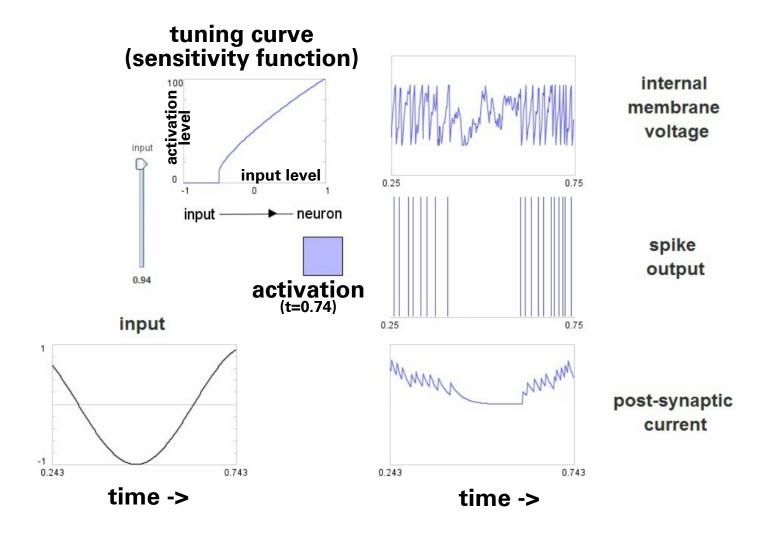
# **NENGO** comprises

- basic neuron model for cortex, for subcortical regions (LIF-model)
- concepts or strategies for connecting neurons:
  - neuron ensembles, neuron buffers, short-term memories, associative memories, binding buffers, ...
- strategies for coding higher level items (concepts <-> S-pointers)
- fully developed model for for neural process control over time (BG-Thalamus-complex: cortico-cortical control loop)
- strategy for connecting the periphery: -> model avatar
  - Sensory input: eyes and/or ears
  - Motor output: arms and/or the speech articulators

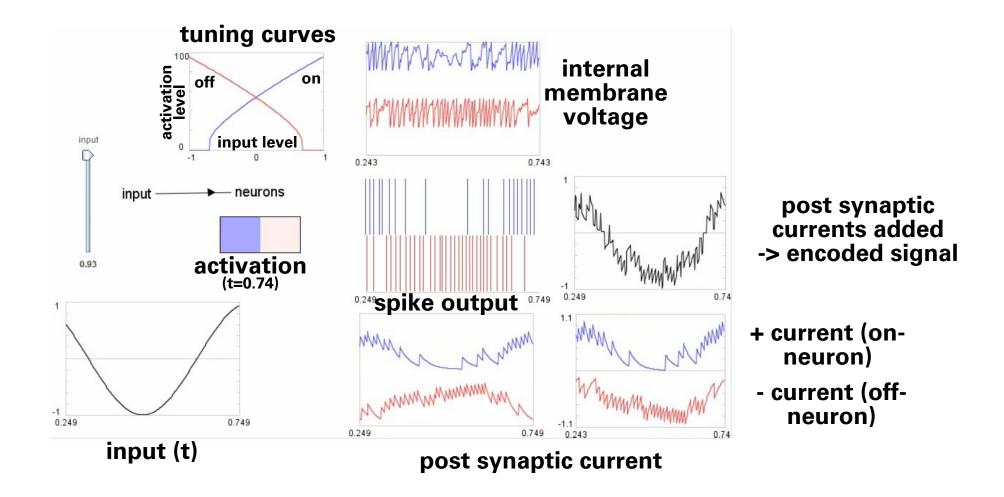
### NENGO: single neuron and spikes

- Neuron = Information processing unit:
- Presynaptic activity (potential) triggers postsynaptic activity (current, based on number of spikes per time interval)
- ... triggers activity level of each neuron cell (soma)
- First goal: representing an input value range by neural activity
- -> we need more than one neuron!

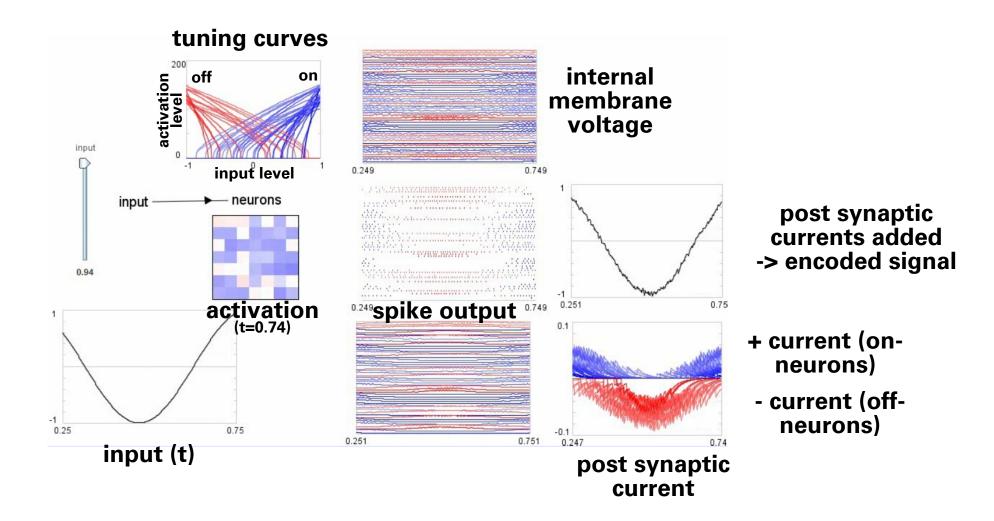
#### Example: one neuron, sensory input



#### Example: two neurons

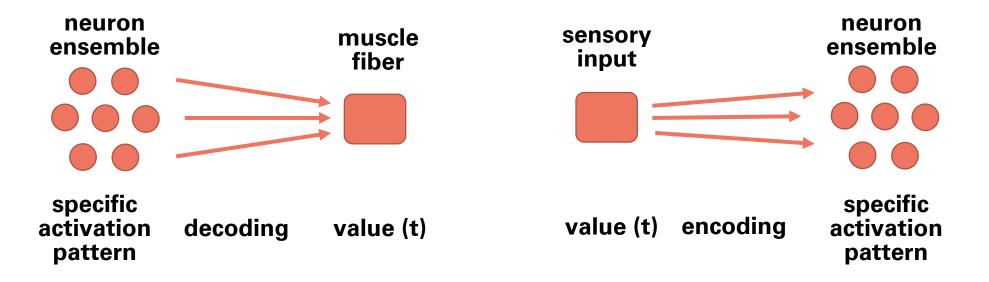


#### Example: 50 neurons



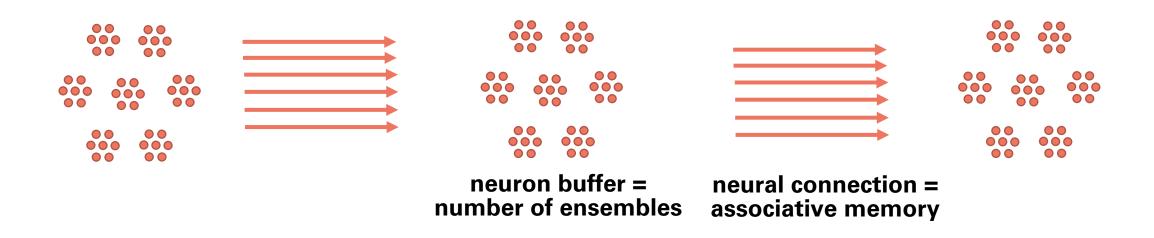
#### 50-100 neurons -> neuron ensemble

- represent "values" (e.g., a muscular activation level; a specific sensory input intensity level over time)
- each value (winthin the range) is
  - represented by specific neural activity within the ensemble
  - en-/decoded by synaptic weights (connections between nenurons)



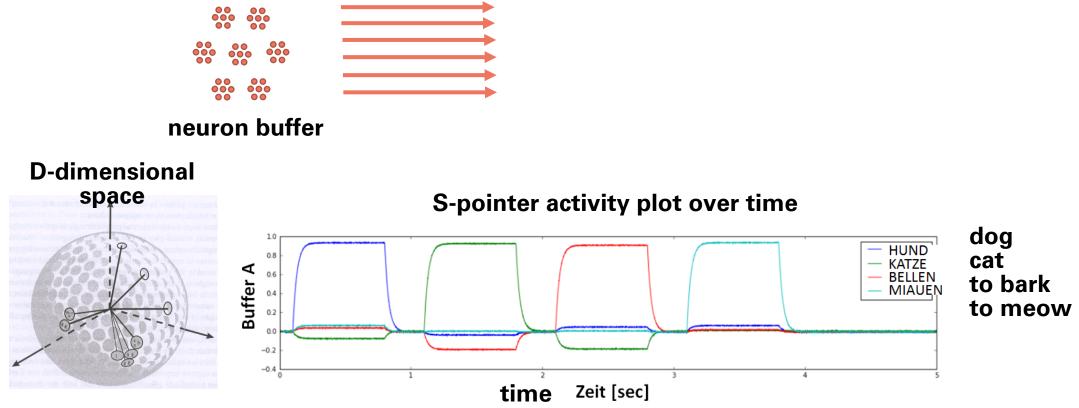
#### 50-100 ensembles -> neuron buffer

- SPA (Semantic Pointer Architecture as part of NENGO)
- buffer represents complex information (items)
- -> coded by a number of "values"; mathematically represented by "vectors" (S-pointers)
- S-pointers represent:
  - phonetic or phonological items (sounds, syllables words)
  - lexemes (language-specific)
  - concepts (meanings) ((-> "thoughts"))



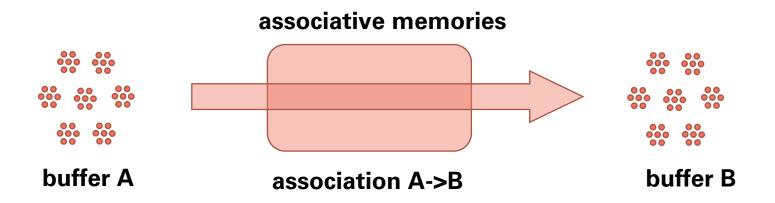
#### NENGO: Decoding complex info (S-pointers)

- by: S-pointer activity plots "similarity plots" for each neuron buffer
- only the most activated items within a buffer are shown (similar items are co-activated)



#### **NENGO:** associative memories

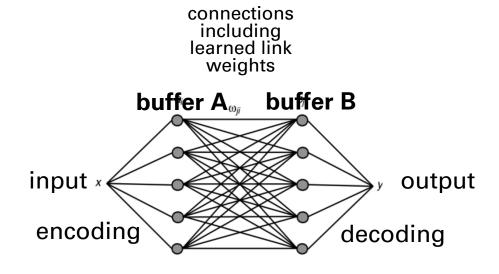
- define the associations of items activated in different buffers: e.g., from phonological form /dOg/ -> lexeme "dog"-> concept <dog/chien/Hund>
- neural connections between different buffers
- develop by learning -> adjustment of synaptic link weights



#### **Neural associations**



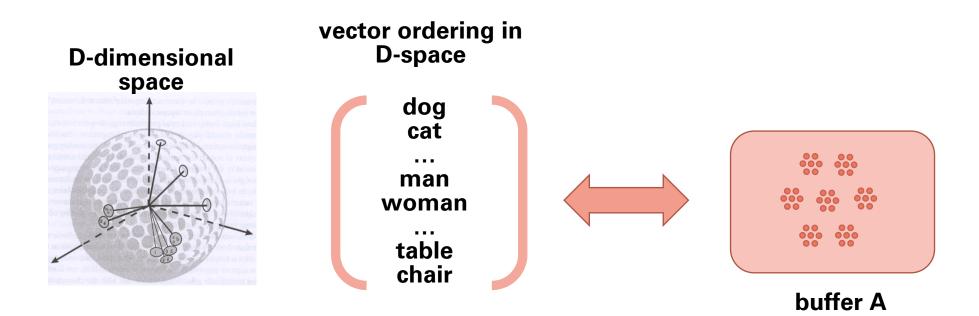
connect each neuron with each other neuron



transformation of states by neural associations

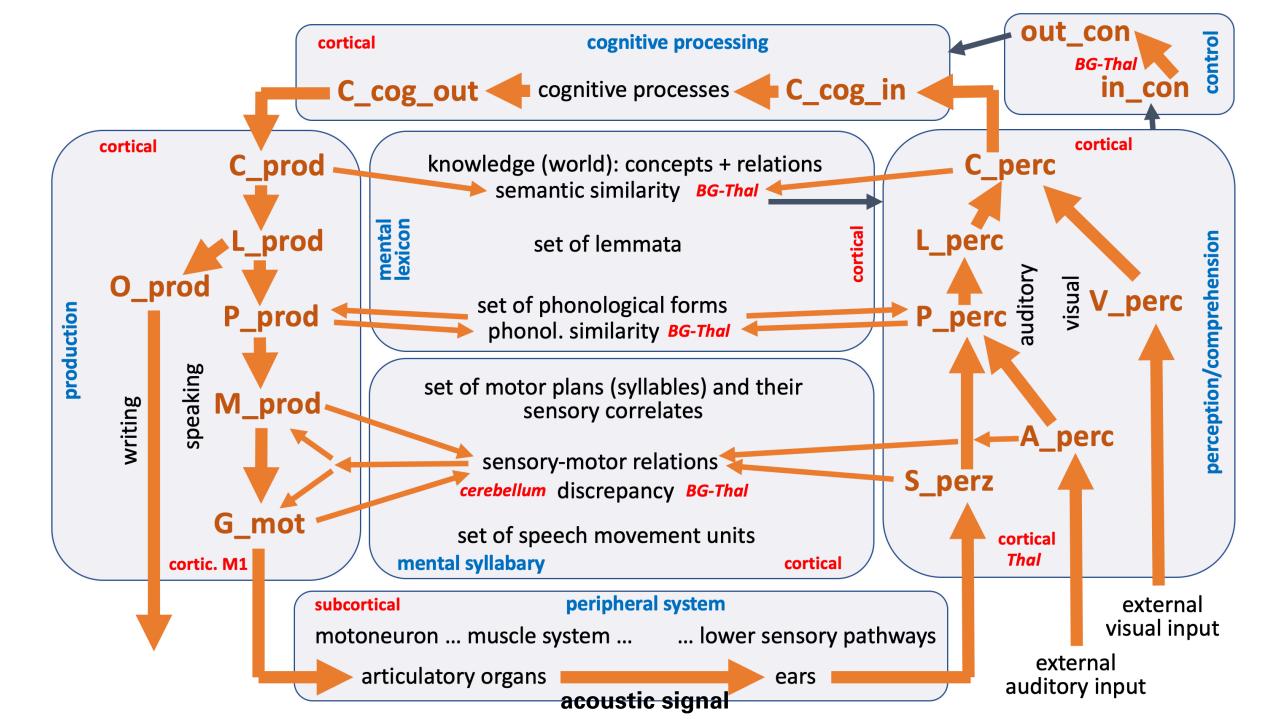
#### NENGO: S-pointer-networks

- define the associations of items within a buffer: e.g.,:
  - at concept level: <dog> <cat> -> animals; <chair> -> furniture
  - at phonological level: /car/ /cat/ -> begin with /k/; /far/ /fat/ -> with /f/
- similar S-pointers point in a similar direction:



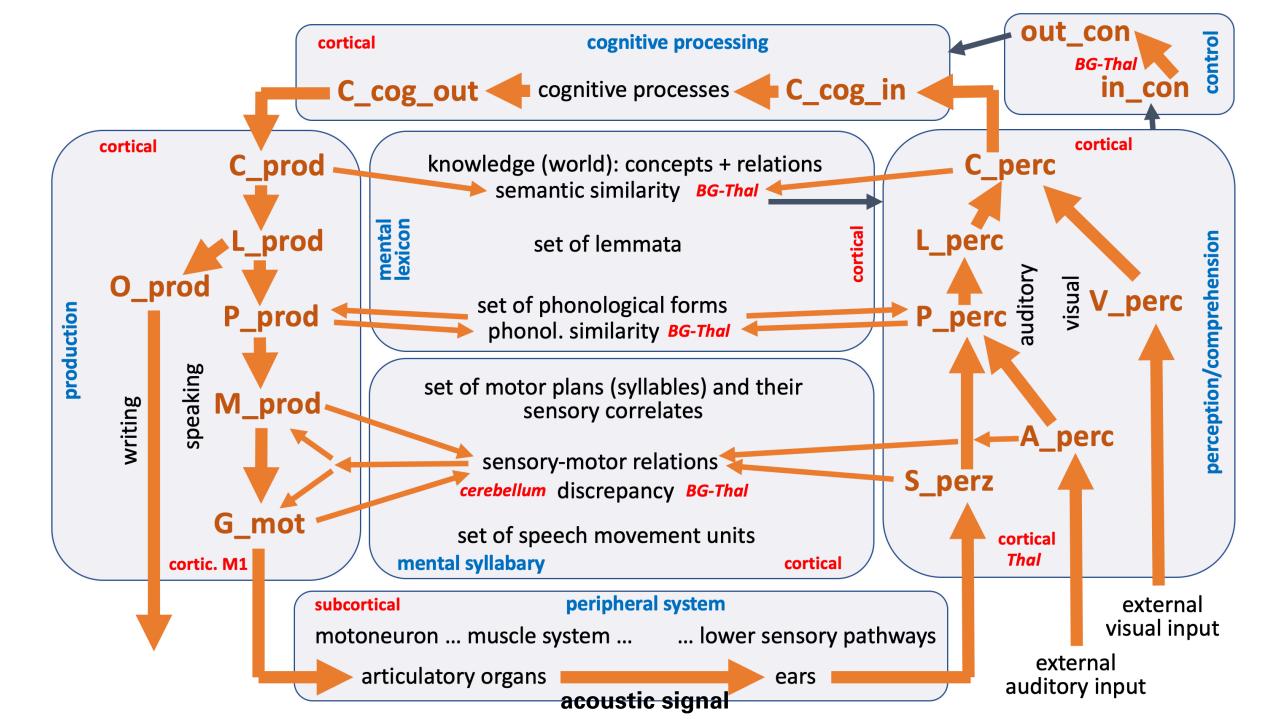
#### Speech processing model: Architecture

- The complete large-scale model
- Perception pathway: Auditory input (values, processed in neuron ensembles) -> cognition (understanding; comprehension) (( buf -> assoc buf -> buf -> assoc buf ... ))
- Production pathway: message, word (S-pointer, processed in neuron buffers) -> articulatory-acoustic output (( same ))
- Including: mental lexicon, mental syllabary as skill or knowledge repositories (( learned S-pointer-networks! ))



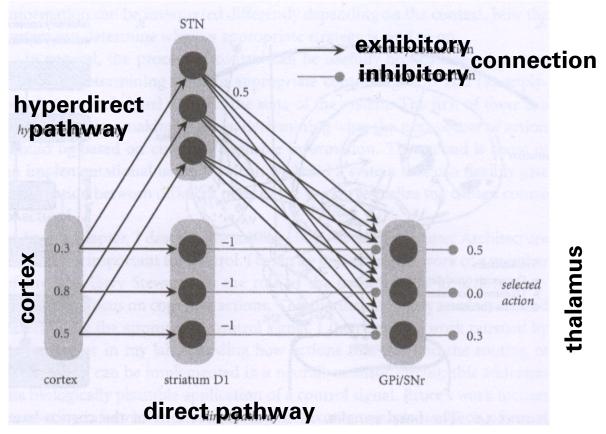
#### Speech processing model: Architecture

- In addition:
- Internal feedback loops: semantic level, phonological level, sensorimotor level
- External feedback loop: motor-articulatory-acoustic-auditory



# NENGO: Basal ganglia and Thalamus

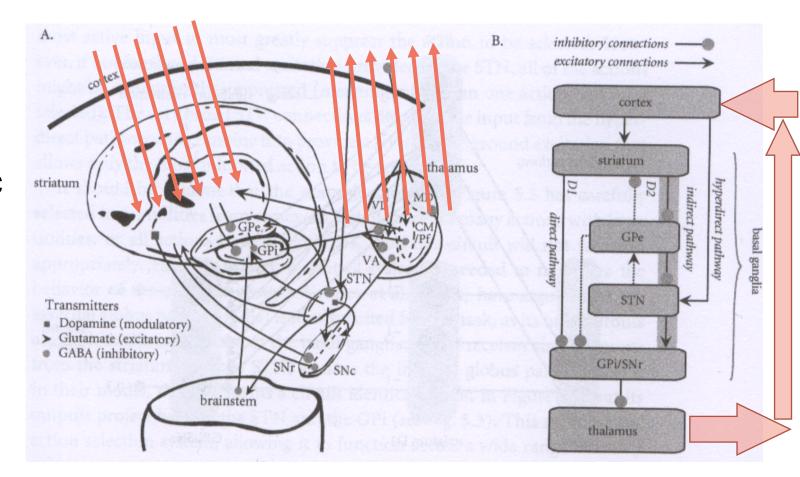
- the cortico-cortical control loop
- ... controls the temporal sequence of (neural) actions:
  - calculate utility values for each available action (cortex -> striat)
  - choosing the most appropriate action as next action based of the set of current utility values (thalamus)
  - Modules (neuron clusters) of BG:
    - striatum,
    - STN: subthalamic nucleus
    - GPi: globus pallidus, ...



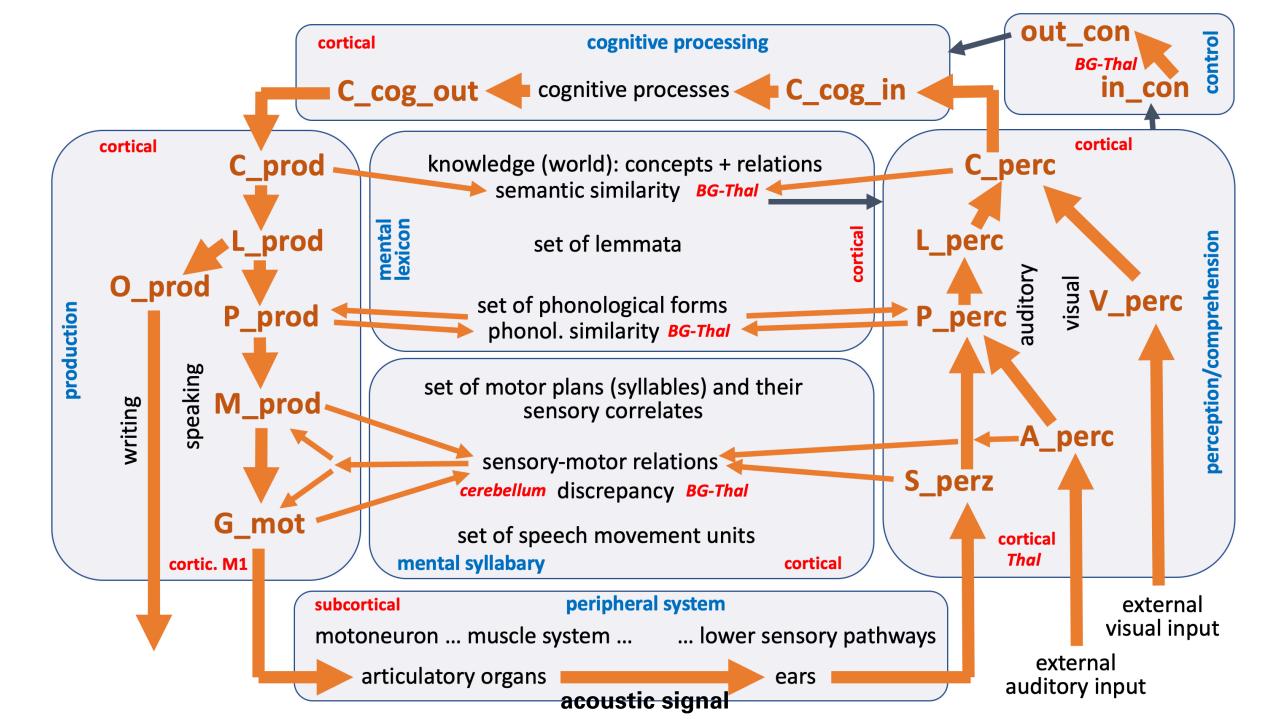
Eliasmith (2013)

#### **BG-Thal: the cortico-cortical loop**

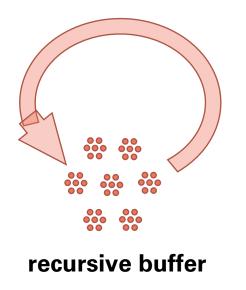
- direct / hyperdirect pathway
- different types of neurons (of synaptic connections) in different neuron clusters
- inhibitory and exhibitory pathways



Eliasmith (2013)

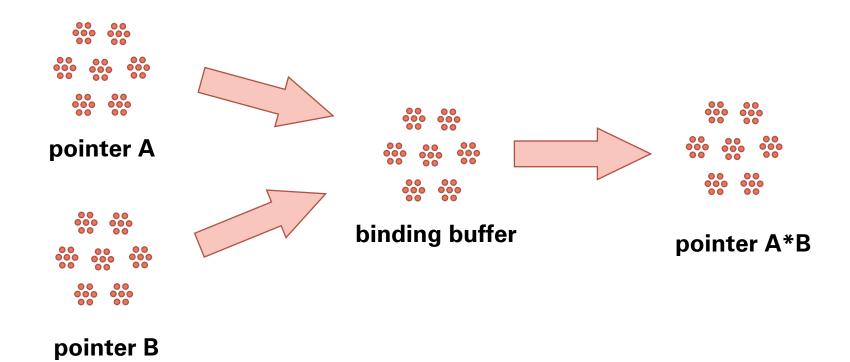


- Short-term memories, e.g., for memorizing word lists (-> serial recall task)
- realized as recursive buffers (neural within-associations)



- Cognitive processing needs:
- Binding and inverse binding (unbinding), e.g., for building sentences (-> the dog barks and the cat meows)
  - binding: <dog>\*<to bark> ; <cat>\*<to meow>
  - Sentence: (<dog>\*<actor> + <bark>\*<acting>
  - unbinding: what can the dog / cat do? -> <acting><sup>-1</sup>
- ... needed for memorizing orderings of list items (-> serial recall vs. free recall)

Binding / unbinding is realized by specific neural structures
 -> binding buffers vs. normal S-pointer buffers

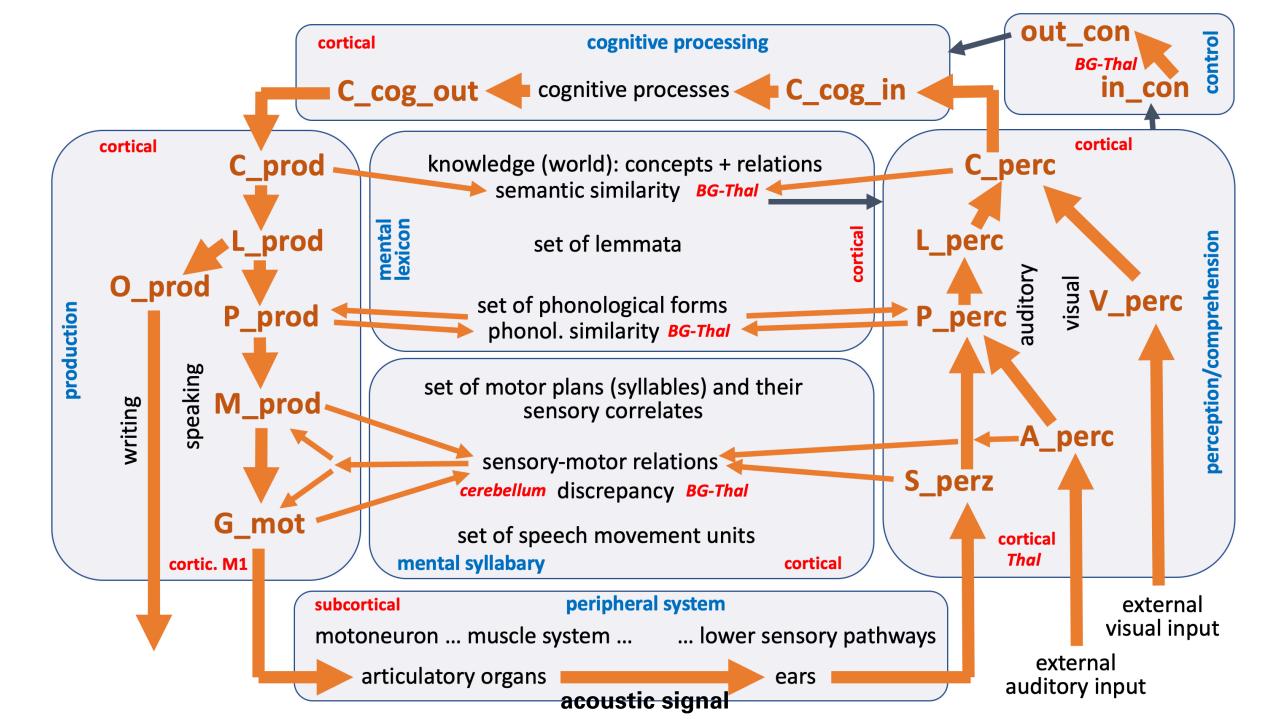


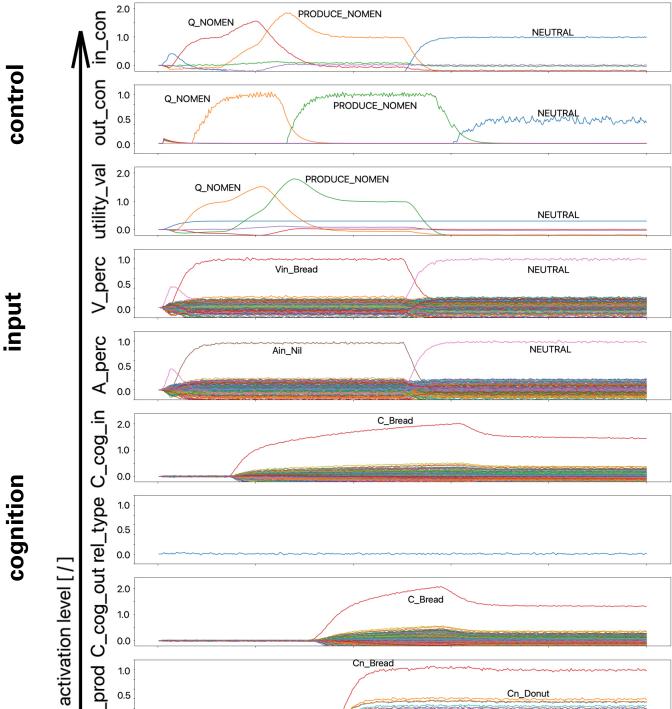
- Activating superordinate terms (e.g., a dog is an animal; a chair is a furniture, is an object)
- binding processes in S-pointer networks, using relations like "is a": <chair>\*<is\_a> => <object>

# The definition of scenarios

- Our large-scale model is more than just an architecture:
- In addition: we can define the "world" in which the speaker has to "act" (define scenarios)
- Scenario + model (avatar) -> simulated behavior
- Scenarios: speech screening tasks: (medical screenings)
  - Picture naming (visual input -> production)
  - Word comprehension (auditory input -> naming of superordinate term)
  - Nonsense word (logatome / syllable) repetition

Kroeger BJ, Stille C, Blouw P, Bekolay T, Stewart TC (2020) Hierarchical sequencing and feedforward and feedback control mechanisms in speech production: A preliminary approach for modeling normal and disordered speech. Frontiers in Computational Neuroscience, 14:99. <u>www.speechtrainer.eu</u> -> publications

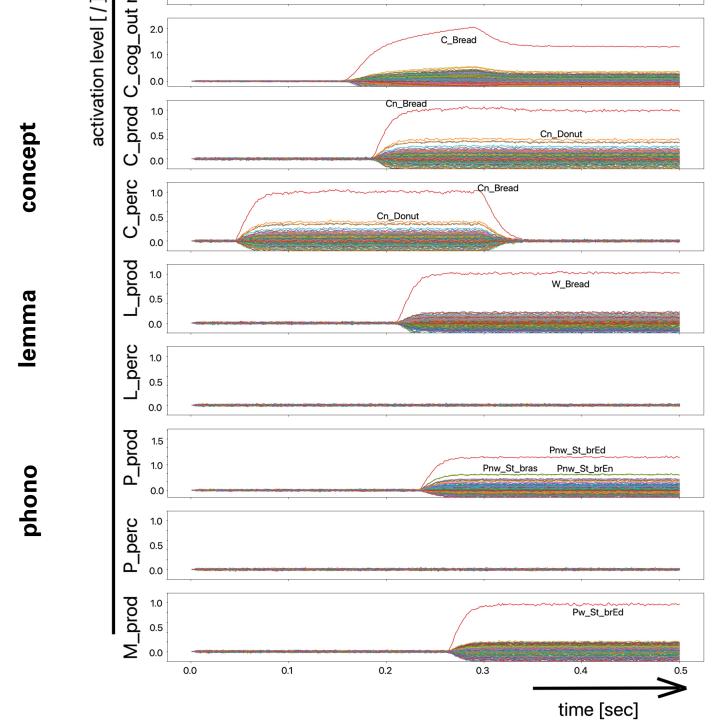




← priming: 100 msec: "look@pic and activate noun" +150 msec : "utter noun"

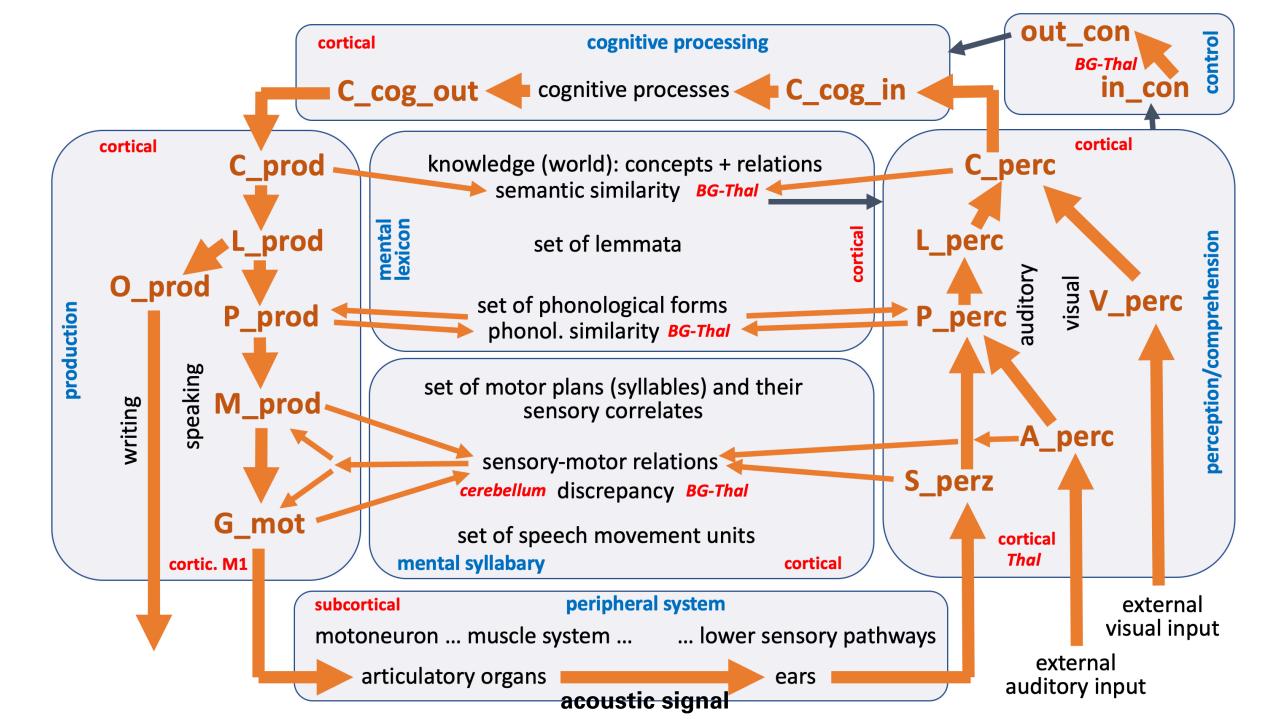
input

#### Kröger et al. (2020)



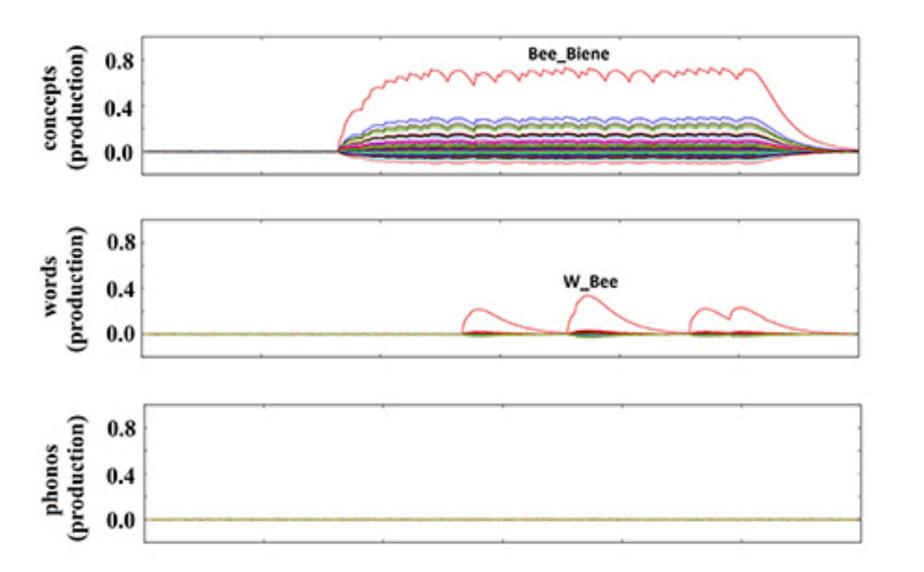
Kröger et al. (2020)

# pathways



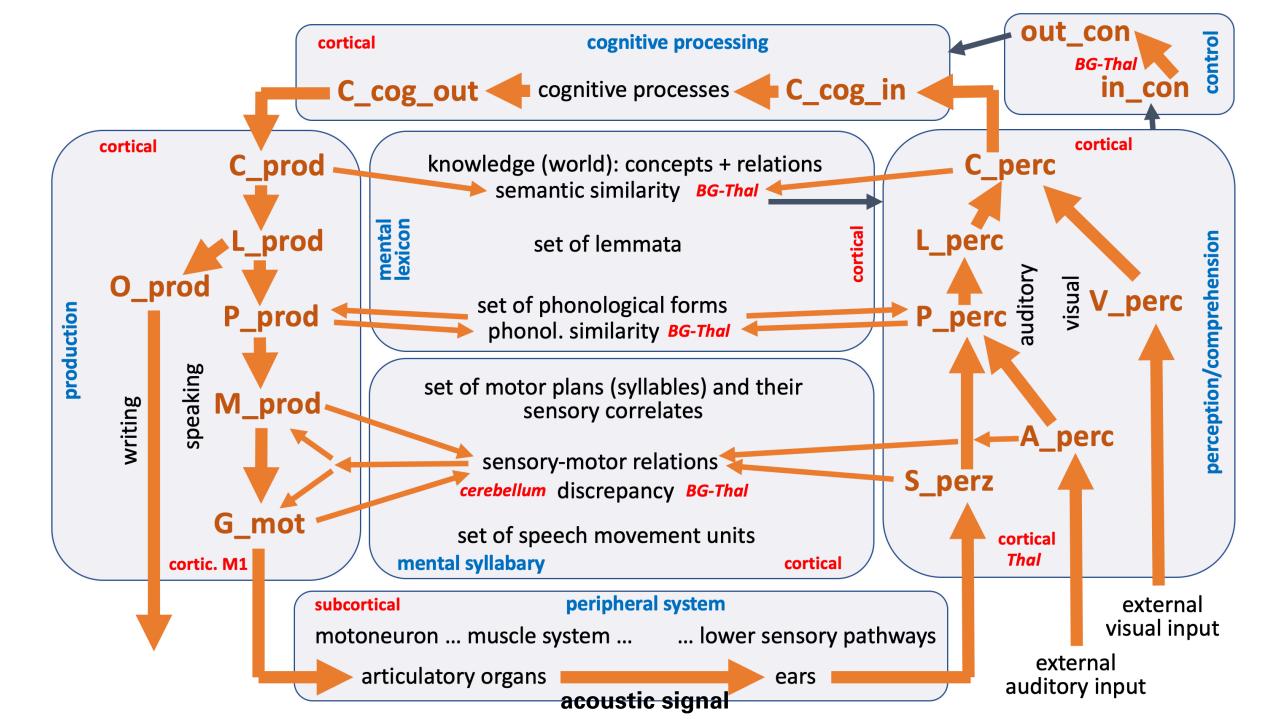
#### The definition of scenarios

- Speech screening tasks:
  - Picture naming (visual input -> production) rare cases (<1% of simulations)</li>
  - -> underline the neural character of the model (neural spikes are statistic events! -> neural signals are noisy!)



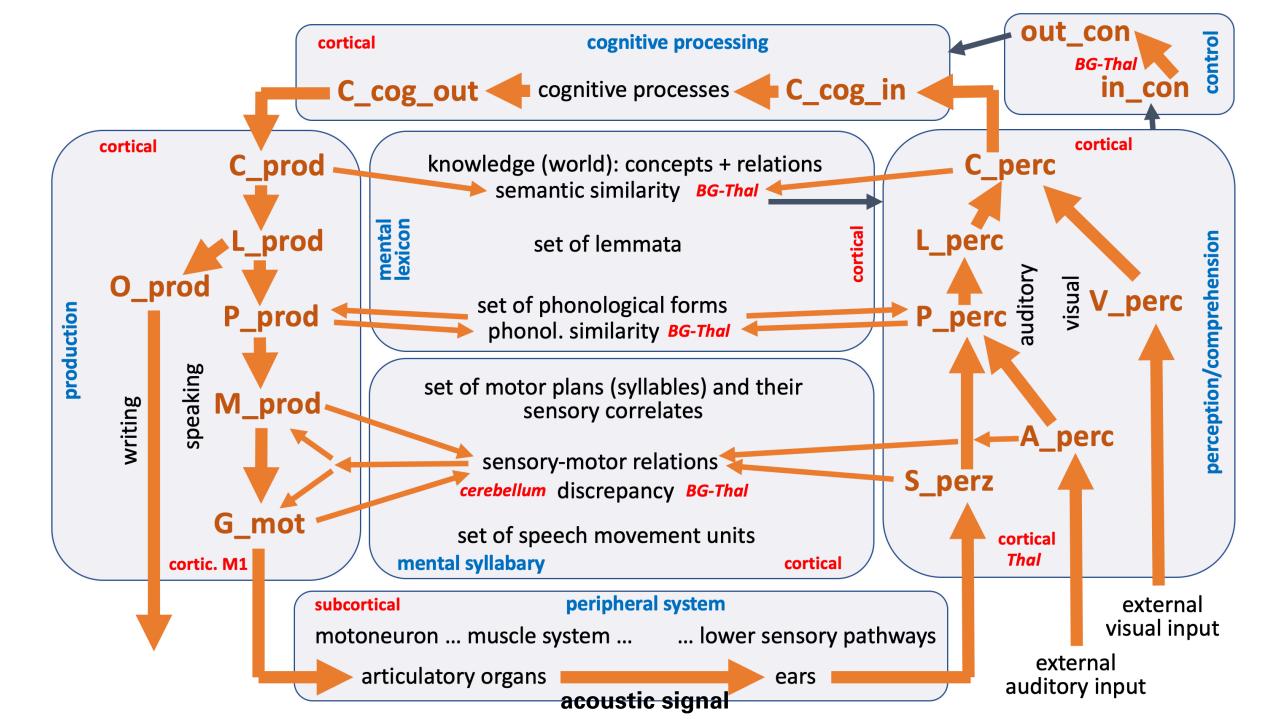
normal model; no neural dysfunctions implemented!

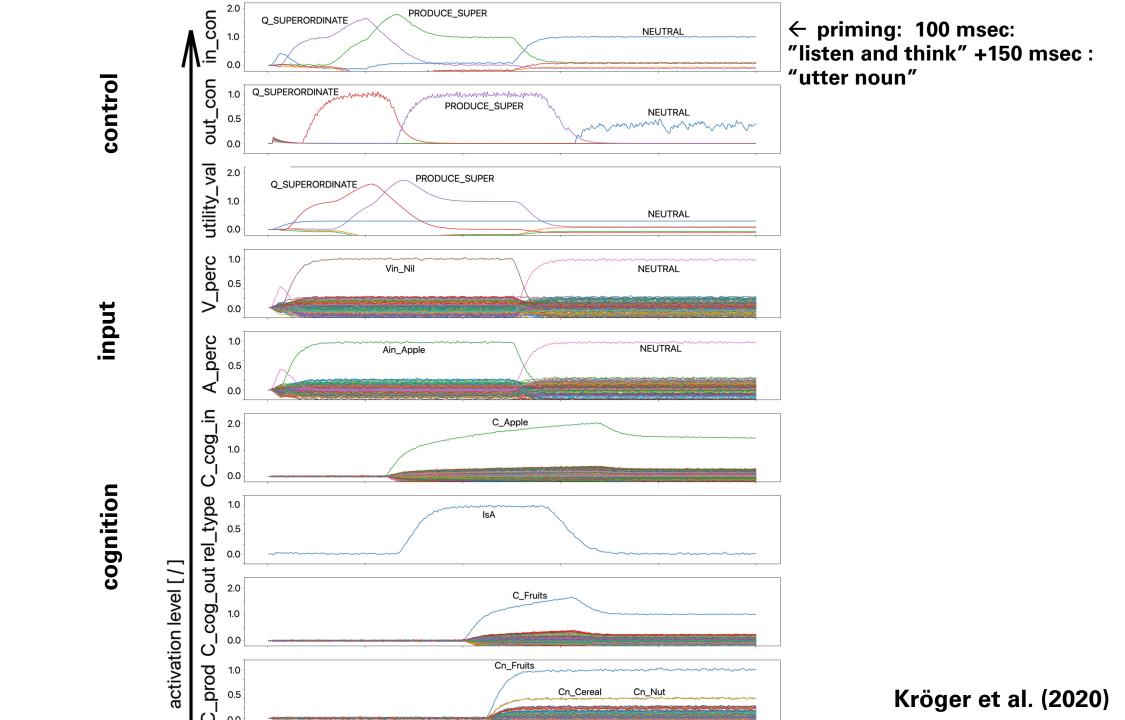
Kröger BJ, Crawford E, Bekolay T, Eliasmith C (2016) Modeling interactions between speech production and perception: speech error detection at semantic and phonological levels and the inner speech loop. Frontiers in Computational Neuroscience 10:51 <u>www.speechtrainer.eu</u> -> publications

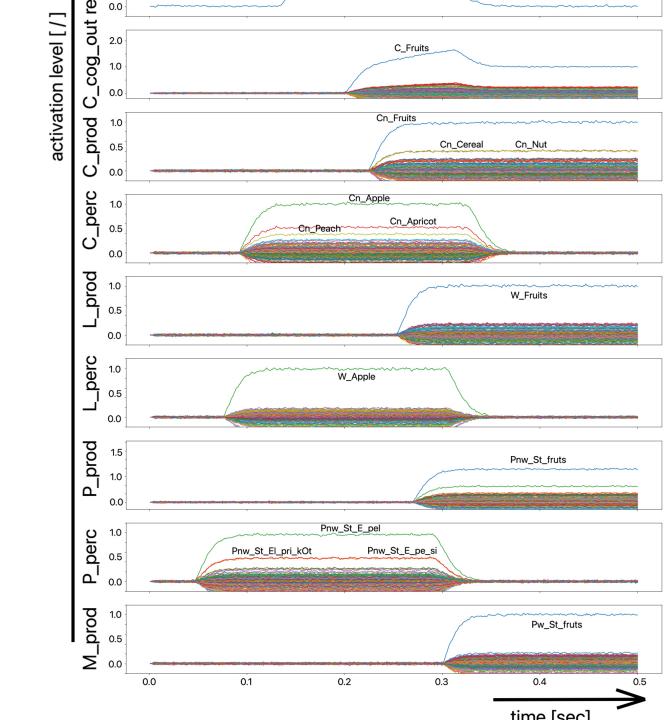


- defines the "world" in which the speaker has to "act"
- -> behavior
- Speech screening tasks:
  - Picture naming (visual input -> production)
  - Word comprehension (auditory input -> naming of superordinate term)
    - Is a cognitive task including binding
  - Nonsense word (logatome) repetition

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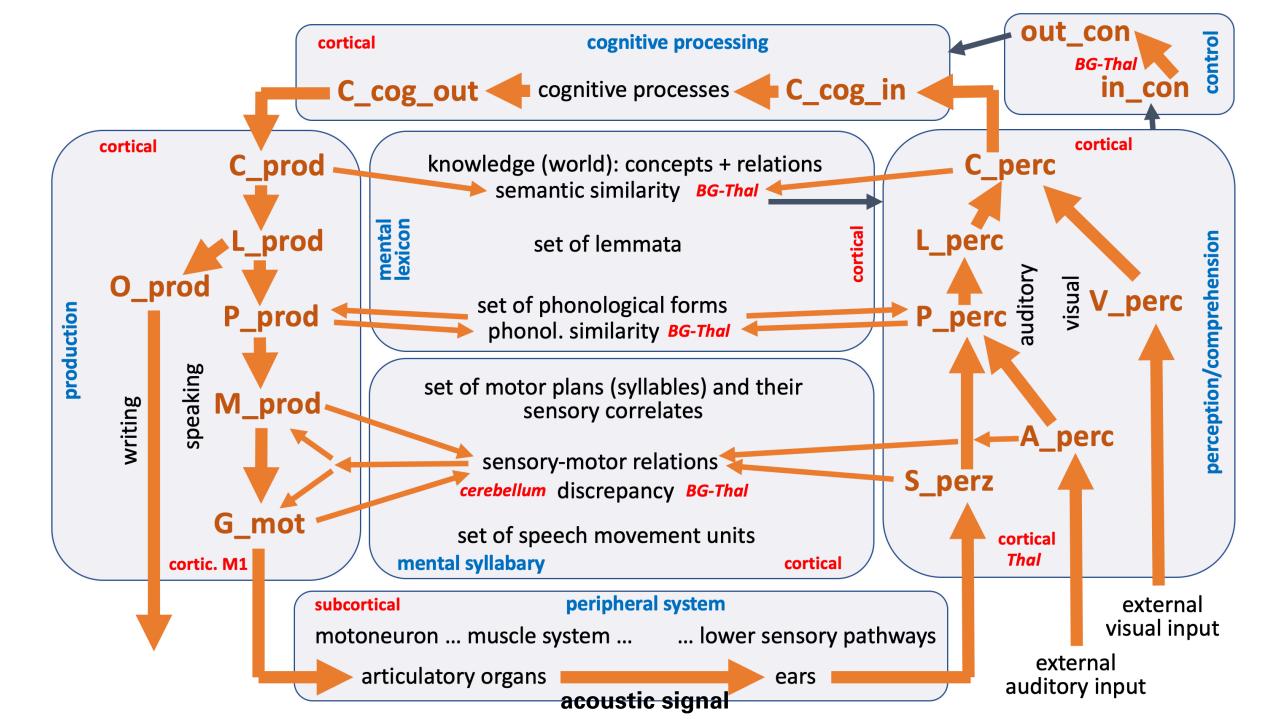
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## pathways

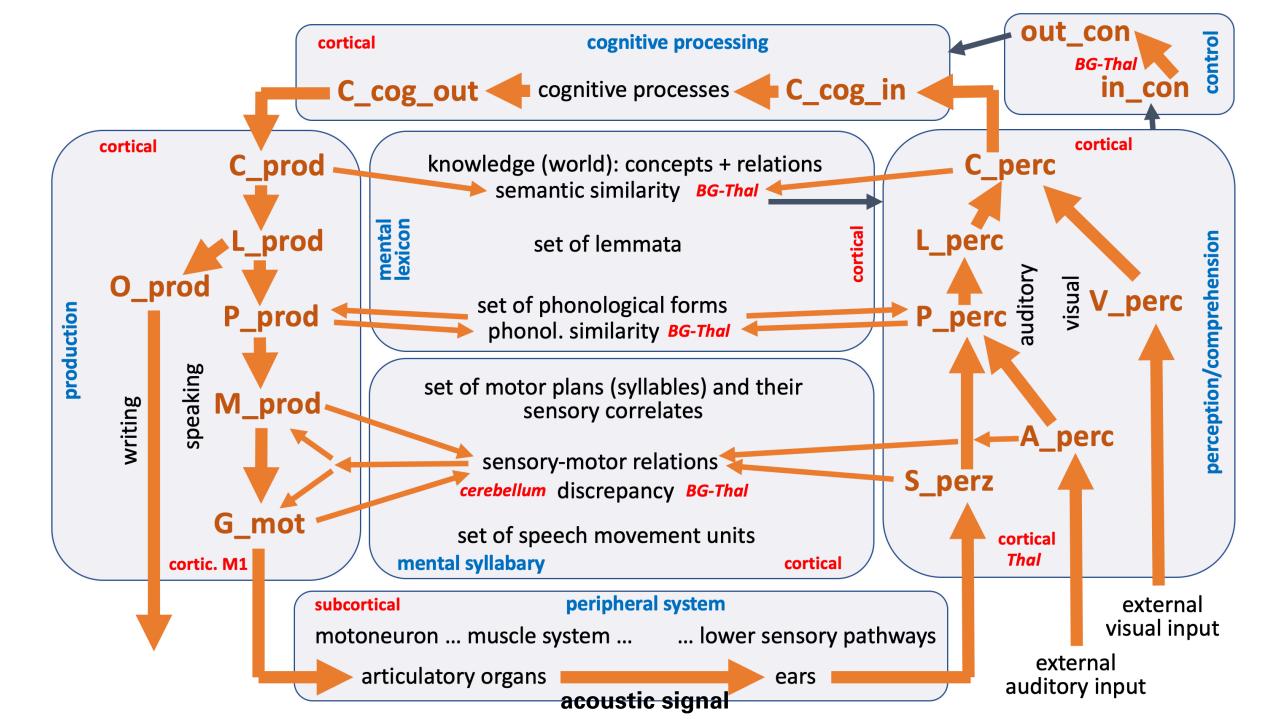
# lemma

phono

## concept



- defines the "world" in which the speaker has to "act"
- -> behavior
- Speech screening tasks:
  - Picture naming (visual input -> production)
  - Word comprehension (auditory input -> cognitive activation and naming of superordinate term)
  - Nonsense word (logatome) repetition
    - Do not need higher level processing; just shortcut at phonological level
    - Phono-shortcut means: repetition without comprehension; without any lexical access

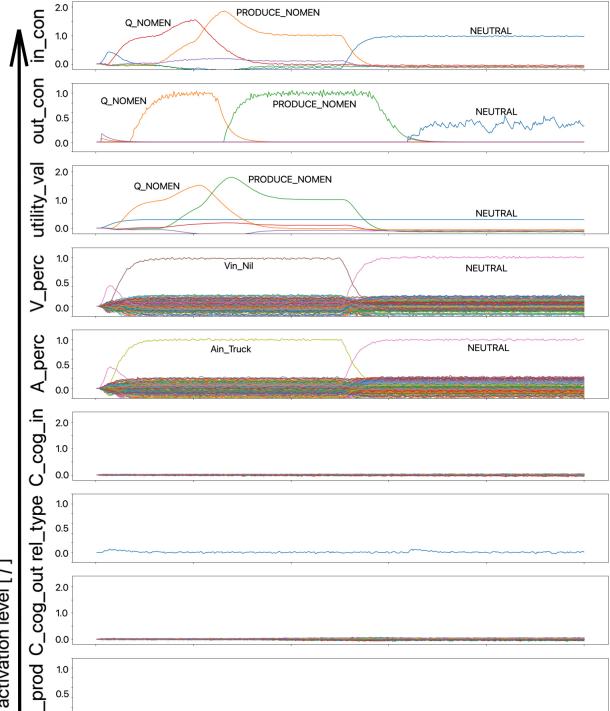






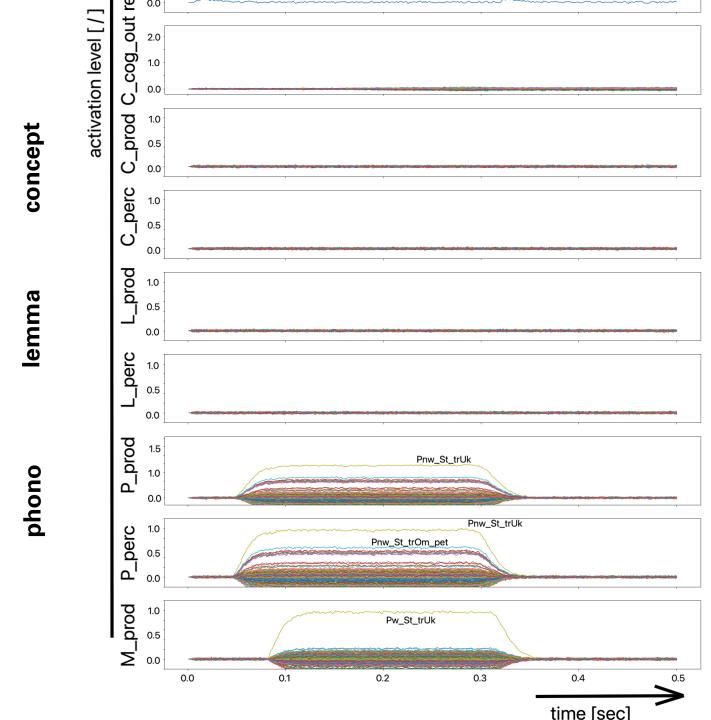
control





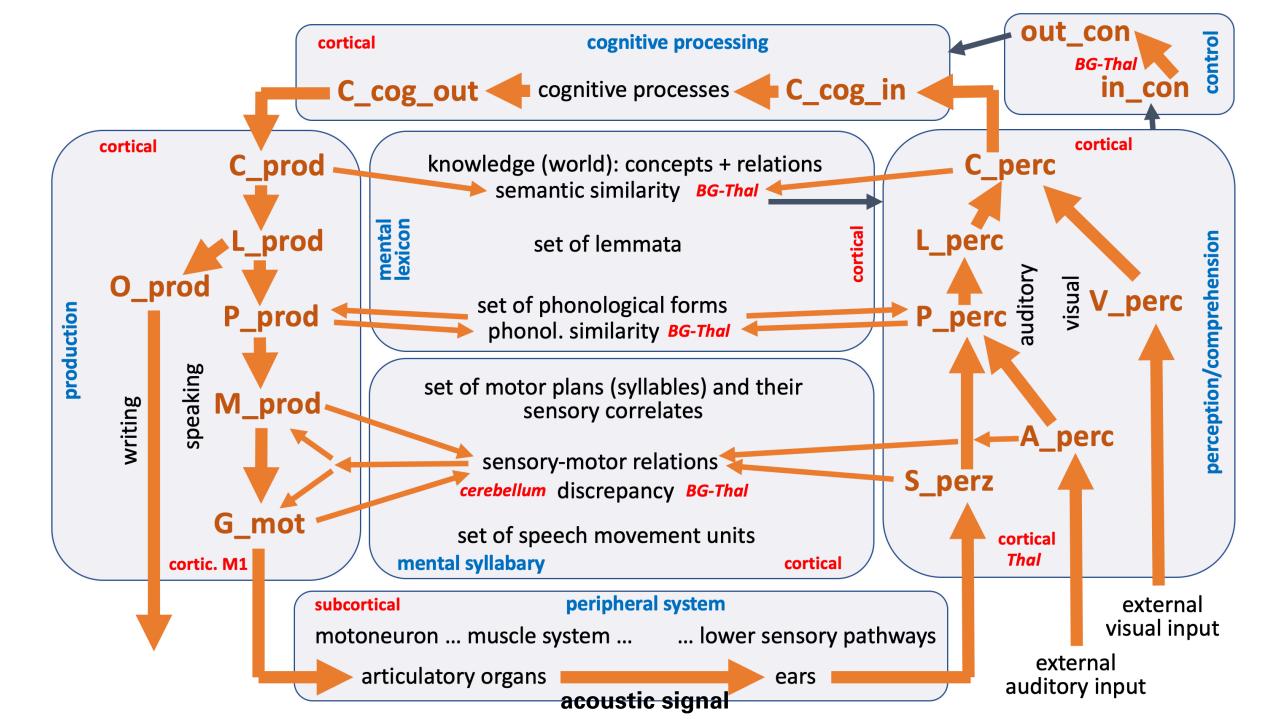
← priming: 100 msec:
"listen" + "reproduce"
150 msec : "further reproduce"

Kröger et al. (2020)

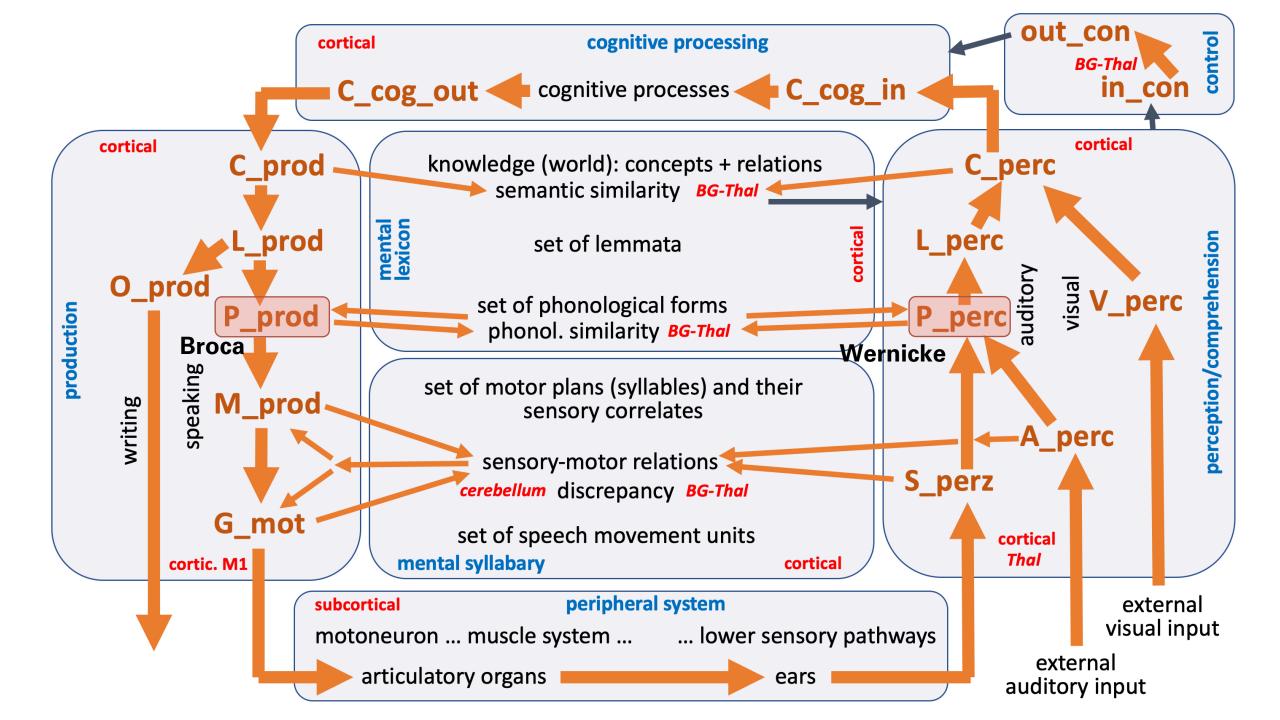


Kröger et al. (2020)

# pathways

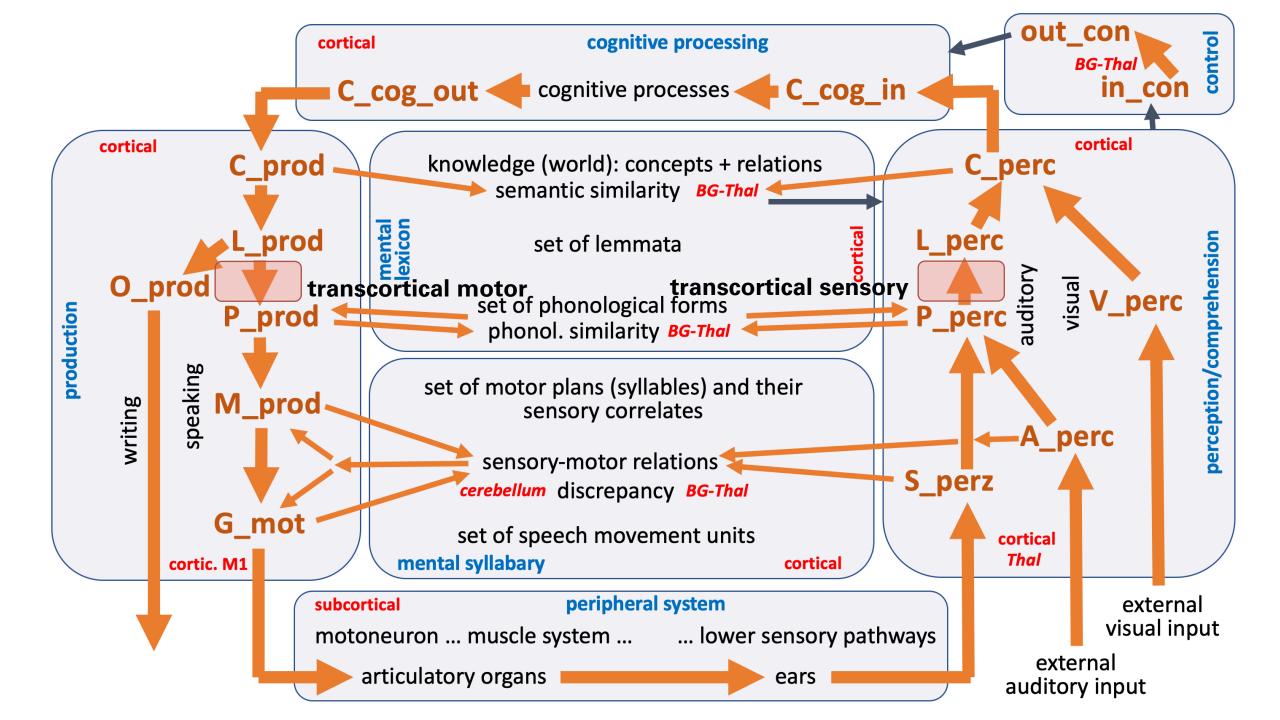


- We will perform the same tasks, but now including:
- Lesions at different levels of the model:
  - Phonological state buffer -> Broca / Wernicke aphasia
  - Associative memories between phono-lemma buffer -> transcortical motor / sensory aphasia
  - Associative memories between lemma-concept buffers -> mixed aphasia
  - Neural association between phono-phono buffers -> conduction aphasia
- Lesions -> decrease in task performance



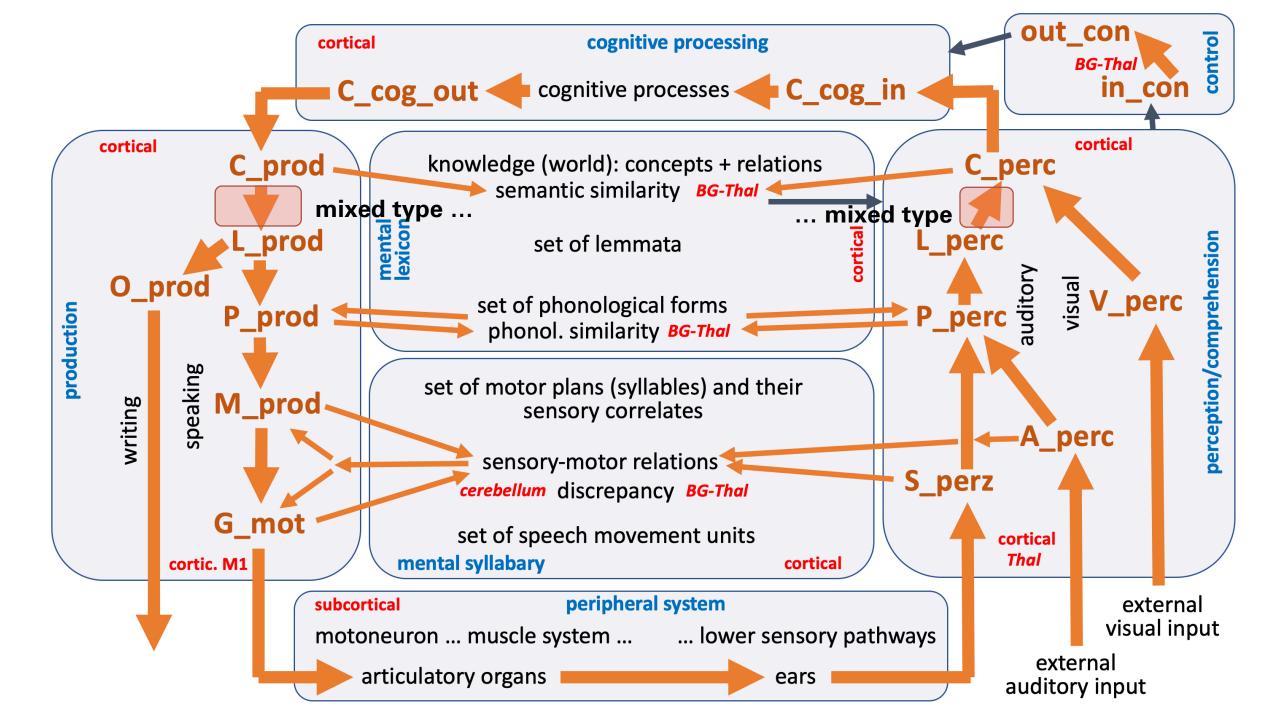
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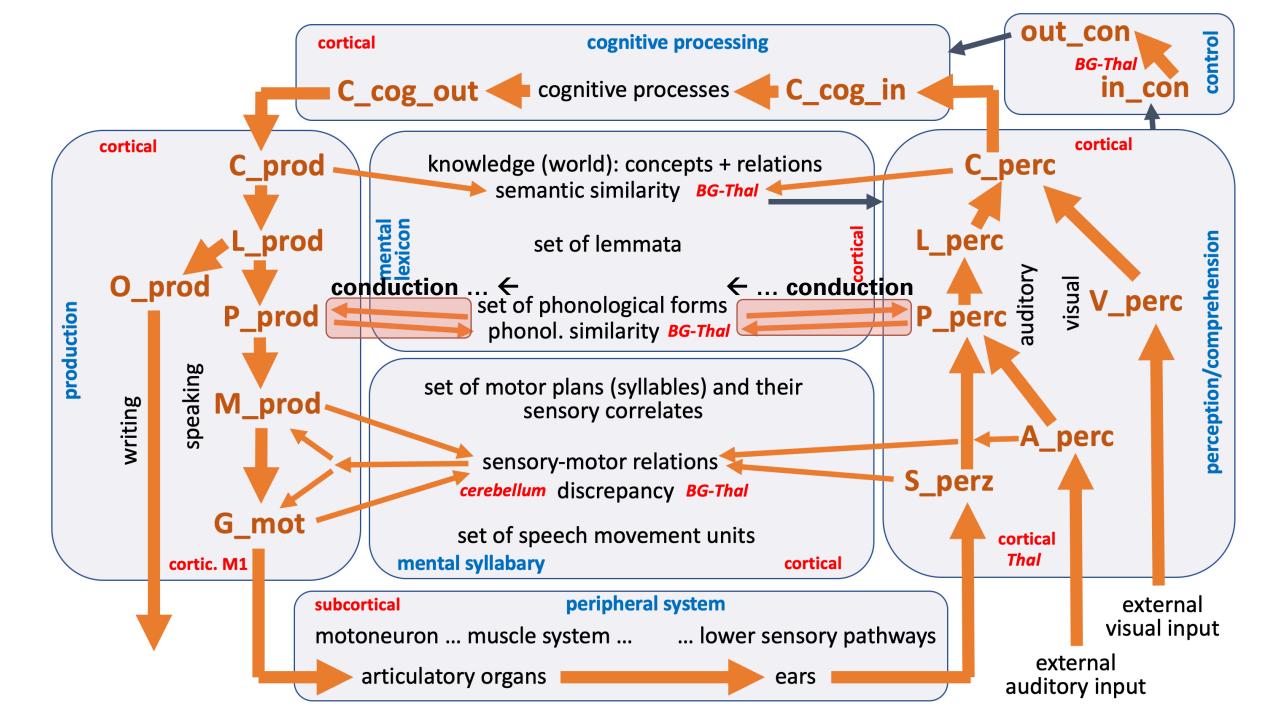


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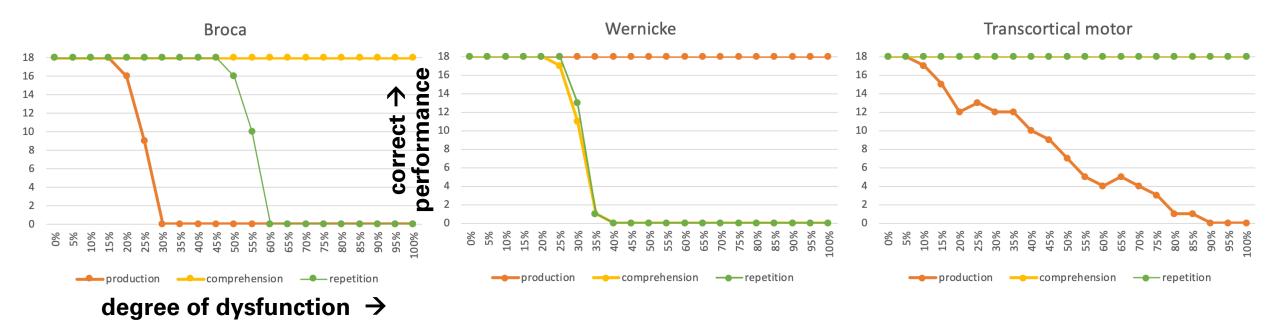
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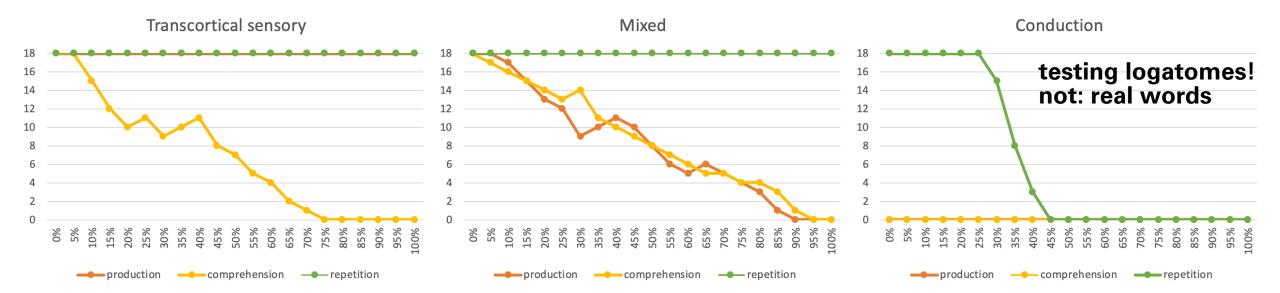


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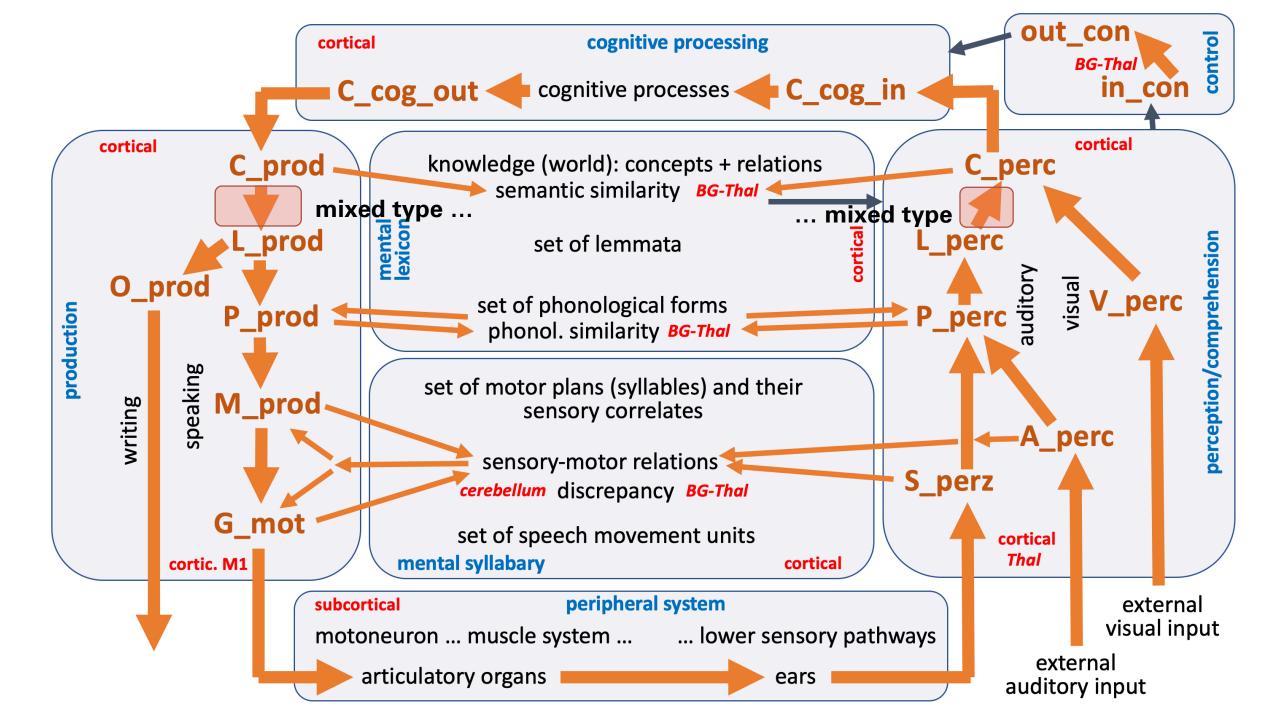


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  - Neural association between phono-phono buffers -> conduction aphasia
- Lesions -> decrease in task performance (symptoms!)





Kröger et al. (2020)



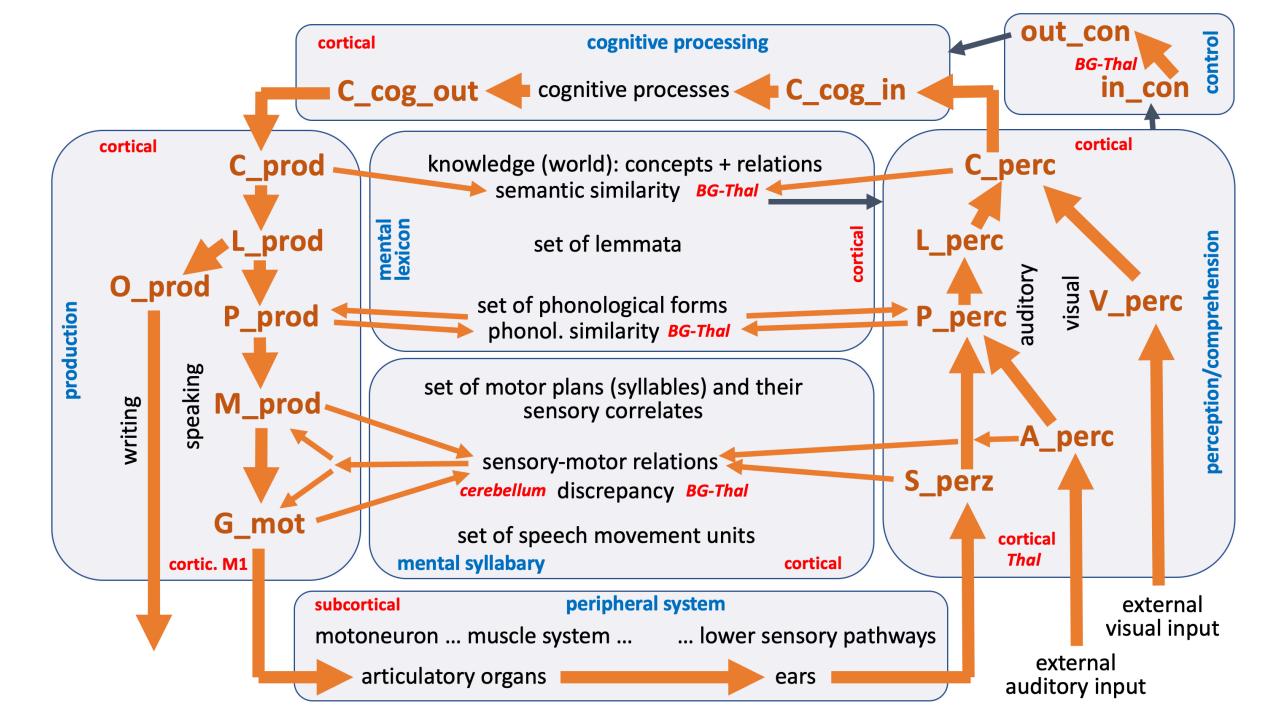
Download Video4: <u>https://www.youtube.com/watch?v=xzAgOnna9Mw</u>

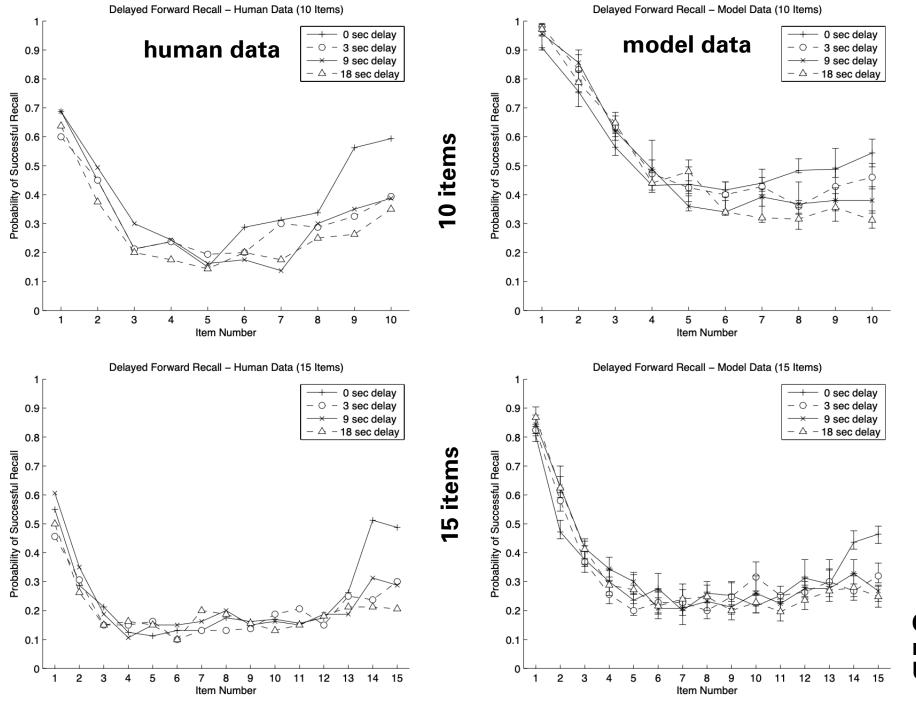
- So far:
  - simple production task (picture naming)
  - comprehension tasks
  - simple repetition task
- more complex tasks:
- Further speech screenings: (-> demonstrating realism of model)
  - Serial recall (10/15 words; Choo 2010, master thesis, Waterloo Univ.)
  - Picture naming with phonological or semantic distractors words
  - Picture naming in case of lexical access/retrieval problems: "tip of the tongue" case: -> help by introducing phonological and semantic cues

### Why more simulations?

- Demonstrating the realism of neural models
- -> we already demonstrated:
  - normal behavior (naming, comprehension, repetition tasks)
  - rare cases (fail in naming in < 1% of all simulations)</li>
  - introduction of neural dysfunctions -> simulation of symptoms in case of different speech disorders (diff. types of aphasias)
- Further simulations:
  - simulating "overburdening" (-> serial recall of 15 words)
  - simulating "difficult task": -> get more faults / mistakes
    - e.g.,: simulation of "speech errors" by using distractors;
    - e.g.,: simulating "tip of the tongue" cases;

- more complex tasks:
  - Serial recall (10/15 words; Choo 2010, master thesis, Waterloo Univ.)
    - -> lowest repetition rate for words in the middle (concave result graph)
    - Short term memory and binding of positions and concepts
  - Picture naming with phonological or semantic distractors (acoustically introduced) -> decrease in performance
  - Picture naming in case of lexical access problems ("tip of the tongue" case): help by introducing phonological and semantic cues -> increase in performance





Choo FX (2010), master thesis, Univ. Waterloo

- Further speech screenings: (-> demonstrating realism of model)
  - Picture naming with phonological or semantic distractors (acoustic)
    - specific word list: including phonological and semantic dis-/similar words
    - -> generation of direct HALTS / no or late HALTS in different cases by including a evaluation of similarity of items at phonological and at semantic level as part of the internal feedback loops

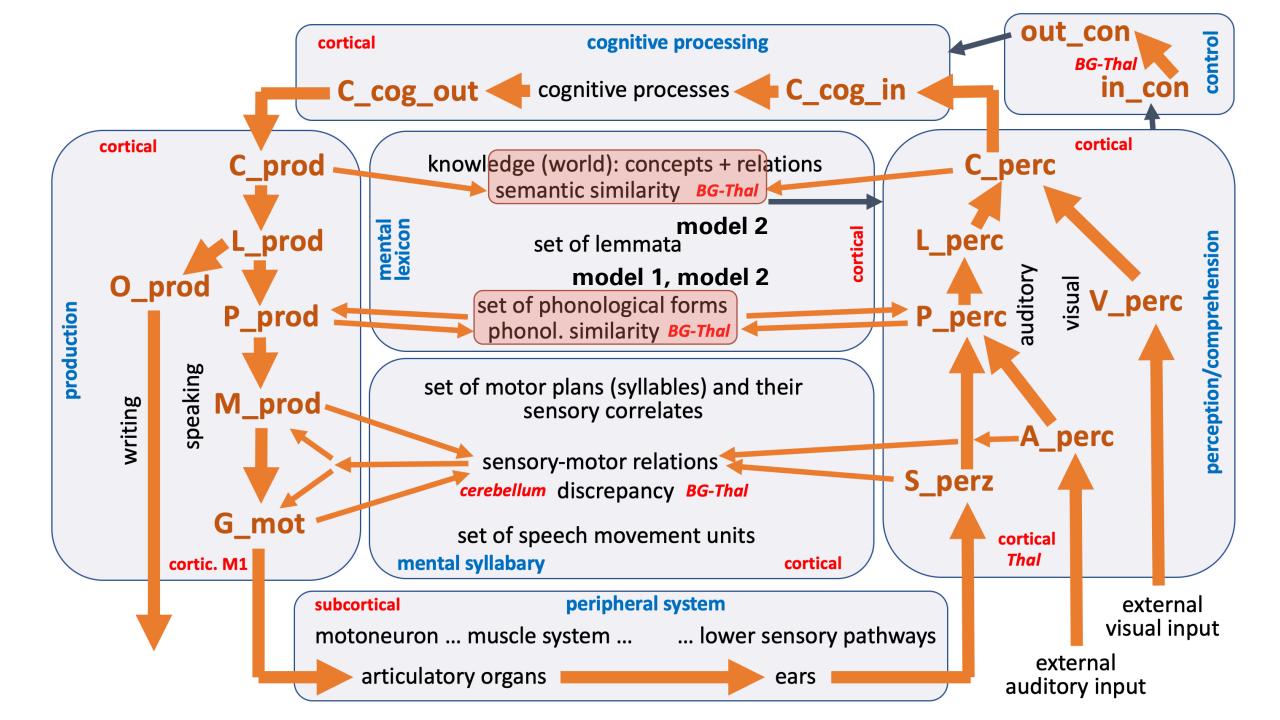
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#### Word list -> mental lexicon: 18x6 =104 words

Target word	Semantically similar word	Phonologically similar word	Phonologically and semantically similar word	Dissimilar word	Semantically superordinate item (semantic cues)	Similar sound segments (phonological cues)
Apple	Peach	Apathy	Apricot	Couch	Fruits	/Ep/
Basket	Crib	Ban	Bag	Thirst	Bin	/bE/
Bee	Spider	Beacon	Beetle	Flag	Crawler	/bi/
Bread	Donut	Brick	Bran	Nail	Cereal	/br/
Camel	Pig	Cash	Calf	Bucket	clovenHooved	/kE/
Carrot	Spinach	Cast	Cabbage	Evening	Veg	/kE/
Duck	Raven	Sub	Dove	Brass	Bird	/da/
Elephant	Moose	Elm	Elk	Stripe	HornAnimal	/El/
Fly	Moth	Flu	Flea	Rake	Bluebottle	/fl/
Lamp	Candle	Landing	Lantern	Package	LightSource	/IE/
Peanut	Almond	Piano	Pecan	Dress	Nut	/pi/
Rabbit	Beaver	Raft	Rat	Coffee	Rodent	/rE/
Snake	Eel	Snack	Snail	Fire	Invertebrate	/snE/
Spoon	Ladle	Sparkle	Spatula	Cable	Lifter	/sp/
Squirrel	Mole	Skate	Skunk	Chain	HairySkin	/sk/
Train	Bus	Trophy	Trolley	Fox	PublicTrans	/tr/
Truck	Jeep	Trap	Tractor	Celery	UtilityVehicle	/tr/
Trumpet	Horn	Traffic	Trombone	Corner	BrassWind	/tr/

- Further speech screenings: (-> demonstrating realism of model)
  - Picture naming with phonological or semantic distractors (acoustic)
    - specific word list: including phonological and semantic dis-/similar words
    - -> generation of direct HALTS / no or late HALTS in different cases by including a evaluation of similarity of items at phonological and at semantic level as part of the internal feedback loops

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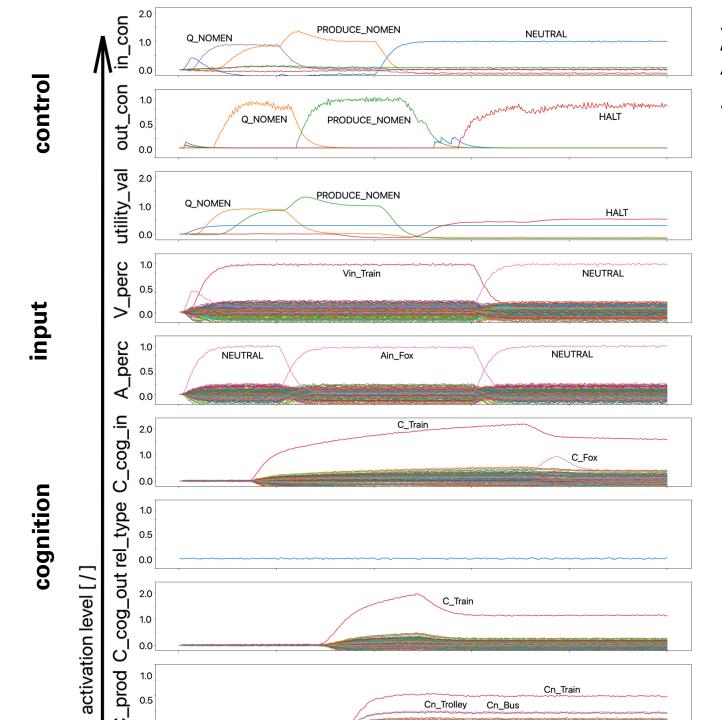


#### Results: picture naming with distractors

- 18 items, 3 runs -> 54 simulations per sub-experiment
  - 4 different distractor words, 2 different models -> 432 simulations
- -> HALT events in case of dissimilar distractor words
  - No HALT in case of similar distractor words (suppress early HALT)

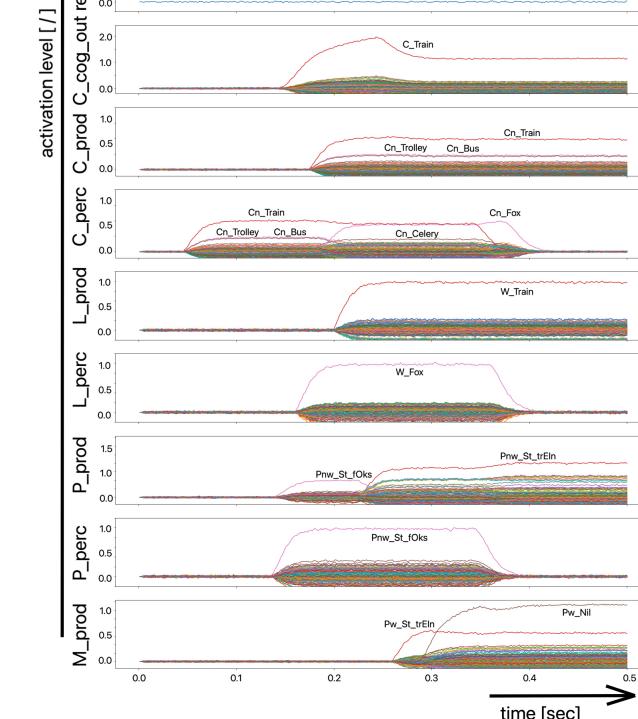
Type of model	Type of distractor word	Number of stops (test series 1)	Number of stops (test series 2)	Number of stops (test series 3)	Number of stops (sum)	Number of simul. Without stop (sum)
1	Semantic similar	1	2	0 no strou	3 ng suppression	51
<sup>1</sup> diff-eval @ semantic +	Phonological similar	10	10	9 effect h		25
1 phono level	Sem + phono similar	1	1	0	> 2	52
1	Dissimilar	17	18	15	> 50	4
2	Semantic similar	15	15	16	46	8
2 diff-eval @	Phonolgical similar	2	2	2	6	48
2 phono level only	Sem + phono similar	0	1	1	2	52
2	Dissimilar	17	14	13	44	10

- Speech screenings: (-> demonstrating realism of model)
  - Picture naming with phonological or semantic distractors (acoustic)
    - Case: phonological and semantic dissimilar word
    - -> generation of HALT signal (-> word is NOT produced)



← priming: 100 msec:
"look and think" +150 msec :
"utter noun" + later:
... HALT may be generated!

Kröger et al. (2020)



Kröger et al. (2020)

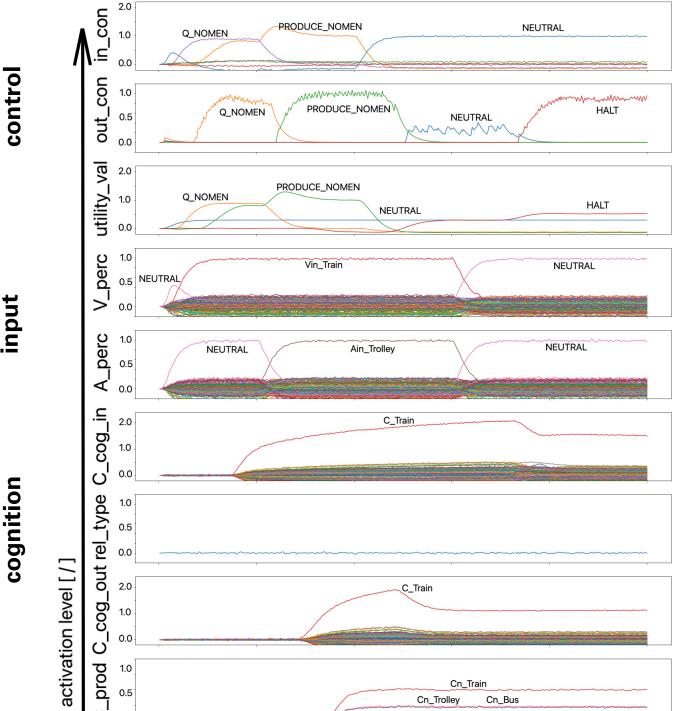
# pathways

### phono

lemma

concept

- Speech screenings: (-> demonstrating realism of model)
  - Picture naming with phonological or semantic distractors (acoustic)
    - Case: phonological and semantic similar word
    - -> suppress early HALT signal (-> word is produced)



← priming: 100 msec: "look and think" +150 msec : "utter noun" + later: ... HALT may be generated!

Kröger et al. (2020)

cognition

input

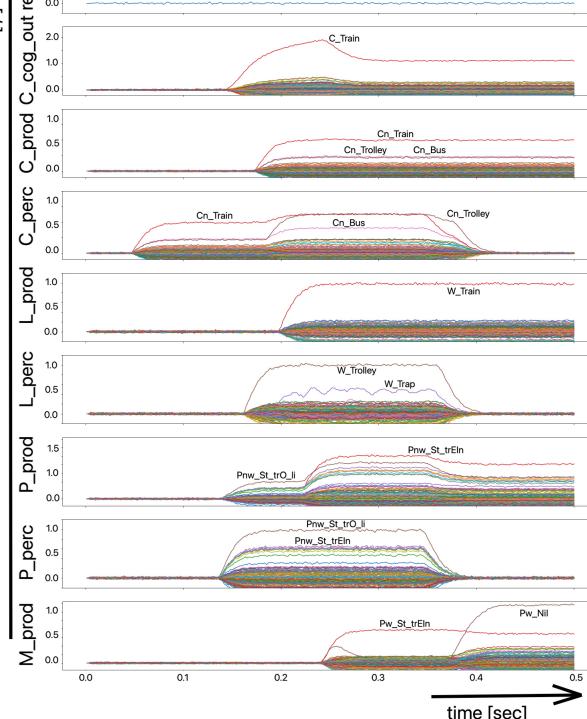


# phono

lemma

# concept





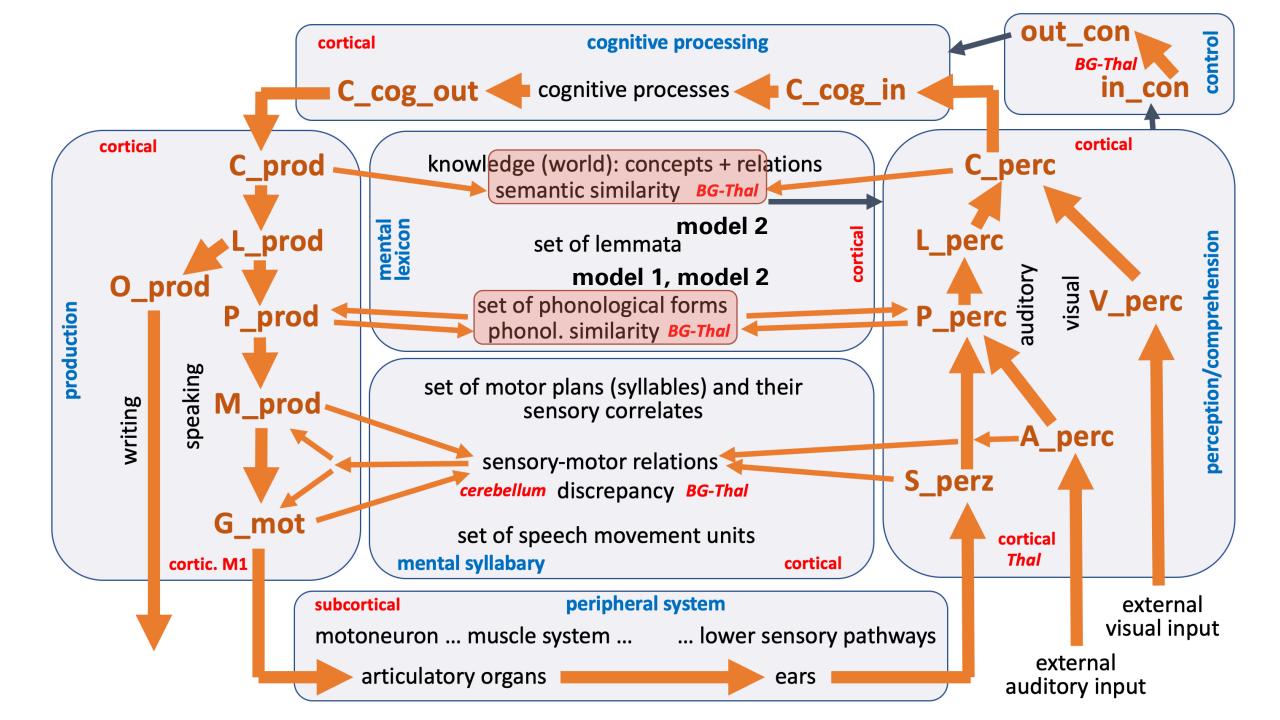
# Results: picture naming with distractors

- -> HALT events in case of dissimilar distractor words
  - No HALT in case of similar distractor words (suppress early HALT)
- Which model is realistic? Neither!
  - Need: diff-eval of 0.5 \* semantic + phono level (??)
  - But: Basic law in modeling: do not adapt the model parameters (← exp. results)

Ту	pe of model	Type of distractor word	Number of stops (test series 1)	Number of stops (test series 2)	Number of stops (test series 3)	Number of stops (sum)	Number of simul. Without stop (sum)
1		Semantic similar	1	2	0 too strong sup -> 3		51
	diff-eval @ semantic +	Phonological similar	10	10	9 too weak sup -> 29		25
1	phono level	Sem + phono similar	1	1	0 too strong sup -> 2		52
1		Dissimilar	17	18	15	50	4
2		Semantic similar	15	15	16	46	8
2	diff-eval @ phono level only	Phonolgical similar	2	2	2 too stro	ng sup -> 6	48
2		Sem + phono similar	0	1	1 too stro	ng sup -> 2	52
2		Dissimilar	17	14	13	44	10

# Models need to be straight forward

- define the model qualitatively AND quantitatively on the basis of rules for neural functioning
- Generate (qualitative and quantitative) simulation resultson this basis! without changing / adjusting the model
- See the model:
  - in all cases (neurons, neural connections) we use the default parametrization given by NENGO
  - we suggest a post-adaptation of the model only in case of one black arrow:



Download Video5: https://www.youtube.com/watch?v=KdJ6KV0sA-8

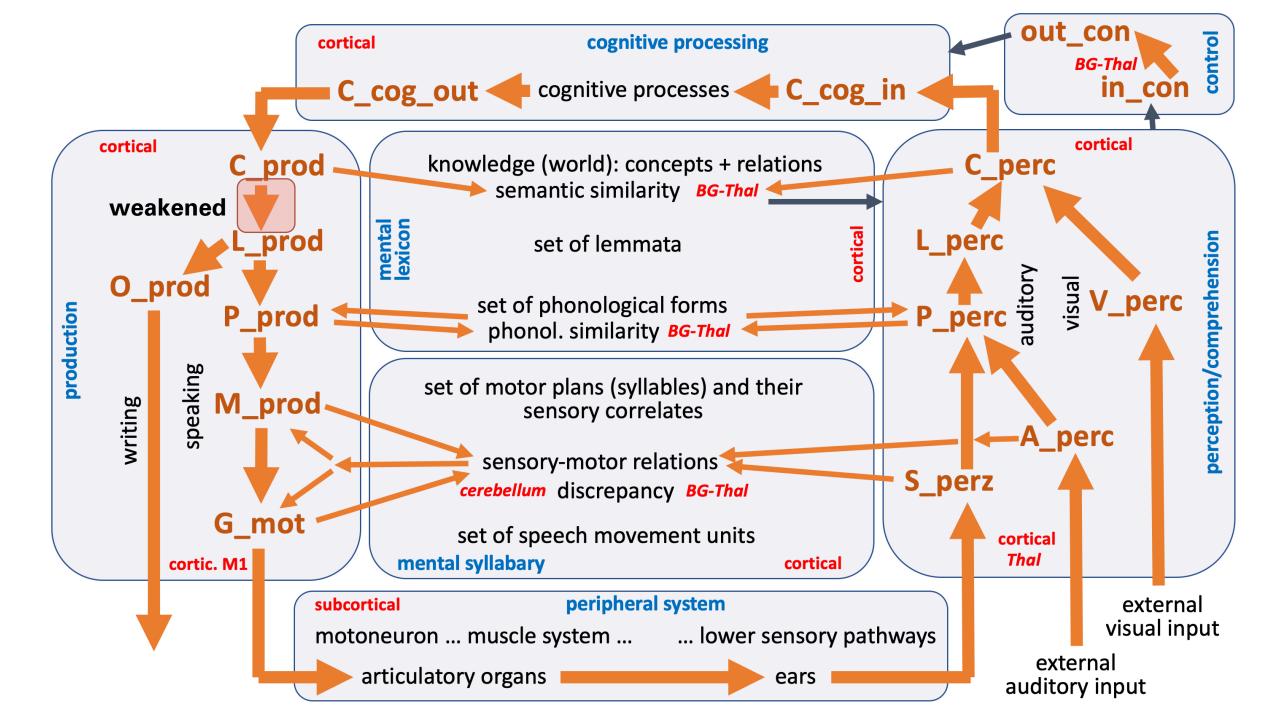
# The definition of scenarios

- more complex tasks:
- Further speech screenings: (-> demonstrating realism of model)
  - Serial recall (10/15 words; Choo 2010, master thesis, Waterloo Univ.)
    - -> lowest repetition rate for middle words (concave result graph)
  - Picture naming with phonological or semantic distractors (acoustically introduced) -> decrease in performance; induces HALTs
  - Picture naming in case of lexical access problems ("tip of the tongue" case): help by introducing phonological and semantic cues -> increase in performance

# The definition of scenarios

- Further speech screenings: (-> logopedic research)
  - Picture naming with phonological or semantic cues (acoustic)
    - Naming in case of "word is on tip of my tongue" (patient with problems in lexical access / lexical retrieval); ->
    - only 20-45% of correct productions occur without any cue
  - What if we use cues, if a word is not produced or not produced correctly?
    - Semantic cue: target word "car"; cue: "you can drive it, has four wheels "
    - Phonological cue: target word "car": cue: "it starts with /k/"

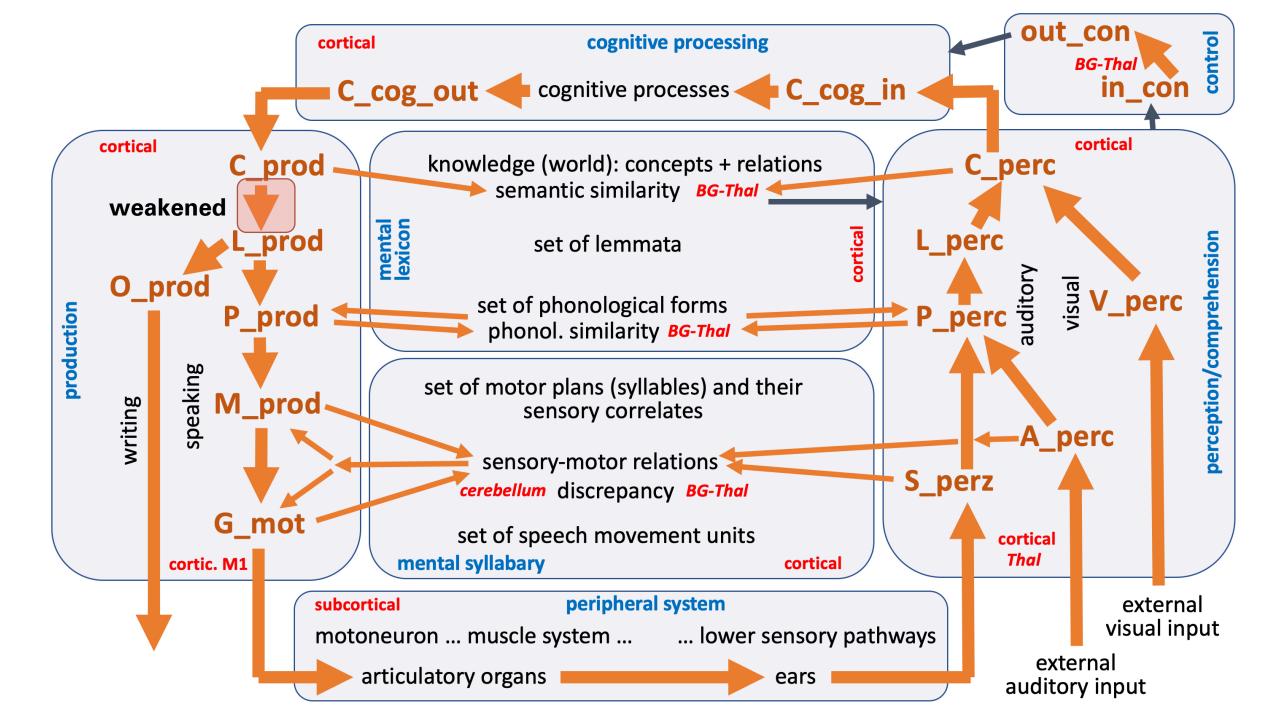
Kroeger BJ, Stille C, Blouw P, Bekolay T, Stewart TC (2020) Hierarchical sequencing and feedforward and feedback control mechanisms in speech production: A preliminary approach for modeling normal and disordered speech. Frontiers in Computational Neuroscience, 14:99. <u>www.speechtrainer.eu</u> -> publications



# The definition of scenarios

- Further speech screenings: (-> logopedic research)
  - Picture naming with phonological or semantic cues (acoustic)
    - Naming in case of "word is on tip of my tongue" (patient with problems in lexical access / lexical retrieval); ->
    - only 20% of correct productions occur without any cue
  - How do cues work in the model, if a word is not produced or not produced correctly?
    - Semantic cue: increases neural activity C\_perz -> lexical loop
    - Phonological cue: increases neural activity at P\_perc -> lexical loop or:
    - phonological shortcut may be more effective!

Kroeger BJ, Stille C, Blouw P, Bekolay T, Stewart TC (2020) Hierarchical sequencing and feedforward and feedback control mechanisms in speech production: A preliminary approach for modeling normal and disordered speech. Frontiers in Computational Neuroscience, 14:99. <u>www.speechtrainer.eu</u> -> publications



# Cues are helpful:

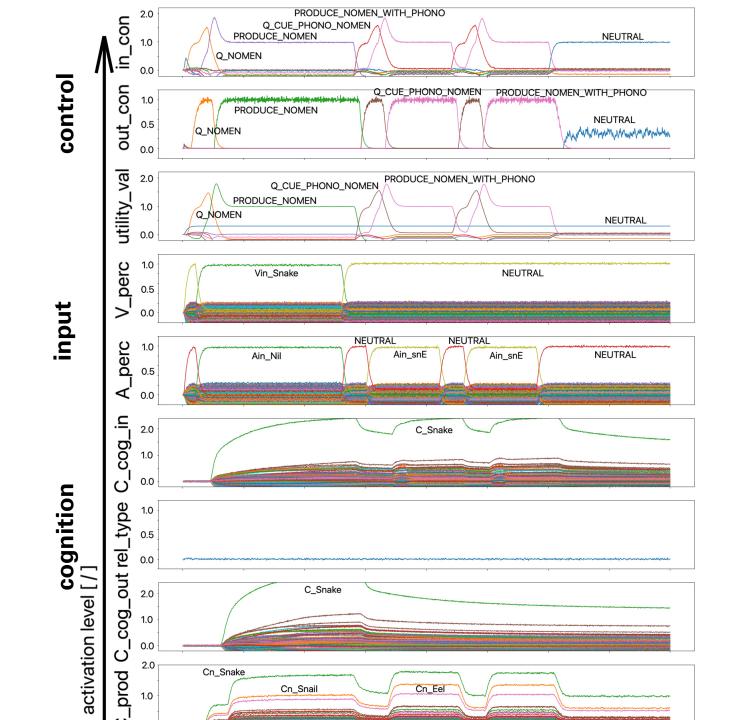
- 2 different cues; 2 different models: with/without shortcut perc->prod at phonological level
- Starting with 20-45% of correct word naming; increase to about 55-70%

Number of test series: type of model and type of cue	Number of target word activation before/after cue (test-series 1)	Number of target word activation before/after cue (test-series <b>2</b> )	Number of target word activation before/after cue (test-series <b>3</b> )	Number of correct target word activation before cue (sum)	Number of further correct target word activation after cue (sum)	Number of activation of incorrect target words (sum)
1: L-route, phono	8/5	6/6	10/3	24	14 -> 38	16
2: shortcut, phono	4/6	2/10	8/4	14	20 -> 34	20
3: L-route, semantic	6/7	4/5	4/4	14	16 -> 30 increase in correct productions	24

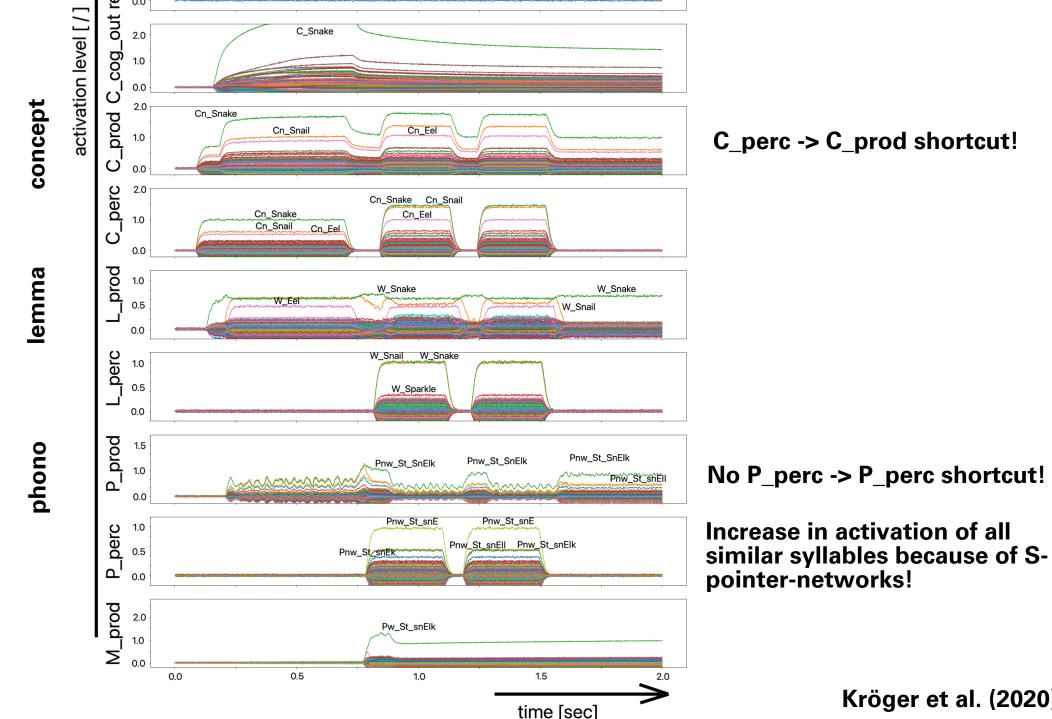
The maximum number of simulations per test series is 18. With a number of correct target word activation of 6, correct target word activation was not achieved in 12 out of 18 simulation cases. The total number of simulations per table row is 54. A distinction is made in this table regarding whether a target word is already correctly and completely activated in the motor plan state buffer before cues are given, i.e., within the first 750 ms of the simulation time, or whether the target word was only correctly activated after occurrence of cues. "L route" is model type "lexical route" without short-circuit at the phonological level; "Shortcut": model type with decoupled mental lexicon, but with short-circuit of the state buffers at the phonological level P\_perc -> P\_prod.

# The definition of scenarios

- Further speech screenings: (-> logopedic research)
  - Picture naming with phonological or semantic cues (acoustic)
    - Naming without cues in case of "word is on tip of my tongue" -> no or incorrect word activation at P\_prod after picture presentation
    - 4 cases:
      - Word activation starts with phonological cue (not earlier); 2 cases: "snake" and "apple"
      - ... starts with semantic cue (not earlier) 2 cases: "duck" and "fly"
      - fourth case: -> rare case ...



← priming: 100 msec:
"look and think" + 600 msec :
"utter noun"
+ later: "listen to cues" +
"utter noun"



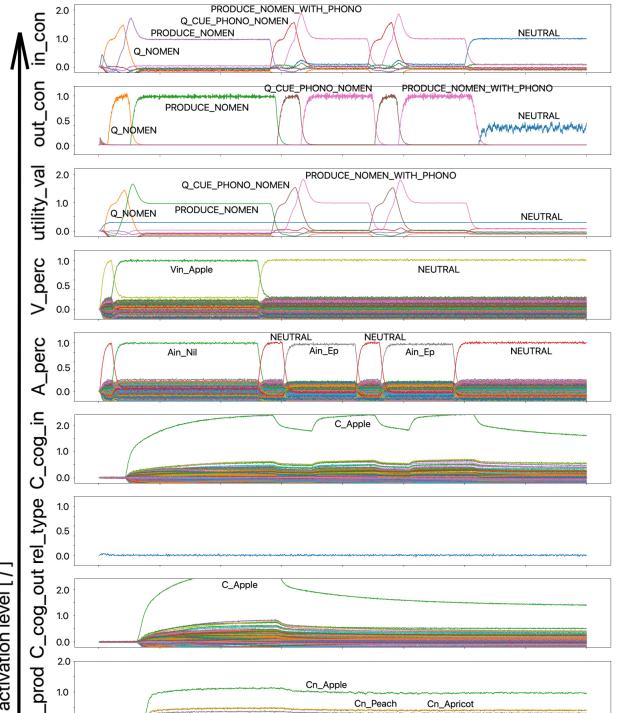
pathways



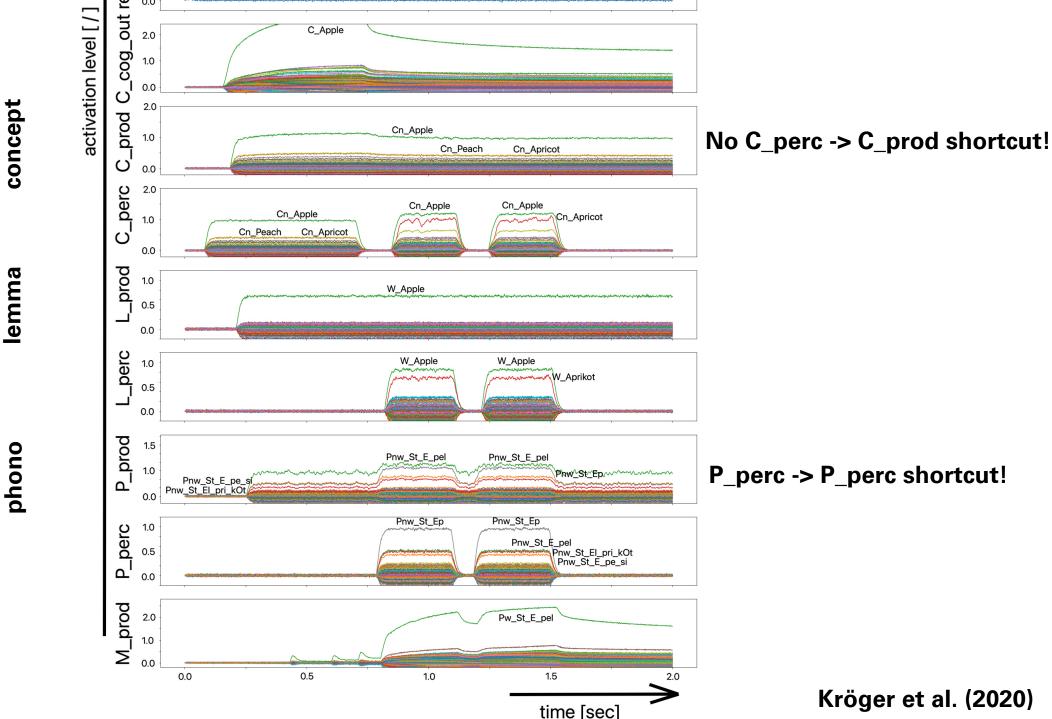


control

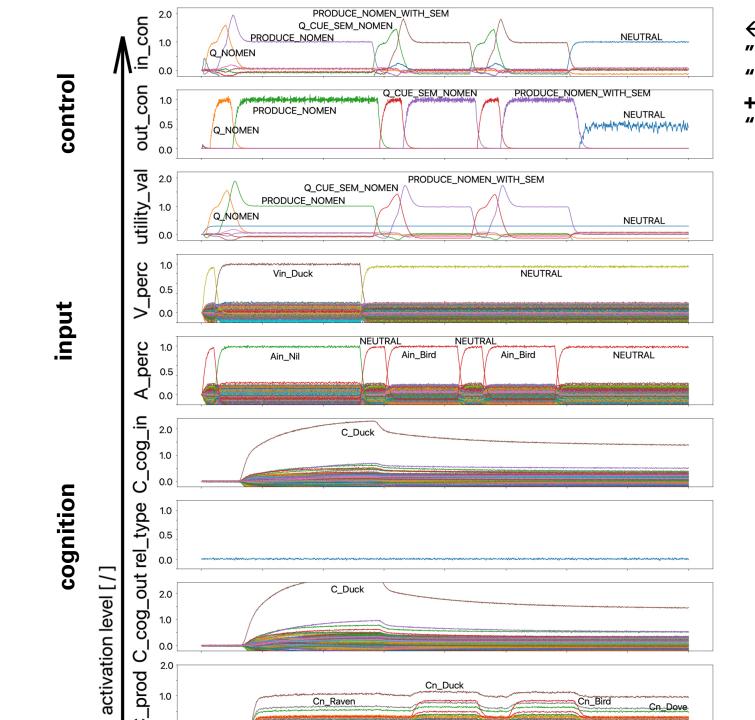
activation level [/]



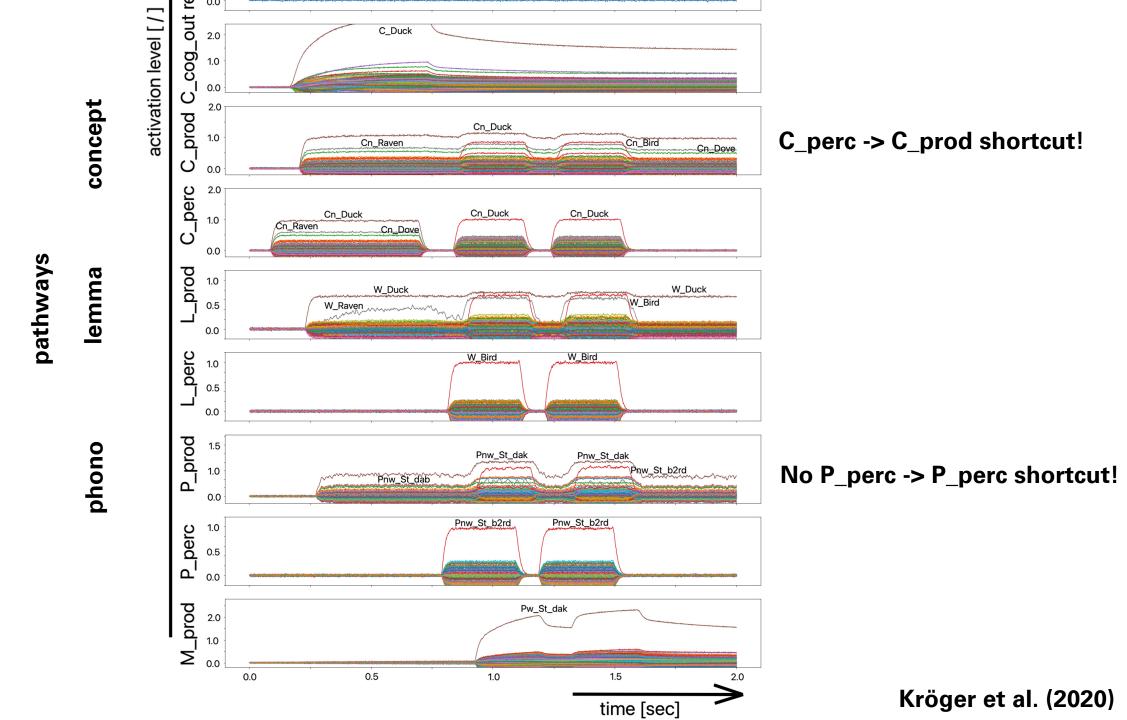
← priming: 100 msec: "look and think" + 600 msec : "utter noun" + later: "listen to cues" + "utter noun"



pathways

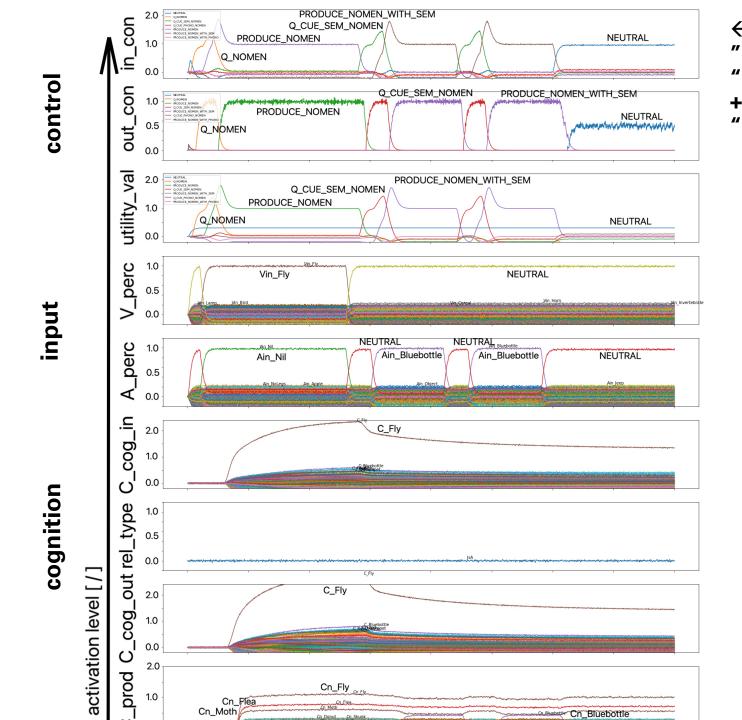


← priming: 100 msec:
"look and think" + 600 msec :
"utter noun"
+ later: "listen to cues" +
"utter noun"



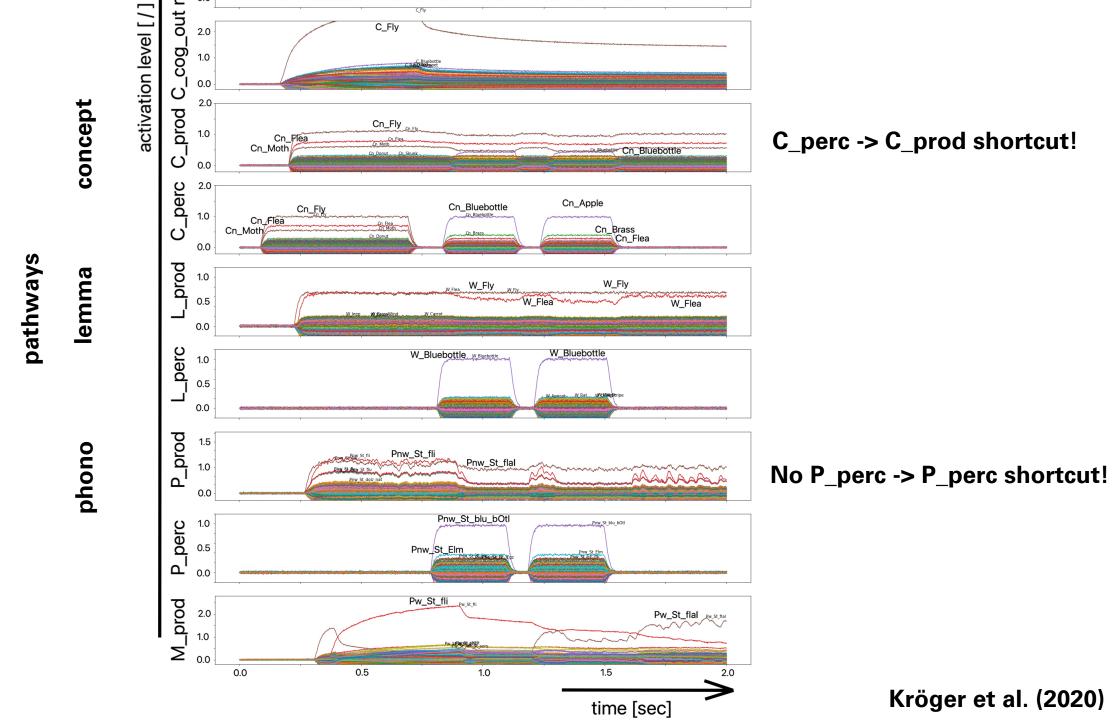
# The definition of scenarios

- Further speech screenings: (-> logopedic research)
  - Picture naming with phonological or semantic cues (acoustic)
    - Naming without cues in case of "word is on tip of my tongue" with cues:
    - 4 cases: ->
      - Word activation starts with phonological cue (not earlier); 2 cases: "snake" and "apple"
      - ... starts with semantic cue (not earlier) 2 cases: "duck" and "fly"
    - Fourth case: -> rare case
      - fourth case: A (semantic and) phonological similar word ("flea" (Floh )) will be activated first and corrected later by a semantic cue (("bluebottle" (Schmeißfliege), also: "meat fly"))
      - -> rare case: correction processes may occur in neural models! (realism!!!)



← priming: 100 msec:
"look and think" + 600 msec :
"utter noun"
+ later: "listen to cues" +
"utter noun"

Kröger et al. (2020)



# Realism of the model:

- Is the model (modeling of single neurons, of connections, of buffers, etc.) biologically "realistic"?
- Most biological realism for
  - the single neuron models (leaky integrate and fire) -> spiking
  - the cortico-cortical BG-Thalamus loop for process control
- But: using the NENGO-concepts of neuron buffers, memories, associative buffers, binding buffers S-pointer-networks etc. leads to a good modelling of all types of behavior occurring in human data

# Realism of the model:

- the large-scale model approximates human behavior surprisingly well in many cases: e.g.,
  - the serial recall task -> concave form: middle words have lower production probability (Choo 2010)
  - the rare cases; even for normal picture naming -> < 1% production errors or production fails (Kröger et al. 2016)
  - cues are helpful in case of mild mixed aphasia -> increase of correct word naming from 20-45% (no cues) to 55-70% (-> increase in performance)

# Realism of the model:

- the large-scale model approximates human behavior surprisingly well in many cases: e.g.,
  - Modelling symptoms (task performance) for different types of aphasia (-> decrease in performance with increasing strength of disorder)
  - Modelling picture naming with distractor words (-> decrease in performance (fails or errors) in case of occurrence of distractor words)
- so: in some cases: approximation of human data on quantitative level is moderate;
  - But all "human behaviors", all "effects" are modeled correctly (qualitative level)

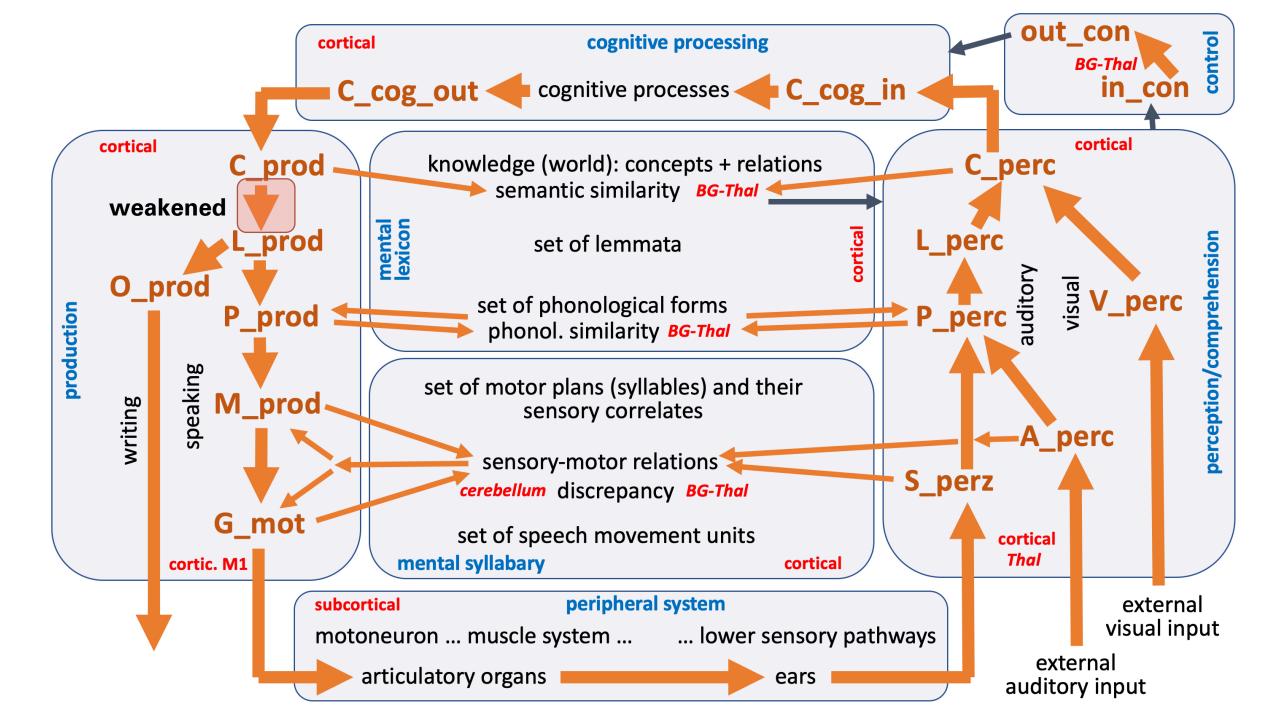
# Benefits from the model:

- Medicine / speech therapy research:
- Simulation of behavior of patients in speech screenings is quite close to reality (model data <-> human data)
- Creation of "simulated patients" with specific neural dysfunctions is already possible!
  - -> that allows checking of different versions of a screening with respect to its increase in sensitivity for detecting a specific neural dysfunction (a specific type of speech disorder)

# Addendum: Future work: https://www.youtube.com/watch?v=wz\_K Integration of articulatory-acoustic model

Download Video6:

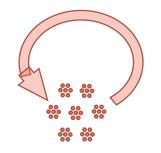
- Results shown so far -> cortical part of the model
- Further work: temporal control for syllable and speech gesture coordination -> Motor plans and their execution: time series of actions / gestures:
  - Vocalic gestures / actions: vocal tract form actions: /a/ , /i/, /u/,
  - Consonantal gestures / actions: oral full/near closing actions, labial, apical, dorsal ... -> /b/, /p/, /m/, ... /s/, /z/, ...
  - Glottal gestures / actions: phonation, opening actions (voiceless sounds), closing actions (glottal stop)
  - Velum: opening / closing actions of velopharyngeal port



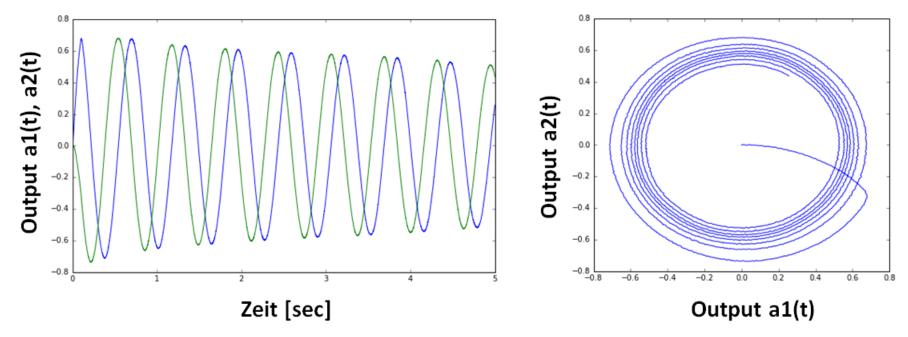
# Addendum: Future work: articulatory-acoustic model integration

- -> neuro-biological control concept for speech movements including generation of articulatory movements and acoustic speech signal
- Modeling of articulatory gestures, phasing concept for timing of gestures, ... -> part of quantitative "Articulatory Phonology" (Goldstein et al. USC, Haskins Labs)
- but our control concept is implemented as a part of a neurobiologically based brain model: neuron ensembles -> oscillators; based on spiking LIF-neurons, ... (differs from quantitative AP)

# Buffers as oscillators

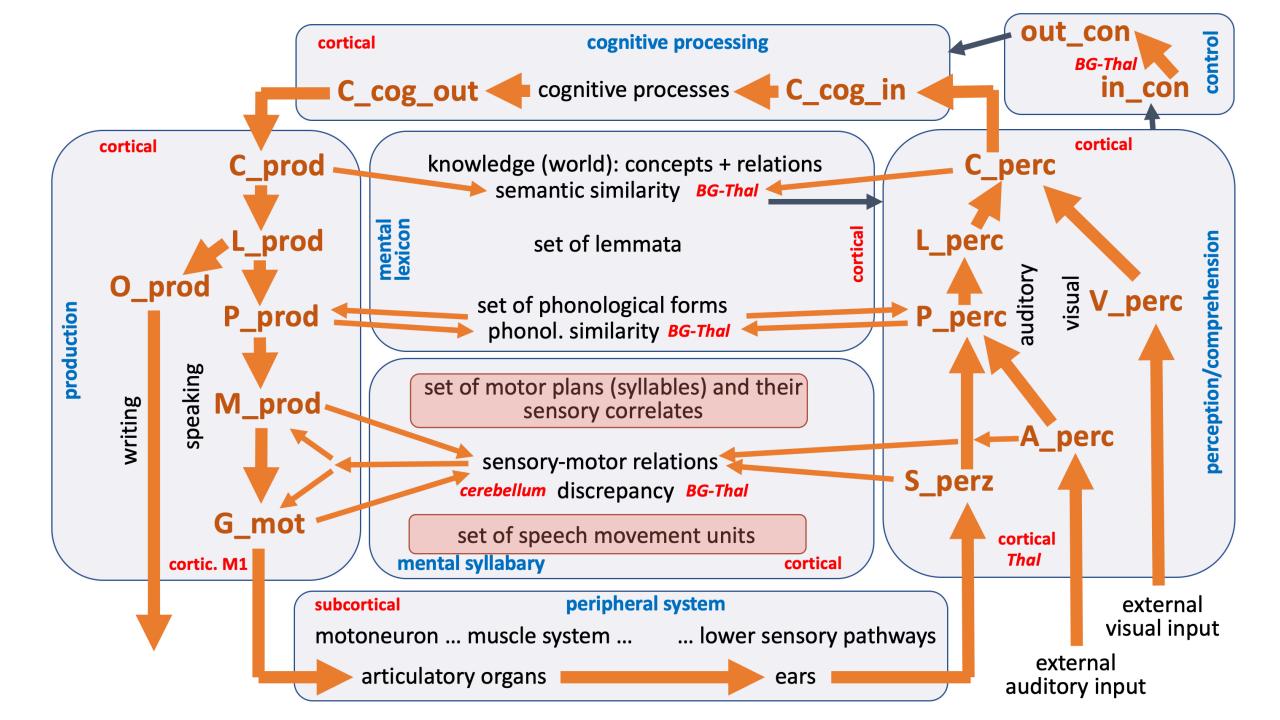


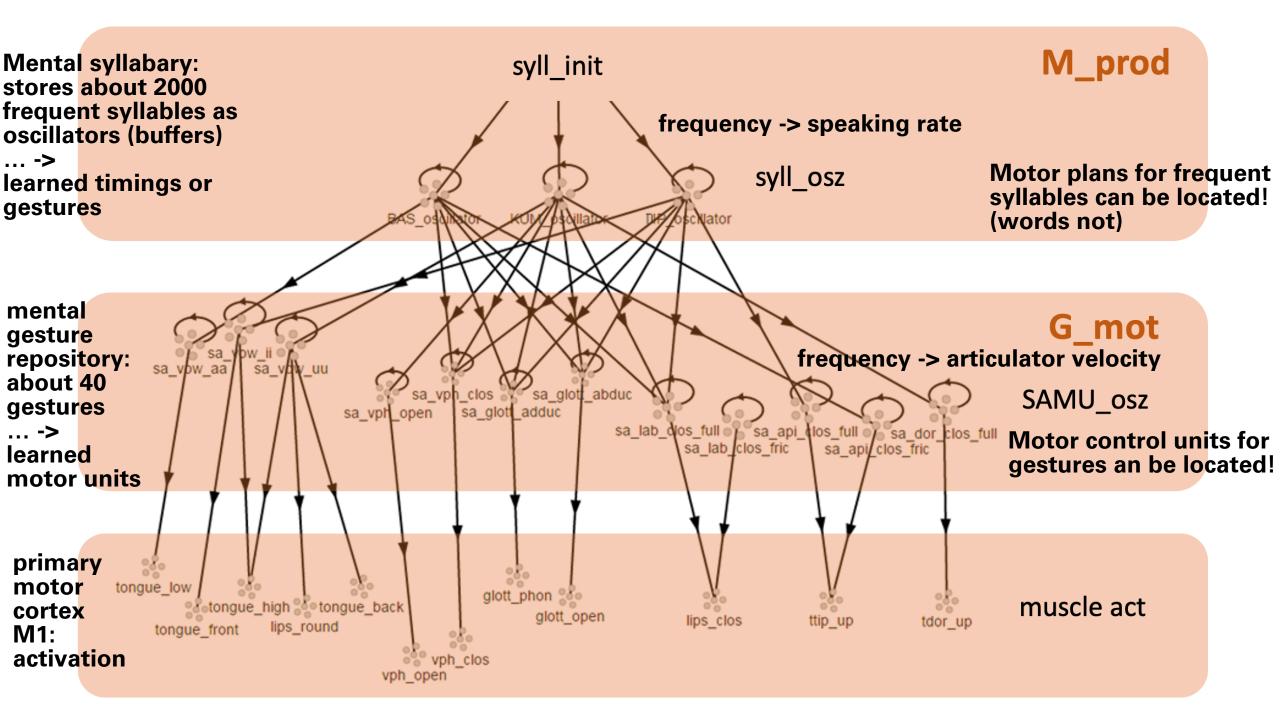
- Buffers with recursive neural connections -> short-term memories or oscillators
- activation transfer from 50% of the neurons to the other 50% of the neurons of the buffer and back ...



# The motor part of the neural model

- from phonological form to motor plans to a temporal series of gestures / speech actions :
- Syllable as basic unit:
- motor plans are stored as a whole "mental syllabary"
- Motor plans store the timing (phasing) of smaller units: gestures or vocal tract action units (= basic motor control units)
- Learning of motor plans? -> somatosensory and auditory feedback is important

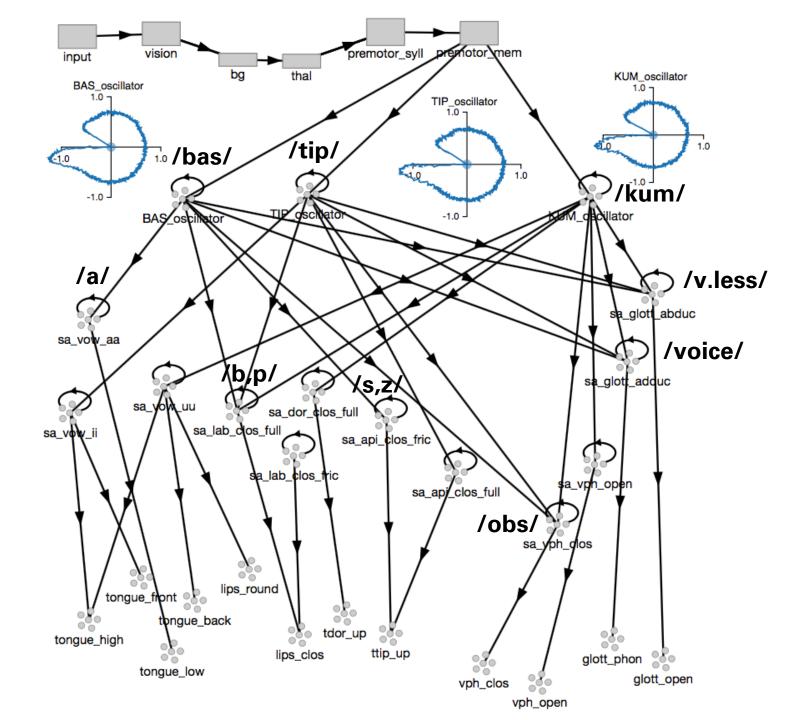




## Example word: /bas kum tip/

run Nengo\_gui →

(( or: run Nengo in python interpreter ))

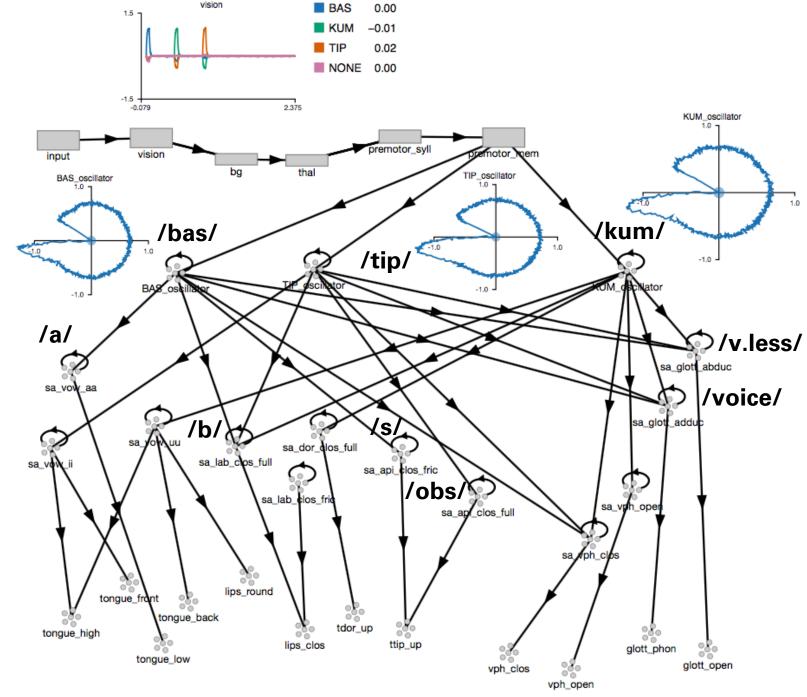


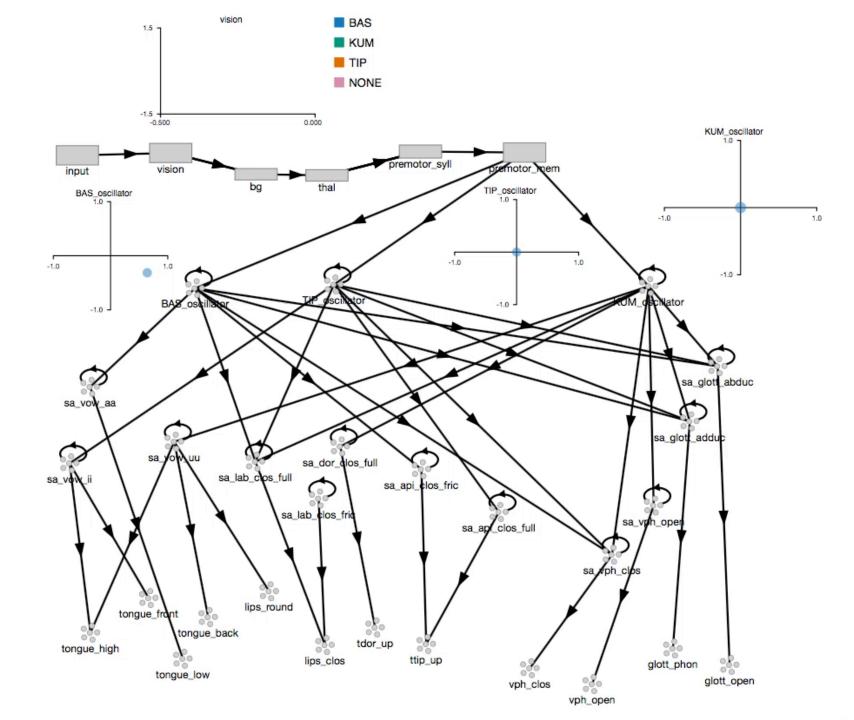
# Why oscillators?

- Oscillators define time intervals (<-> duration of one oscillation cycle)
- Oscillators define timing of actions:
  - actions may start at specific points in time within an oscillation cycle = phase value
  - Timing of articulation remains phase-stable even for different frequencies of oscillation (for different duration intervals for each syllable; for different speaking rates)

# Example word: /bas kum tip/

see the timing of the syllable oscillator activation: <u>Video1 start100</u>: simulation of three syllables



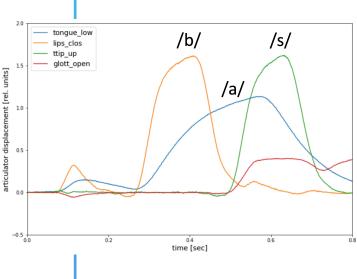


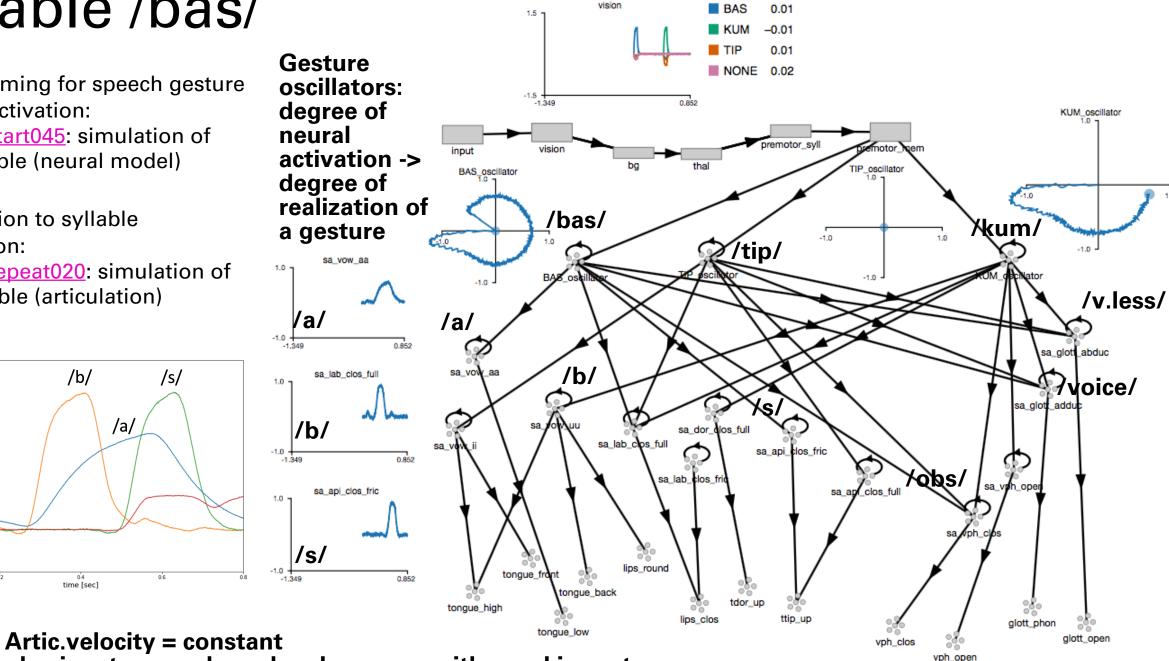
# Syllable /bas/

See the timing for speech gesture (action) activation: Video2 start045: simulation of first syllable (neural model)

Introduction to syllable articulation: Video<sup>B</sup> repeat020: simulation of

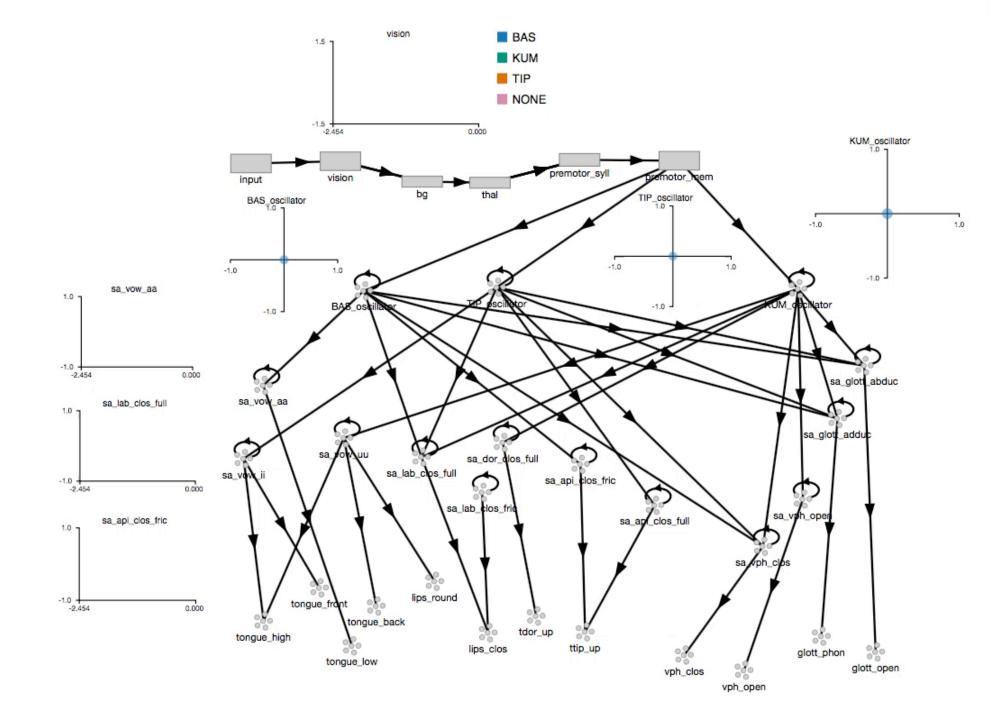
first syllable (articulation)



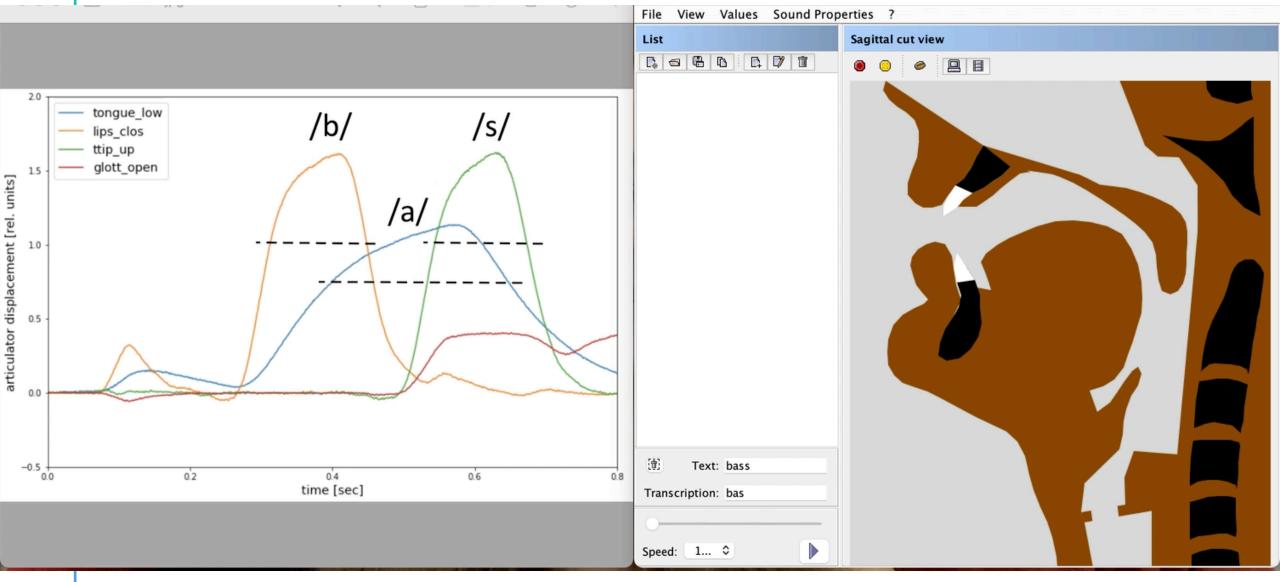


phasing: temporal overlap decreases with speaking rate





#### Gesture oscillators: degree of neural activation -> degree of realization of a gesture -> "abstract" movement trajectory



**2D-articulatory model** 

# Application of our approach: Modeling different speaking rates

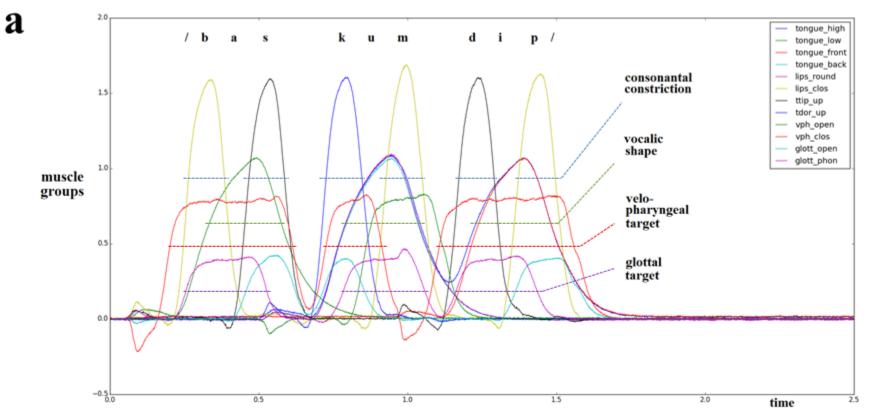
- phase values (motor plan / syllable level) stay constant
- oscillator frequency for gestures stays constant (differentiates vocalic and consonantal gestures)
- oscillator frequency of syllable oscillators (motor plan level) changes

Kroeger, BJ, Bekolay T, Blouw P, Stewart TC (2020) Developing a model of speech production using the Neural Engineering Framework (NEF) and the Semantic Pointer Architecture (SPA). Proceedings of the International Seminar on Speech Production ISSP2020. Yale University, New Haven, CT. <u>www.spechtrainer.eu</u> -> publications

### Results

- constant phasing of actions leads to correct production of speech sounds at each speaking rate:
- slow speaking rate:

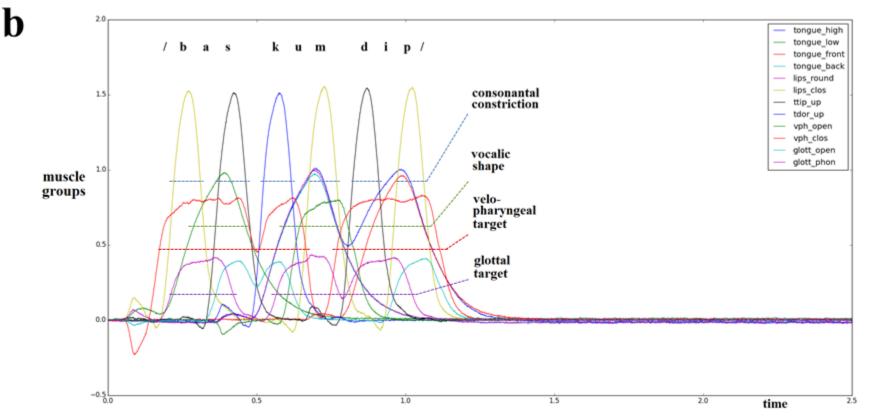
no increase in articulatory effort per speech action



### Results

- constant phasing of actions leads to correct production of speech sounds at each speaking rate:
- normal speaking rate:

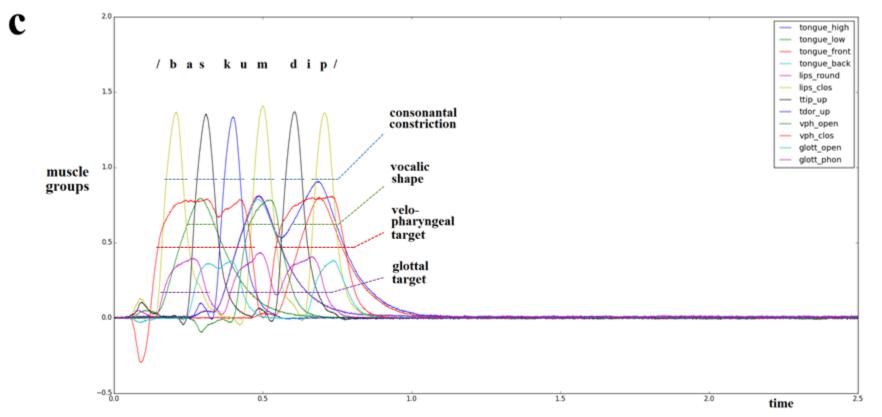
no increase in articulatory effort per speech action



### Results

- constant phasing of actions leads to correct production of speech sounds at each speaking rate:
- fast speaking rate:

no increase in articulatory effort per speech action



# Further work

- Further work: real-time integration of the articulatoryacoustic model into the neural brain model
- Integrating acoustic and somatosensory feedback allows motor (plan) learning
  - Timing of begin and release of constrictions / closures
  - Learning acoustic-motor relations
- -> implementation example for the articulatory model: see:
- SpeechArticulationTrainer (AppStore, GooglePlay)
- <u>https://www.youtube.com/watch?v=C09EYber\_T4</u>

