# Statistical information in (early) language learning

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# Some contextualization...





Language comprehension before the first word(s): speech perception and lexical development



# Language as a system

- Multimodal social behavior  $\rightarrow$  communicative function
- Limited inventory of signs that are recombined to encode meaning → symbolic function



### What needs to be learned?



# What is the problem?

The input is noisy and does not present clear-cut categories. Children have to deduce abstract representations from a broad range of overlapping concrete exemplars.



#### Word segmentation





#### Morphological regularities

Hund *dog* – Hunde *dogs* Puppe *doll* – Puppen *dolls* Kater *male cat* – Kater *male cats* Tochter *daugther* – Töchter *daugthers* 

#### Word order

....

Le petit prince (DAN) the little prince En infant terrible (DAN) a terrible child

# **Ressources to solve the learning problem**

countours, ...)

Innate biases

Scaffolding

Adult behavior that supports language learning (infantdirected speech, imitation, turn-taking, ...)

Predispositions that help children to attend to relevant

language, infant-directed communication, pitch

input, e.g. preference for socially relevant stimuli (faces,

Cognition



### **Principles**

General cognitive mechanisms supporting learning (categorical perception, associative learning, fast mapping, statstical learning, ...)



Domain-specific linguistic principles that facilitate learning (mutual exclusivity, Gricean maximes, universal grammar?)

# **Statistical learning**



→ finding a predictive function that maps input/output pairs based on a (limited) data set

→ often supervised learning with training on an annotated set and testing prediction on an untrained set





### Human learning (domain-general)

→ detect statistical regularities in the world to categorize and learn about the environment

→ mostly unsupervised, implicit learning (or reinforcement learning) with opaque input from which categories have to be derived



Linguistic forms are associated with specific frequency and co-occurrence information that may help to infer/learn linguistic categories.

### **Distributional information**

### Example: phonemic categories

Even though tokens of phonemes differ, they cluster around frequency peaks in specific acoustic dimensions. This distributional information can be exploited for category formation.



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### Transitional probabilities

#### Example: speech segmentation

Syllables within a word reliably cooccur, while syllables that span word boundaries have smaller transitional probabilities. This statistical information can help to find word boundaries (next to, e.g., prosodic boundary information).



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### **Co-occurrence information**

### Example: word learning

In interactions with young children, words tend to co-occur with their referents (e.g., during play or picture book reading). Across multiple situations this allows disambiguation of potential referents.



Linguistic forms are associated with specific frequency and co-occurrence information that may help to infer/learn linguistic categories.

### Non-adjacent dependencies

### Example: Tense marking

Linguistic elements are not randomly ordered and so a specific element allows to make predictions about upcoming elements. This dependency information can be used to learn morpho-syntactic rules.

#### Past tense:

He has <u>push</u>ed She has <u>pull</u>ed She has <u>move</u>d He has <u>call</u>ed

-> S/he h**as <u>X</u>-ed** 

# Evidence for different types of statistical learning in language acquisition

#### Distributional information

(e.g., phonological categories)



**Co-occurrence information** (e.g., word learning)



**Transitional probabilities** 

(e.g., speech segmentation)

The -> gui -> tar	<sup>.</sup> -> is
ball	looks
cookie	sounds

. . .

(Non-) adjacent dependencies (e.g., word order, grammatical markers)

[[Determiner + [Noun+PL]] ... ... has .... -ed / ... is ...-ing

# Learning phonological categories

Infants (and other primates) perceive speech sounds categorically. And they are sensitive to a wide variety of speech sound contrasts from birth.

Within the first year of life, they attune their perception to the native language, i.e., they improve their perception of native contrasts while they loose sensitivity to (most) nonnative contrasts.

 $\rightarrow$  Distributional statistics as a key to detect relevant acoustic distinctions.



e.g. Kuhl et al., 20066; Werker & Tees, 1998

### **Distributional phonetic information**

### **Preferential listening experiment**

24 American-English 6-month-olds &24 American-English 8-month-olds

2.5 minutes familiarization with six blocks of 24 syllables from the da-ta continuum with either *unimodal* or *bimodal* distribution with respect to VOT; colorful flowers as visual stimulus



Maye et al. (2002)

# **Distributional information**

### **Preferential listening experiment**

Test phase presenting 4 *alternating* and 4 *non-alternating* trials, half presenting tokens 1 and 8 (endpoints) and tokens 3 and 6 (category boundaries) alongside visual checkerboard pattern



- $\rightarrow$  Infants only discriminate da/ta after listening to bimodal distribution
- → Distributional information in the input modulates sound discrimination, i.e. statistical information in the input provides a mean to learn phonemic categories (even with only limited short-term exposure).

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# Distributional learning as domain-general mechanism...

### **DISCRIMINATION OF OBJECTS**

Method: eye tracking 4 alternating (token 1 and 8)/ 4 non-alternating (token 3 or 6) test trials

Age group: Dutch 9-to 11-month-olds

Result: discrimination after bimodal, but not unimodal familiarization



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# Distributional learning as domain-general mechanism...

### **DISCRIMINATION OF FACES IN INFANTS**

Method: ERPs, repetition suppresion paradigm

Age group: 6.5-month-olds

Result: discrimination after bimodal, but not unimodal familiarization

16 14 12 10 ž 0 5 10 15 20 25 30 35 40 45 50 55 60 65 70 75 80 85 90 95 100 Morphing steps



Altvater-Mackensen et al. (2017)

### ... with limits

### **DISCRIMINATION OF FACES AND PHONEMES IN ELDERLY**

Method: similar procedure in familiarization but forced choice task for test, stimuli were syllables (as in Maye et al., 2002) and faces (as in Altvater-Mackensen et al., 2017)

Age group: young (18-24 years) and old (60-84 years) healthy adults

Result: no evidence of learning in either modality without an explicit task

→ Distributional information might be easier to pick up for infants than adults (maybe because learning is less automatic and more top-down regulated in adults).

# Detecting words in the speech stream

Around 7 to 9 months of age, infants are able to extract (and memorize) words from a continuous speech stream

They exploit different cues to word boundaries, such as typical stress patterns, (il)legal consonant clusters and already familiar words.

 $\rightarrow$  Transitional probability on syllable combinations as a further key to detect common syllable sequences = words.

Feet: The feet were all different sizes. This girl has very big feet. Even the toes on her feet are large. The shoes gave the man red feet. His feet get sore from standing all day. The doctor wants your feet to be clean.
Bike: His bike had big black wheels. The girl rode her big bike. Her bike could go very fast.





e.g. Juscyzk & Aslin (1996); Altvater-Mackensen & Mani (2013)

# **Transitional probabilities**

### Artificial language learning experiment using HTPP

24 American-English 8-month-old

2 minutes of familiarization to continous speech stream (fade in and out, no pauses, no prosody): concatenation of four three-syllable nonsense words *bidaku*, *padoti*, *golabu*, *tupiro* 



Saffran et al. (1996)

# **Transitional probabilities**

Exp 1: test phase with two trials each consisting of *familiarized* words (*bidaku, padoti, golabu, tupiro*) *novel* syllable orders (*dabiku, dopati, lagobu, pituro*)

Exp 2: test phase with two test trials each with *familiarized* words (*bidaku, padoti, golabu, tupiro*) *part-words* (*daku-pa, doti-go, labu-tu, piro-bi*)



- → Longer looking to novel/part-words than familiarized words, indicating that infants can discriminate novel word forms from exposed word forms
- → Infants use transitional probability information to extract words from the speech stream.

# Transitional probabilities as domaingeneral (and age-robust) mechanism...

### **TRIPLET LEARNING IN AUDITORY AND VISUAL MODALITY**

Method: familiariziation similar to Saffran et al. (1999), but behavioral forced choice task for test

Age group: young (18-24 years) and old (60-84 years) healthy adults

Result: discrimination in both modalities, but stronger effects for auditory than visual domain and less pronounced effects in elderly





Fiser & Aslin (2001); Schevenels et al. (2023)

# Learning form-meaning pairs

By 6- to 9-month of age infants show first recognition of highly frequent words and accumulate a sizable vocabulary within the following months.

Remarkable as word-meaning mappings are largely arbitrary and naming events are mostly ambiguous (cf. *Quine's problem*), even though the child's view might often be less crowded than the environment suggests.

 $\rightarrow$  Cross-situational co-occurrence of word and referents as an important source for lexical development.

e.g., Yurovsky et al. (2013); Bergelson & Swingley (2012)





# **Tracking of co-occurrence information**

### Word learning task

28 American-English 12-month-olds &27 American-English 14-month-olds(other study successfully tested adults)

Familiarization with 30 slides presenting two shapes on each slide and labelling one of them (drawn from a set of six wordobject pairs)



Smith & Yu (2008)

# **Tracking of co-occurrence information**

### Word learning task

12 test trials presenting each object-word combination twice, using preferential looking procedure with proportion of target looking as dependent variable





- $\rightarrow$  Longer looking to labelled object
- → Infants can use cross-situational cooccurrence information to track (and learn) word-object pairings.

# Tracking of co-occurrences in more natural learning situations ...

### WORD LEARNING FROM PICTURE BOOKS IN TODDLERS

Method: natural exposure in book reading; preferential looking task

Age group: 2- to 4-year-olds

Result: successful learning of novel word-object pair (name of microbe) from crowded display; better learning for toddlers that showed higher attention during reading

Altvater-Mackensen (2021)









# **Evidence for different types of statistical learning in language learning**

### Distributional information

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**Co-occurrence information** (e.g., word learning)



### **Transitional probabilities**

(e.g., speech segmentation)

The -> gui -> tar -> is		
ball	looks	
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. . .

- How can infants keep track of statistical regularities?
  - (in particular if they do not yet know what is the relevant unit of representation)
    - Why do we see differences in performance depending on age?

# **Conceptualizations of learning**

### $\begin{array}{c} \rightarrow \\ \leftarrow \end{array}$ Simple heuristics

- Take every syllable to be a word
- Associate what is co-occuring
- Detect repetitions
- Focus on salient parts of input
- Allow only one meaning for each word (and only one word for each meaning)

### $oldsymbol{ ho}$ Hypotheses testing

- Take first occurrence and check it against the next occurrence. Update if different, strengthen if similar.
- Assume a rule/regularity and tolerate deviation as long as exceptions are not exceeding a certain treshold (cf. Yang's *tolerance principle*)

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• ...

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# Attention and brain oscillation?

Attention is dynamic and synchronizes with sensory information (cf. DAT theory).

This might be reflected in brain oscillations that align with regularities in the input at different levels, which might in turn help to track these regularities (cf. *neural entrainment*).



Jones (1976)

# **Processing and brain maturation?**

Infants seem to be better implicit statistical learners than adults who often need a task.

This might reflect differential impact of bottom-up vs. top-down information in processing, which might in turn be related to the maturation of prefrontal cortex as well as fiber tracts connecting lower- to higher-level processing areas of the brain.



Skeide & Friederici (2016); Gogtay et al. (2004)

# Some open questions/problems...

• Work with complex/more natural stimuli is very limited (e.g. transitional probabilities often reduced to learning of doublets/triplets and a limited inventory)

### ightarrow ecological validity

 It is largley unclear how statistical cues are weighed/integrated with other cues (some work, e.g. showing that prosody wins out over stats)

### $\rightarrow$ comprehensive models

Recommendation for further reading: Conway (2020). *Neuroscience and Biobehavioral Reviews, 112, 279–299*. Frost et al. (2019). *Psychological Bulletin, 145(12),* 1128-1153.

