



# There are patterns in child-directed speech that make first language learning possible

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SWISS NATIONAL SCIENCE FOUNDATION



**How do children learn  
language?**







# Recursion





# Recursion



# Rule-based learning

**house**

**houses**

**child**

**children**

**kid**

**kids**

**mouse**

**mice**

**dog**

**dogs**

**foot**

**feet**

# Rule-based learning

## German

Table 1: The Eight German Plural Classes

affix	-s	-(e)n	-e	-er	zero
- Umlaut	<i>Auto-s</i>	<i>Bahn-en,</i> <i>Fahne-n</i>	<i>Fisch-e</i>	<i>Kind-er</i>	<i>Lehrer-0</i>
+ Umlaut	--	--	<i>B<u>ä</u>nk-e</i>	<i>R<u>ä</u>d-er</i>	<i>K<u>ä</u>sten-0</i>

Behrens 2002

## Dinka

<sup>ˠ</sup> SINGULAR	PLURAL	<sup>ˠ</sup> GLOSS	
cɿn	cìn	‘hand’	(marked singular)
jɛ̃ɛc	jè̃eɛc	‘belly’	(marked plural)
tùuk	tùuk	‘fruit of palm’	(both singular and plural marked)



# Beyond structure



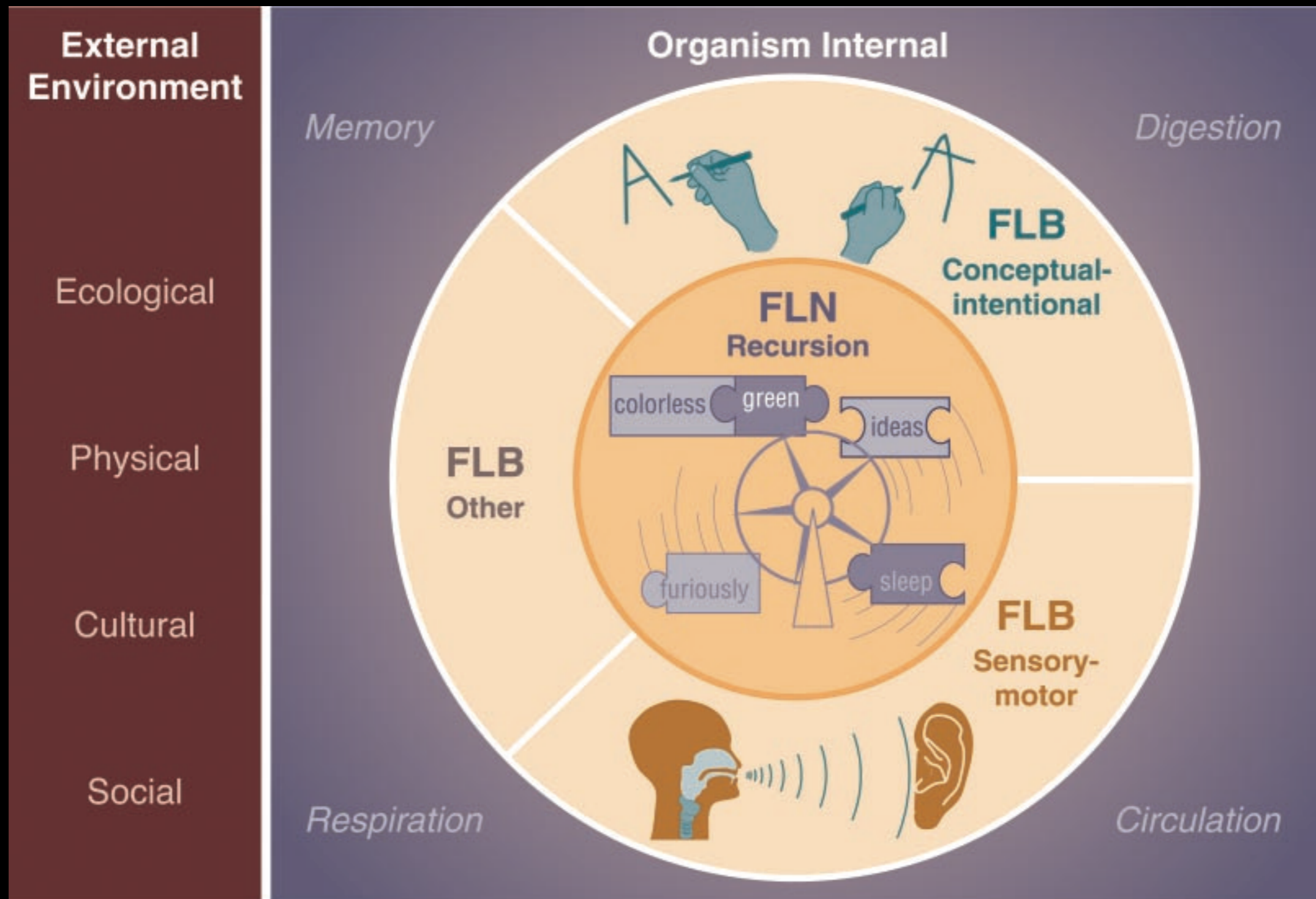
# Beyond structure



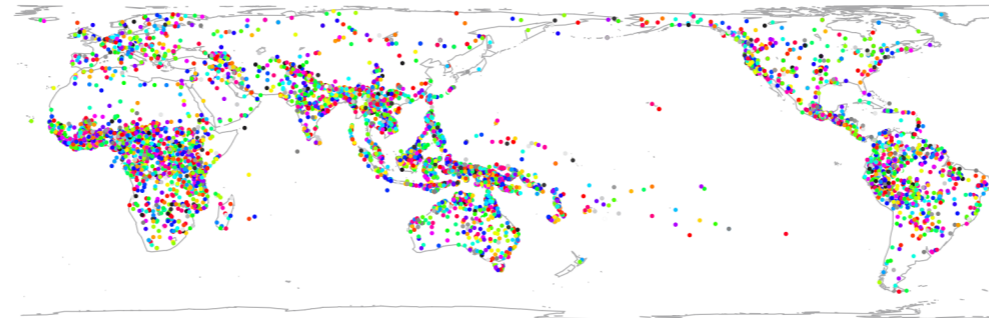
# Beyond structure







# Learning mechanisms



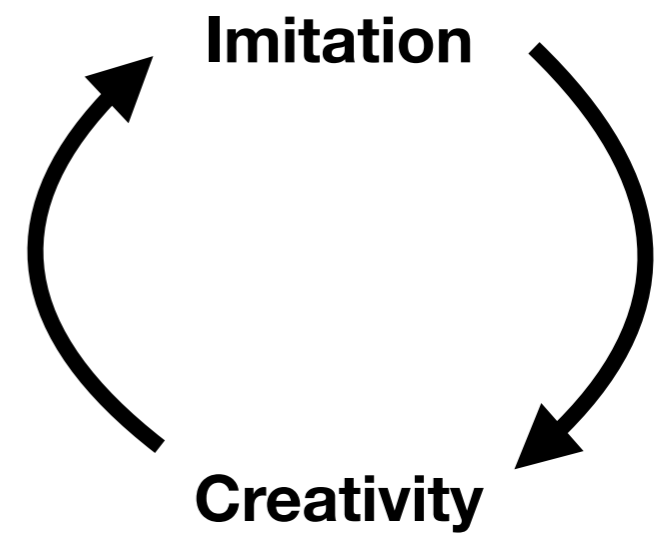
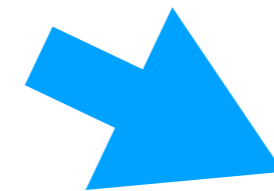
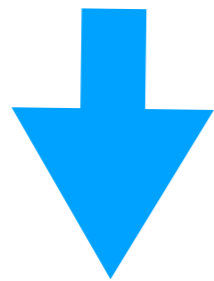
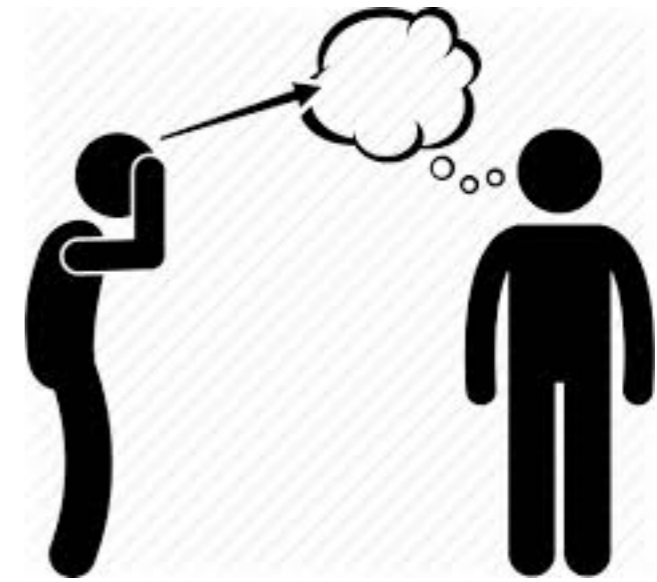
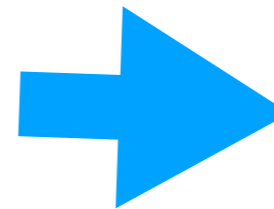
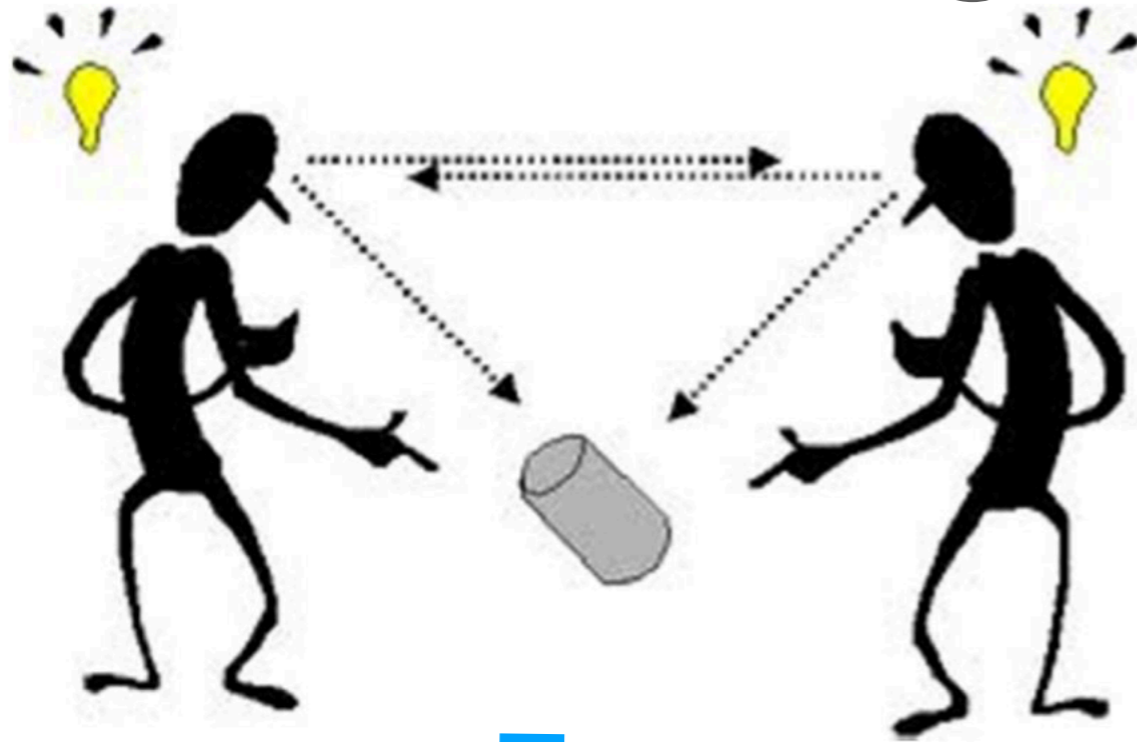
Socio-  
cognitive  
abilities



Experiments

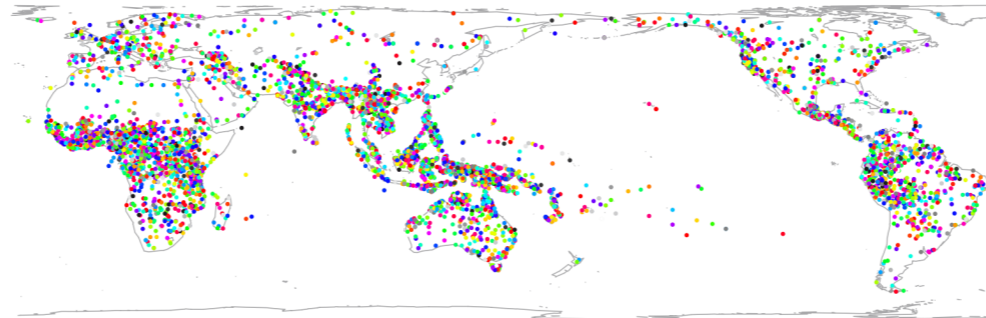


# Learning mechanisms

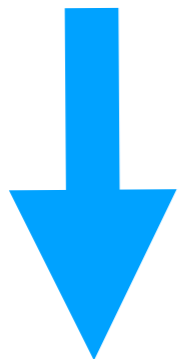




# Learning mechanisms



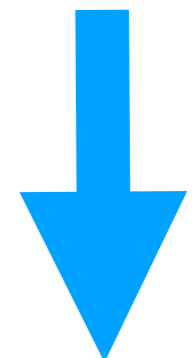
Socio-  
cognitive  
abilities



Experiments



Pattern-  
detection  
abilities



Experiments

# How is language learning possible?

- **Study 1:** An extreme case of language learning: A polysynthetic verb system.
- **Study 2:** Variation sets: How to learn about constructions and meaning.
- **Study 3:** Structural frames: how to detect nouns and verbs
- **Study 4:** Temporal cues: how to detect nouns and verbs



**Balthasar Bickel**



**Damian Blasi**



**Jekaterina Mazara**

# Study 1: An extreme case of language learning: thousands of verb forms in Chintang



# Verbal morphology

## Chintang

English  
 I walk  
 you walk  
 he walk**s**  
 we walk  
 you walk  
 they walk  
 walk**ed**

	1s	1di	1pi	1de	1pe	2s	2d	2p	3s	3ns	intransitive	
1s						tupnaʔā tupnaʔāniŋ tupnehē matupyoknehē	tupnaʔāce tupnaʔāceŋiŋ tupnace matupyoknace	tupnaʔāni tupnaʔāniŋiŋ tupnanihē matupyoknanihē	tubukuj tubukujniŋ tubuhē matupyoktuhē	tubukujcuŋ tubukujcuŋniŋ tubujcihē matupyoktucjihē	tupmaʔā tupmaʔāniŋ tubehē matupyoktehē	
1di						tupcoko tupcokoniŋ tubace matupyoktace	tubumcum tubumcumnim tubumcumhē matupyoktumcumhē	tupceke tupcekeŋiŋ tubace matupyoktace				
1pi						tubukum tubukumnim tubumhē matupyoktumhē	tubiki tubikiŋiŋ tubihē matupyoktihē					
1de						tupnaʔānciyā tupnaʔānciyāniŋ tupnanciyehē matupyoknanciyehē	tupcokoya tupcokoyaniŋ tubacehē matupyoktacehē	tubumcumma tupcekejanij tubumcumhē matupyoktumcumhē	tupcekeja tupcekejanij tubacehē matupyoktacehē			
1pe						tubukumma tubukummaniŋ tubumhē matupyoktumhē	tubikiya tubikiyaniŋ tubihē matupyoktihē					
2s	atupmaʔā atupmaʔāniŋ atubehē {a-ma}tupyoktehē			{a-ma}tupceke {a-ma}tupcekeŋiŋ {a-ma}tubace {a-ma-ma}tupyoktace	{a-ma}tupno {a-ma}tupnikniŋ {a-ma}tube {a-ma-ma}tupyokte					atuboko atubokoniŋ atube amatupyokte	atubukuce atubukuceŋiŋ atubuce {a-ma}tupyoktuce	atupno atupnikniŋ atube {a-ma}tupyokte
2d	atupmaʔānciyā atupmaʔānciyāniŋ atubajcihē {a-ma}tupyoktaŋcihē			atupcoko atupcokoniŋ atubace amatupyoktace	atubumcum atubumcumnim atubumcumhē {a-ma}tupyoktumcumhē					atupceke atupcekeŋiŋ atubace {a-ma}tupyoktace		
2p	atupmaʔāniŋ atupmaʔāniŋiŋ atubajnihē {a-ma}tupyoktaŋnihē			atubukum atubukumnim atubumhē amatupyoktumhē	atubiki atubikiŋiŋ atubihē {a-ma}tupyoktihē							
3s	utupmaʔā utupmaʔāniŋ utubehē {u-ma}tupyoktehē	maitupceke maitupcekeŋiŋ maitubace {mai-ma}tupyoktace	maitupno maitupnikniŋ maitube {mai-ma}tupyokte	matupceke matupcekeŋiŋ matubace {ma-ma}tupyoktace	matupno matupnikniŋ matube {ma-ma}tupyokte	natupno natupnikniŋ natube {na-ma}tupyokte	natupceke natupcekeŋiŋ natubace {na-ma}tupyoktace	natubiki natubikiŋiŋ natubihē {na-ma}tupyoktihē	tuboko tubokoniŋ tube matupyokte	tubukuce tubukuceŋiŋ tubuce matupyoktuce	tupno tupnikniŋ tube matupyokte	
3d	utupmaʔānciyā utupmaʔānciyāniŋ utubajcihē {u-ma}tupyoktaŋcihē								utupcoko utupcokoniŋ utubace {u-ma}tupyoktace	utubukuce utubukuceŋiŋ utubuce {u-ma}tupyoktuce	utupceke utupcekeŋiŋ utubace {u-ma}tupyoktace	
3p	utupmaʔāniŋ utupmaʔāniŋiŋ utubajnihē {u-ma}tupyoktaŋnihē								utuboko utubokoniŋ utube {u-ma}tupyokte	utupno utupnikniŋ utube {u-ma}tupyokte		

*jo-go-yaŋ*

*na-khutt-i-ca-i-hatt-i-bir-i.*

whatever-NMLZ-ADD 3[s]>2-steal-2pO-V2:eat-2pO-V2:move.away.TR-2pO-V2:do.for-[SBJV.]2pO

‘It (a cat) may steal everything from you and eat it all up!’ [story.cat.204]

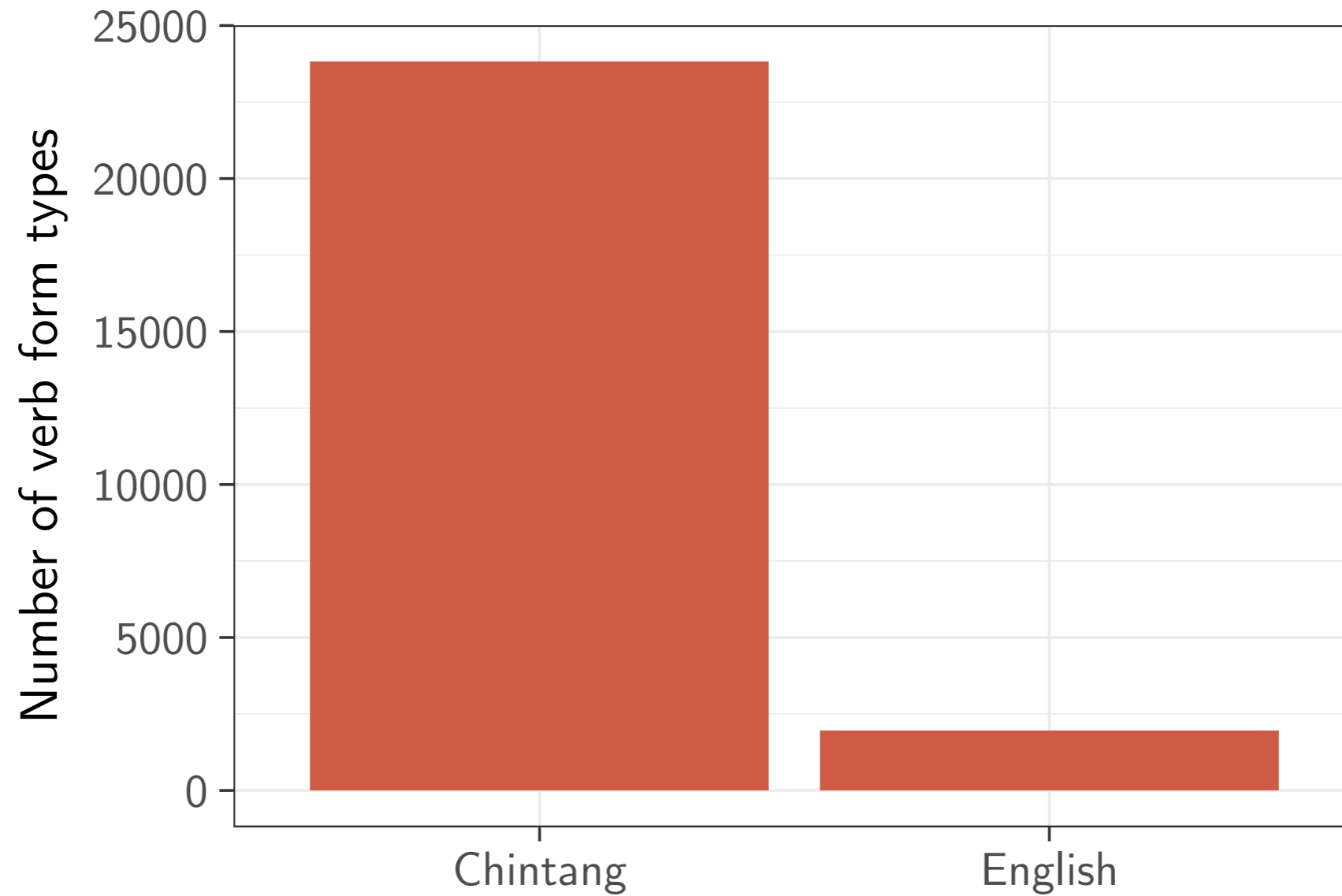
# Verbal structure

	<b>Affixes</b>	<b>Unique synthetic forms</b>
<b>English</b>	<b>3</b>	<b>3</b>
<b>Chintang</b>	<b>148</b>	<b>4745</b>

**How can such an extreme  
system be learned?**

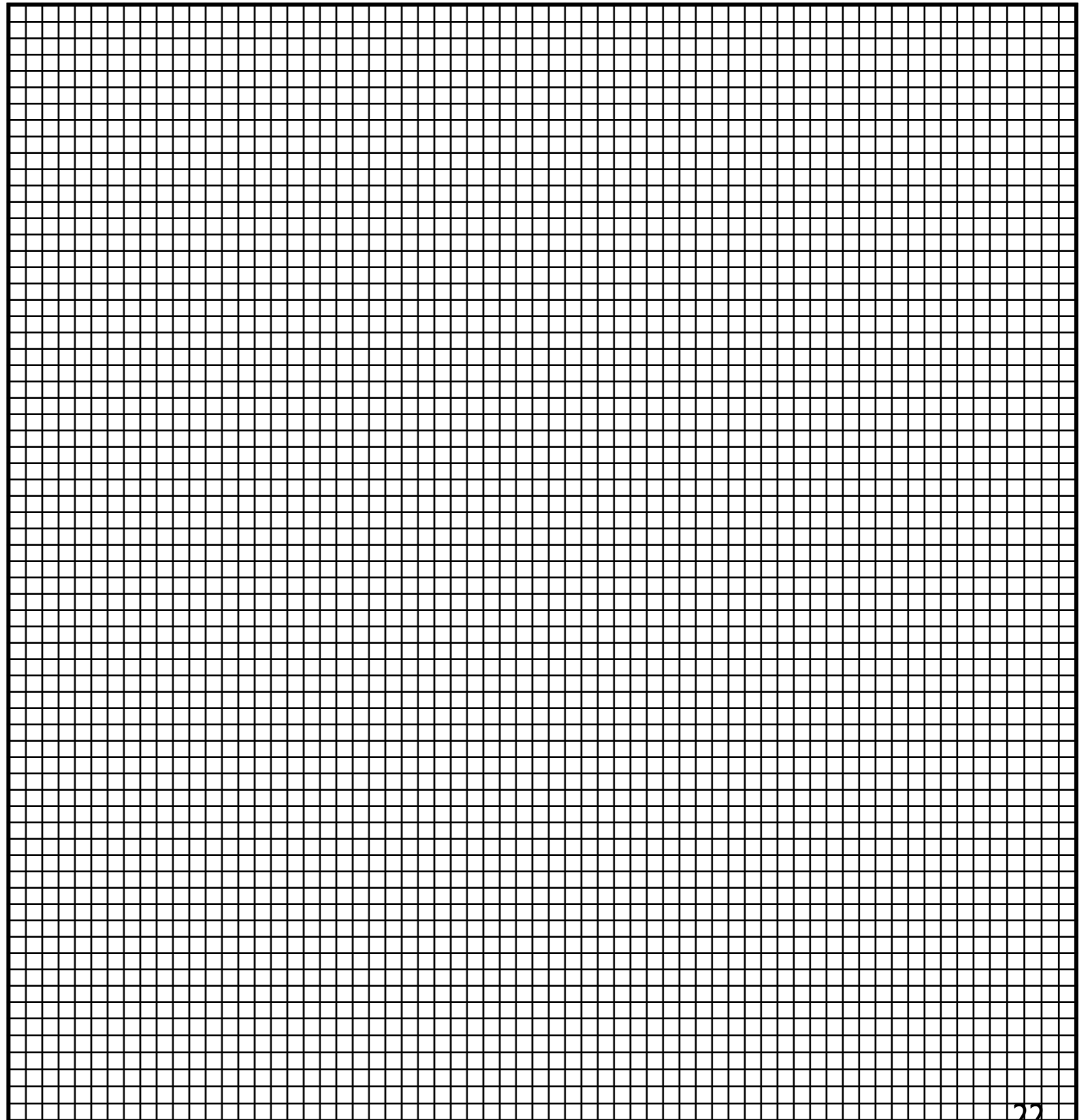


# Verb form types in the input



English

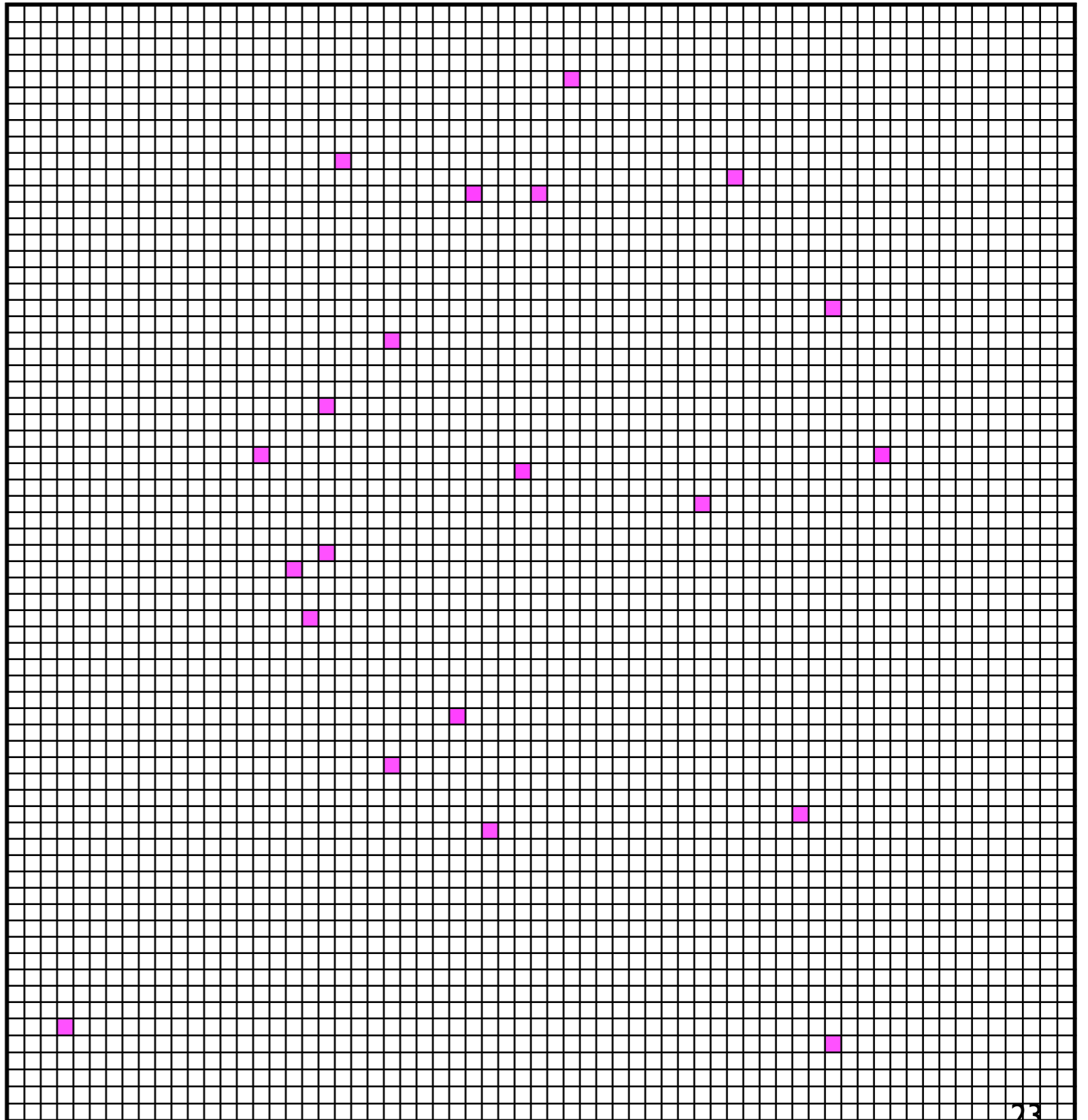
Chintang



# English



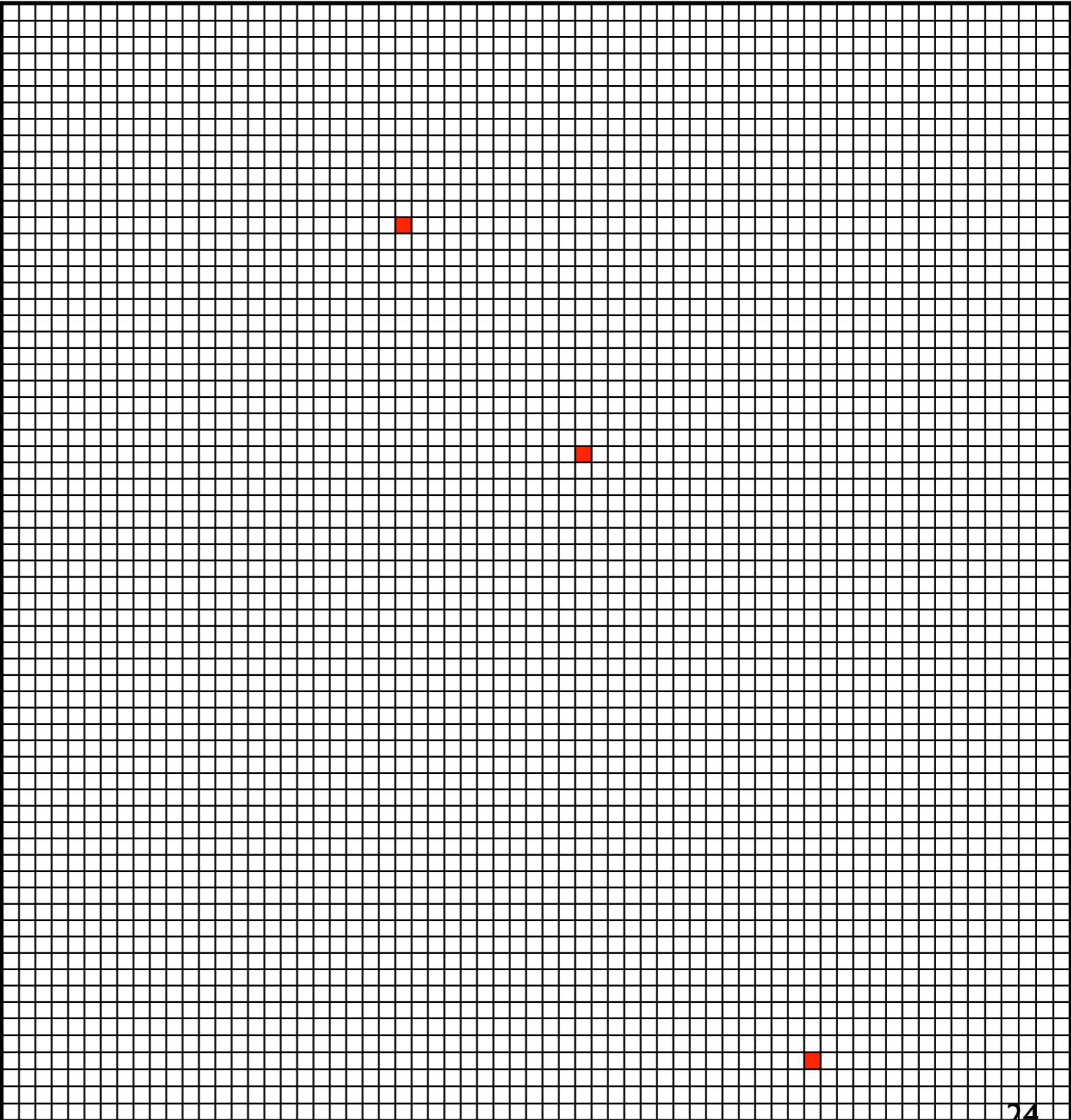
# Chintang





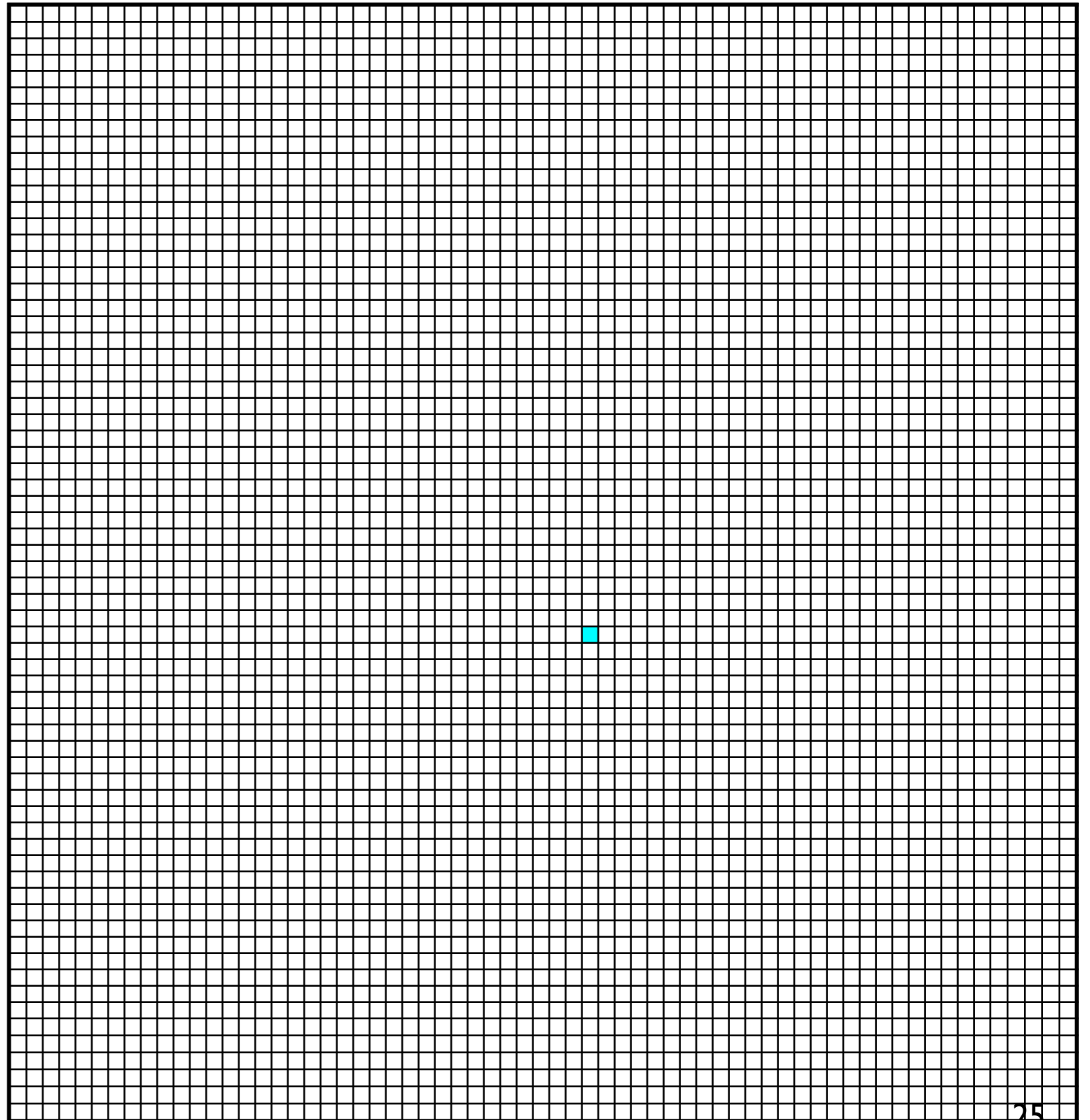
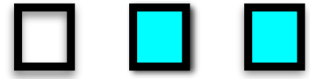
# English

# Chintang

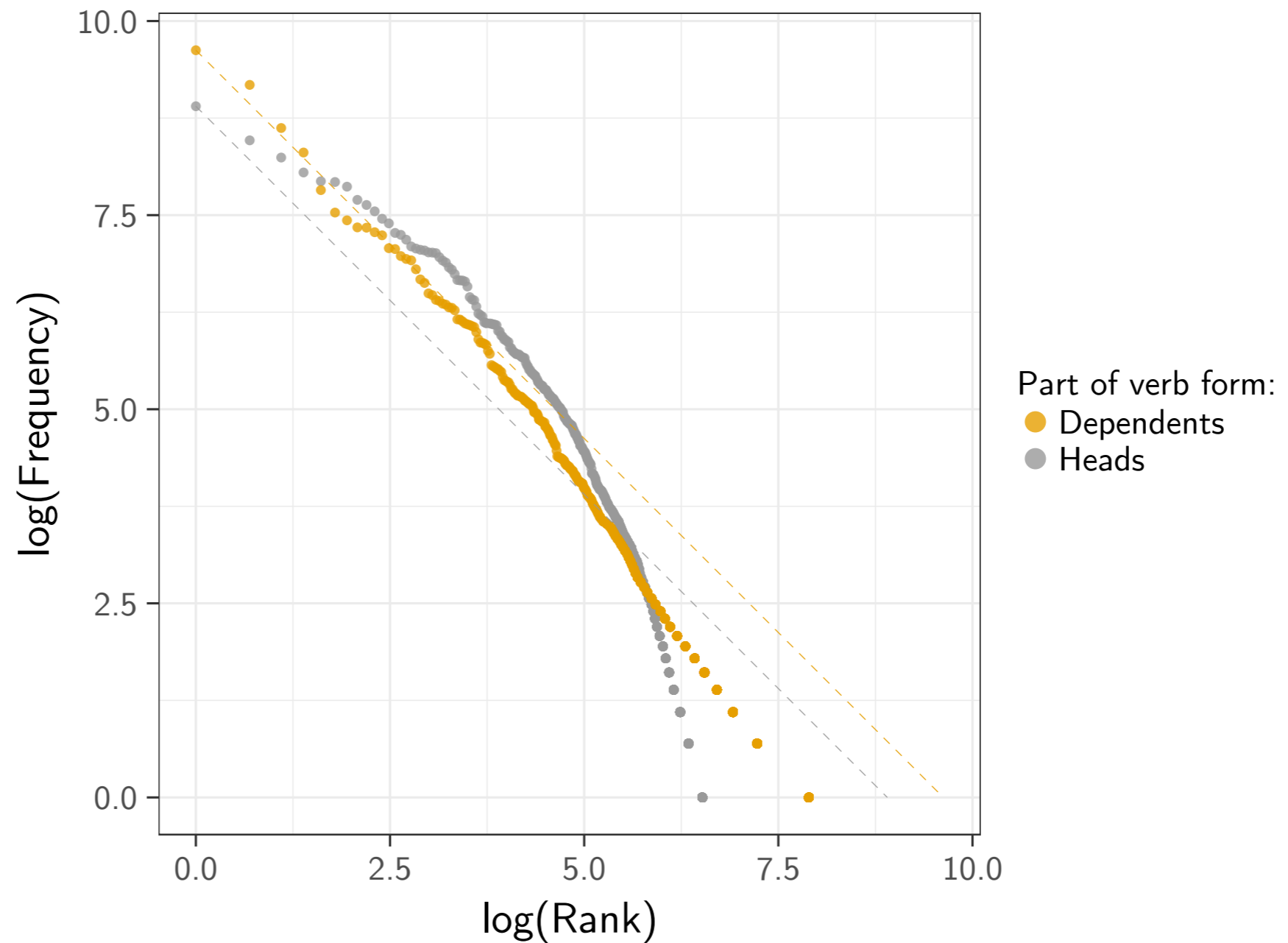


# English

# Chintang



# Frequency distributions stems vs. affixes: Chintang (input)



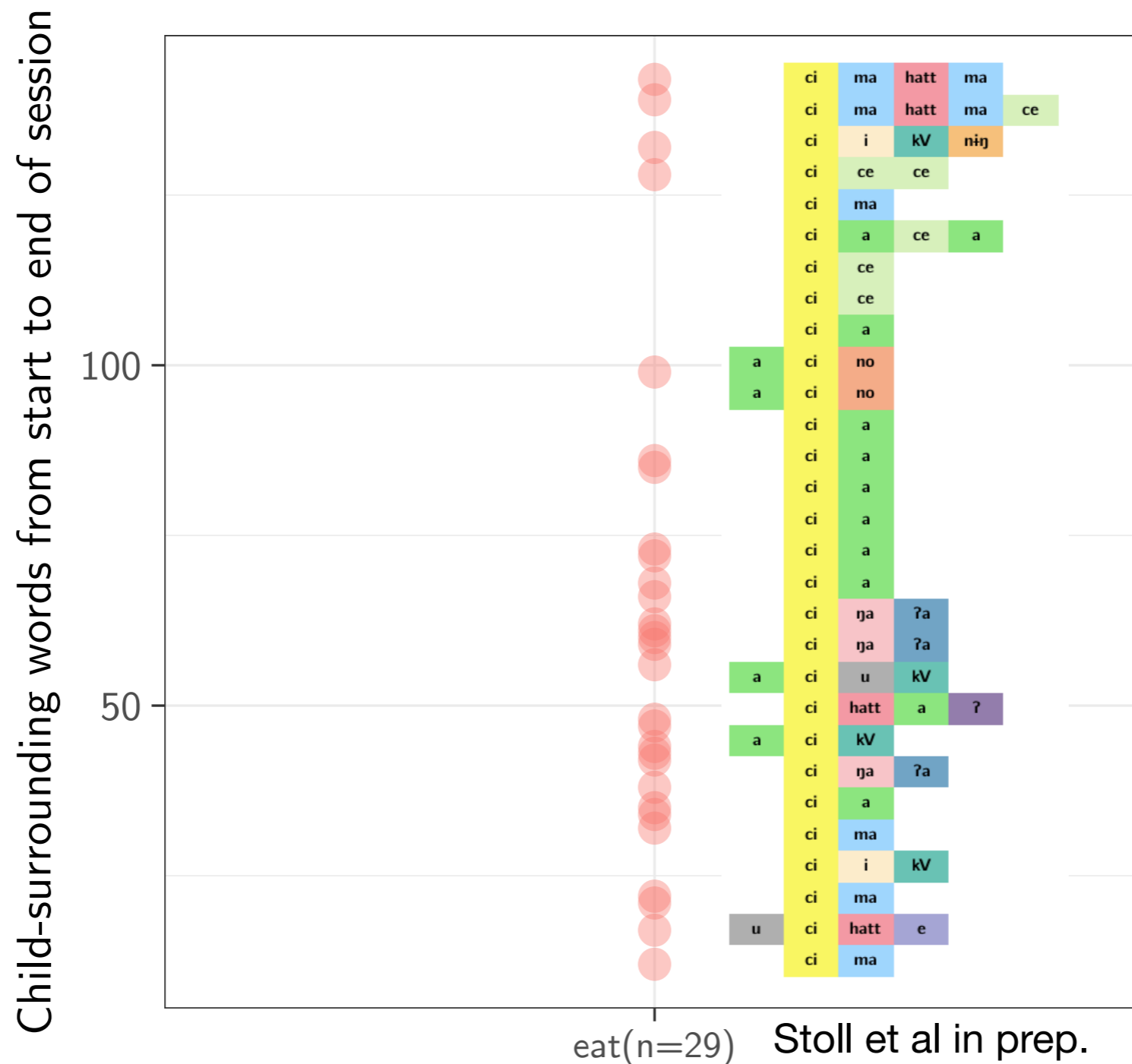


# Most frequent stems in one recording: Chintang (input)

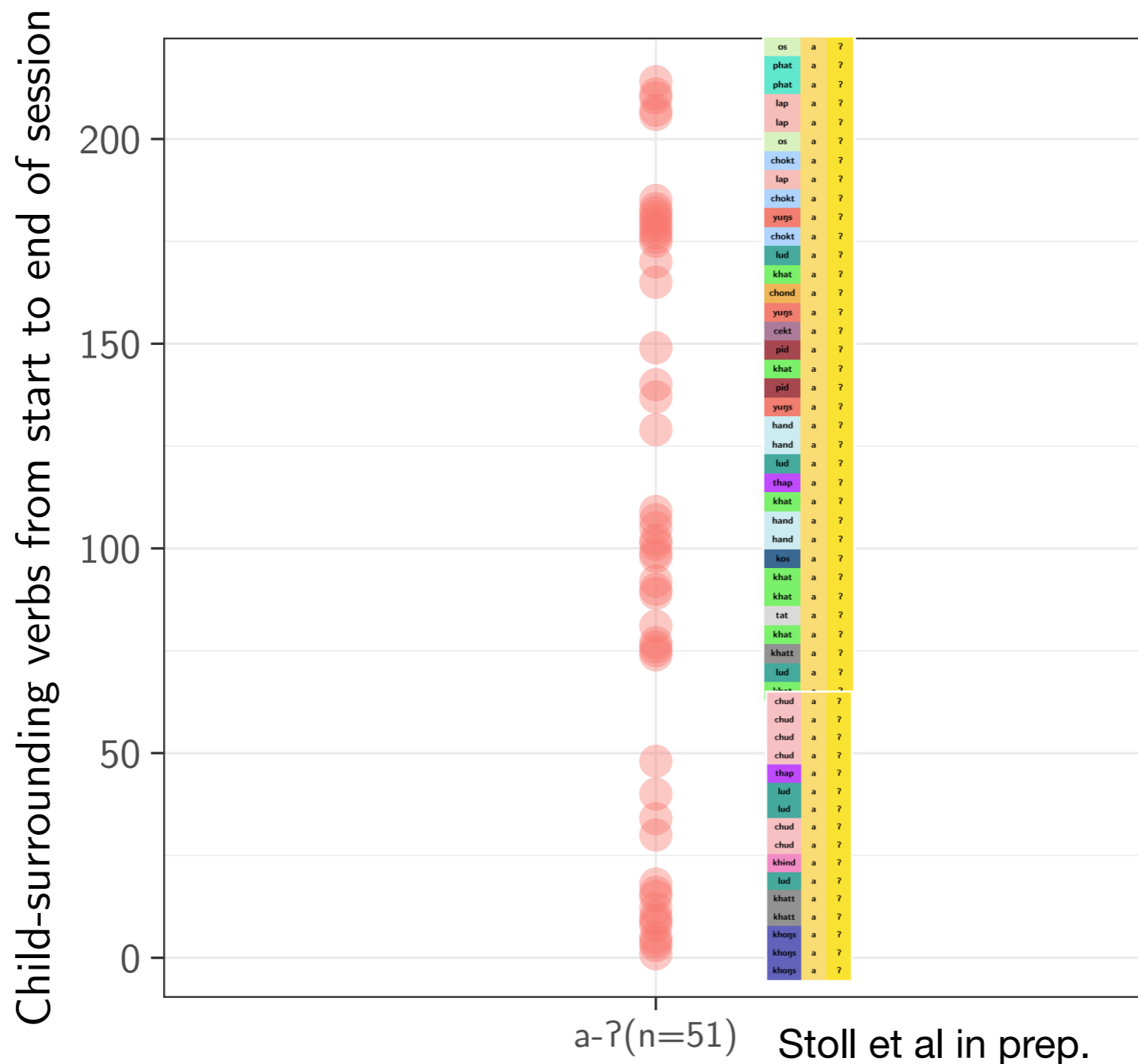


10 most frequent heads in adult production in sample session

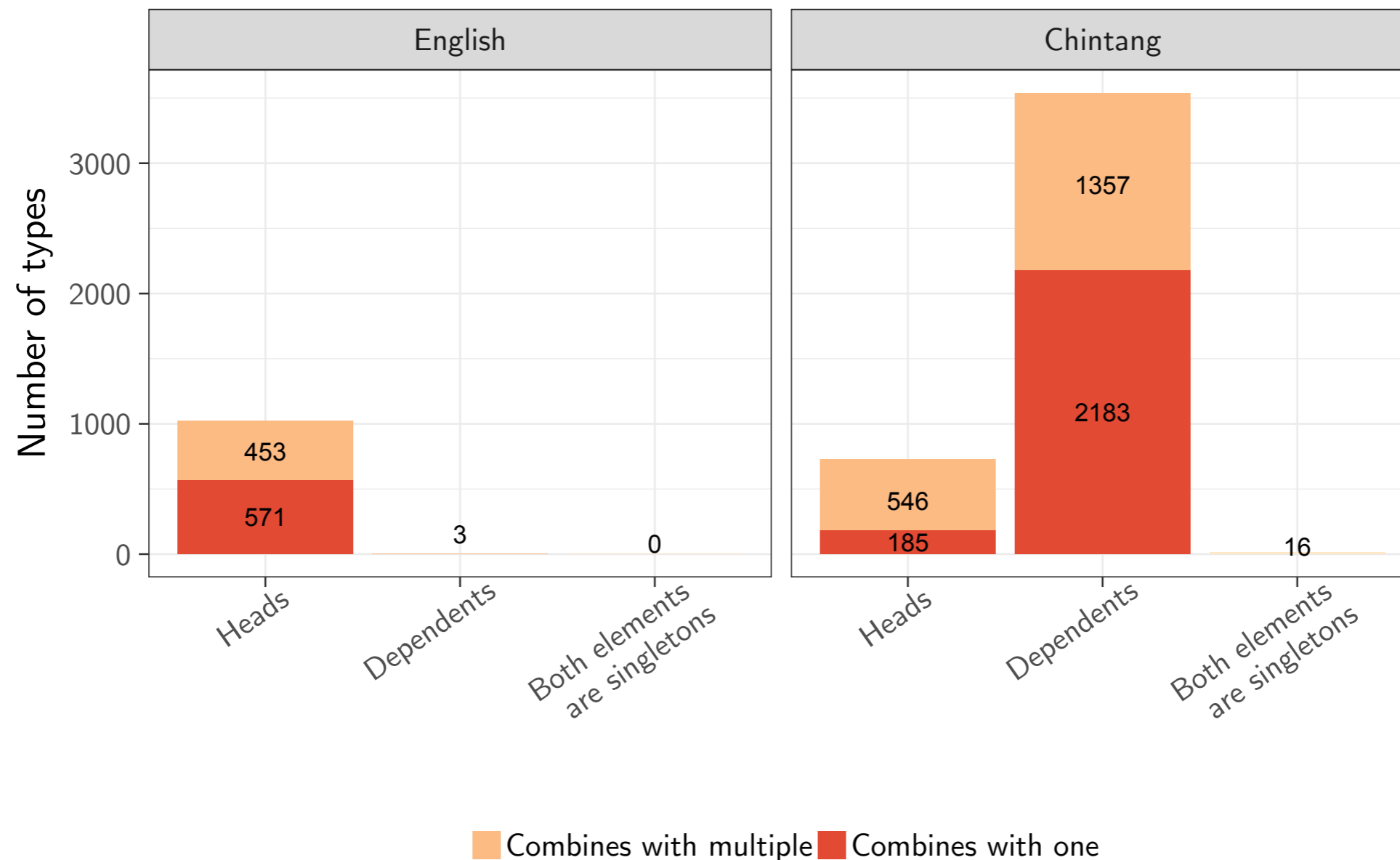
# Forms with the same stem in one recording session: Chintang (input)



# Forms with same affixes in one recording session: Chintang input

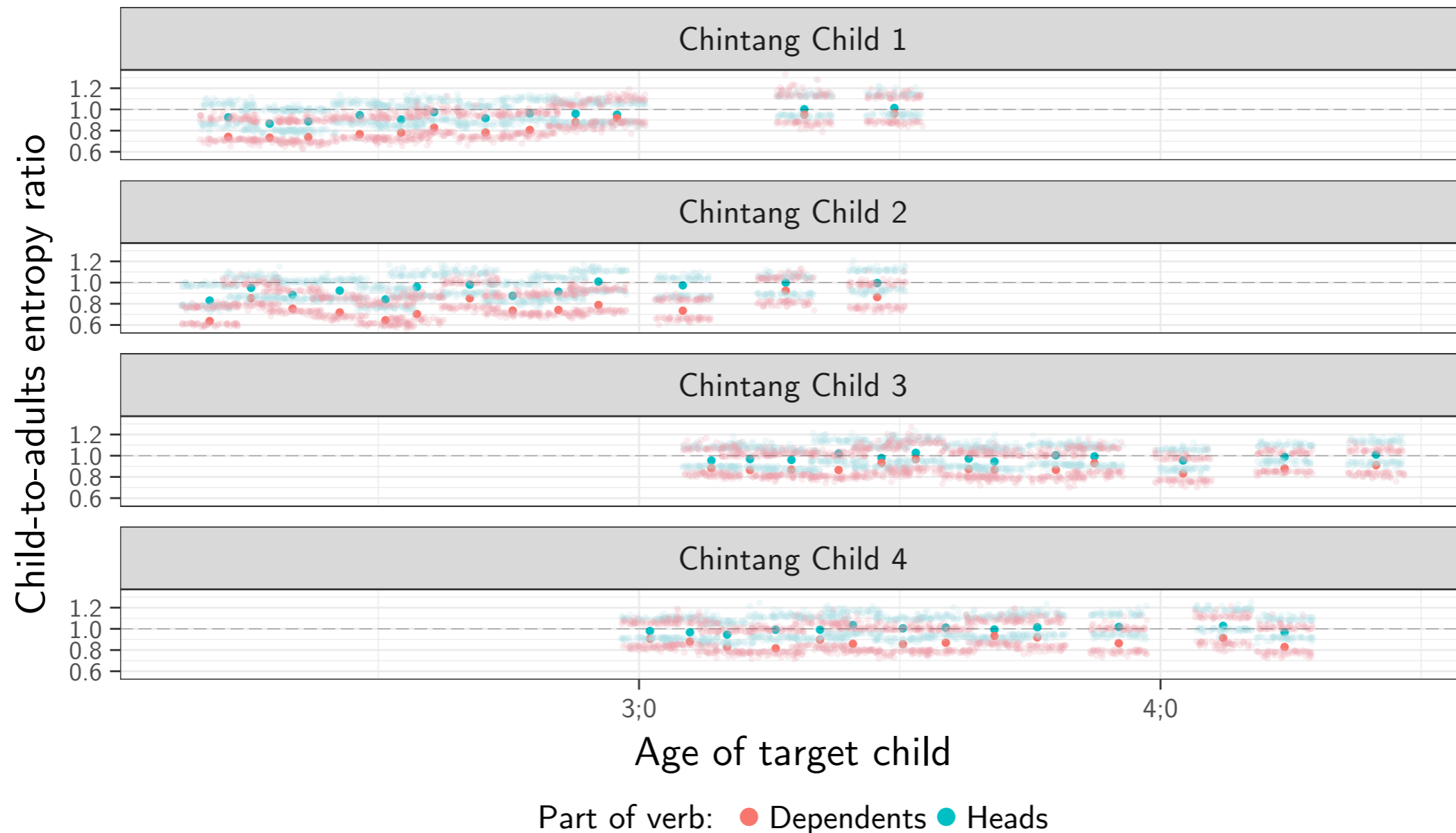


# Stem and affix combinations: Chintang (input)





# Stems vs. Affixes over time in children

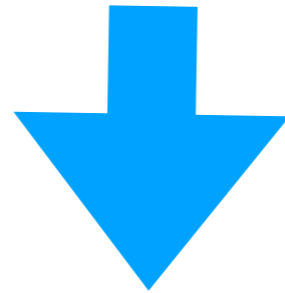


# **Conclusion**

## **Chintang verbal morphology**

**The system can be learned via  
distributional learning,  
generalisation and  
abstraction**

How do children learn  
language?



How come children can  
learn **any** language in a  
few years time?

# Challenge: diversity



# Data

# Grammars

WALS

AUTOTYP

THE WORLD ATLAS OF LANGUAGE STRUCTURES ONLINE

Home Features Chapters Languages References Authors

Browse Genealogy 100-language Sample 200-language Sample

Search a languoid by name. Matching names are formatted in bold font for languages, in italics for genera and underlined for families. This search does also take alternative names into account.

Showing 1 to 100 of 2,679 entries

Name	WALS code	ISO 639-3	Genus	Family	Macroarea	Latitude	Longitude	Countries
Aari	aar	aiw	South Omotic	Afro-Asiatic	Africa	6.00	36.58	Ethiopia
Abau	aba	aaU	Upper Sepik	Sepik	Papunesia	-4.00	141.25	Papua New Guinea
Abaza	abz	abq	Northwest Caucasian	Northwest Caucasian	Eurasia	44.00	42.00	Russia
Aberaki (Western)	abw	abe	Algonquian	Algic	North America	44.00	-72.25	United States Canada

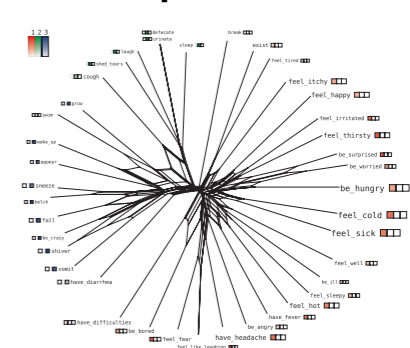
Role	Reference	Predicate class	Co-arg. role	Co-arg. ref.	TAM	Clause	Semantics	Diathesis	Polarity	Domain				
2	A <sub>tr</sub>	2	def_tr		37	99_PTCP	1	main	1	ACT	0	<any>	0	<any>
4	A <sub>str</sub>	3	def_dtr		37	99_PTCP	1	main	1	ACT	0	<any>	0	<any>
1	S	19	S <sub>success</sub>		37	99_PTCP	1	main	0	<7a>	1	ACT	0	<any>
2	A <sub>tr</sub>	404	A <sub>tr</sub> P		37	99_PTCP	1	main	0	<7a>	1	ACT	0	<any>
4	A <sub>str</sub>	832	AT G <sub>loc</sub>		37	99_PTCP	1	main	1	ACT	0	<any>	0	<any>

```
group_by(Language, ReferenceType, TAM, Polarity, Diathesis) %>%
  arrange(Role) %>%
  do({
    # split the roles by GRs
    roles <- lapply(split(.S$Role, .S$GR, drop=T), as.character)

    # convert them to alignment, sorting appropriately
    eqset <- sapply(roles, paste, collapse='=')
    eqset <- factor(eqset, levels=c('S', 'S=Atr=P', 'S=A', 'S=P', 'Atr', 'Atr=P', 'P'))
    eqset <- as.character(sort(eqset))
    alignment <- paste(eqset, collapse='*')
    data.frame(Alignment = alignment, stringsAsFactors=F)
  })
```

```
build_mat <- function(dta) {
  items <- unique(unlist(dta))
  markers <- names(dta)
  mat <- matrix(0, nrow=length(items), ncol=length(markers))
  colnames(mat) <- markers
  rownames(mat) <- items
  for(m in markers) for(i in items) if(i %in% dta[[m]]) mat[i, m] <- 1
}
# S class
rolenat5 <- build_mat(semantic.per.role['S'])
```

Language	ReferenceType	TAM	Polarity	Diathesis	Alignment
Abkhaz	#ANY#	<NA>	<NA>	ACT	S=Atr=P
Acehnese	#ANY#	<NA>	<NA>	ACT	S=Atr=P
Ainu	#ANY#	<NA>	<NA>	ACT	S=Atr=P
Alambak	#ANY#	<NA>	<NA>	ACT	S=Atr=P
Akkadian	#ANY#	<NA>	<NA>	ACT	S=Atr=P
Albanian	2plPro	<NA>	<NA>	ACT	S=Atr=P
Albanian	N-def	<NA>	<NA>	ACT	S=Atr=P
Albanian	3sgPro	<NA>	<NA>	ACT	S=Atr=P
Albanian	1sgPro	<NA>	<NA>	ACT	S=Atr=P
Albanian	2sgPro	<NA>	<NA>	ACT	S=Atr=P
Albanian	N-indef	<NA>	<NA>	ACT	S=Atr=P
Albanian	1plPro	<NA>	<NA>	ACT	S=Atr=P
Albanian	3plPro	<NA>	<NA>	ACT	S=Atr=P





# Data: ACQDIV bank



Cluster 1

Cluster 2

Cluster 3

Cluster 4

Cluster 5

Turkish  
Japanese

Indonesian  
Yucatec

Inuktitut  
Chintang

Sesotho  
Russian

Dene  
Cree

**New Corpora:**

Qaqet  
Nungon  
Pijanjatjara  
Ku Waru  
Tuatschin  
English  
Shipibo  
etc.

# ACQDIV core team



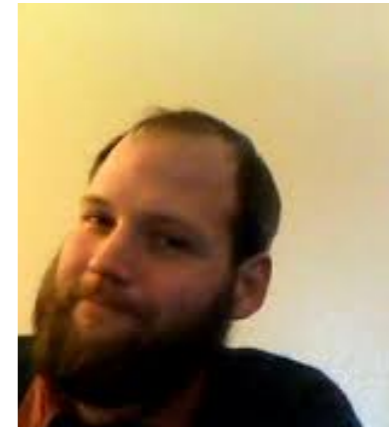
**Steven Moran**



**Damian Blasi**



**Dagmar Jung**



**Nick Lester**



**Géraldine Walther**



**Jekaterina Mazara**



**Guanghao You**



**Claudia Cathomas**



**André Müller**



**Anna Jansco**



**Andreas Gerster**



**Alexandra Bosshard**



**Ruben Moegel**



**Katherina Habel**



**Areezoo Zandy**









ELAN - CLLDCh3R02S04.eaf

File Edit Annotation Tier Type Search View Options Window Help

Grid Text Subtitles Lexicon Audio Recognizer Video Recognizer Metadata Controls

tx@JIT

Nr	Annotation	Begin Time	End Time	Duration
112	kristina	00:41:14.010	00:41:14.800	00:00:00.790
113	e duijana coptana	00:41:15.275	00:41:17.580	00:00:02.305
114	kanchi to	00:41:18.824	00:41:19.800	00:00:00.976
115	kanchi	00:41:20.446	00:41:21.380	00:00:00.934
116	to hanako hanako mama	00:41:22.109	00:41:24.410	00:00:02.301
117	arj	00:41:25.670	00:41:26.240	00:00:00.570
118	bago salo kha	00:41:27.540	00:41:28.720	00:00:01.180
119	topattigo	00:41:29.905	00:41:30.580	00:00:00.675
120	ena bago salo kha	00:41:30.660	00:41:31.870	00:00:01.210
121	hale cekta	00:41:32.888	00:41:33.860	00:00:00.972

00:41:15.380 Selection: 00:41:27.540 - 00:41:28.720 1180

Selection Mode Loop Mode

los1@LDCh3 [0]

ref@JIT [202] CLLDCh3R02S04. 1022 CLLDCh3R02S0 CLLDCh3R02S CLLDCh3R02S04. 1026 CLLDCh

tx@JIT [202] e duijana coptana kanchi to kanchi to hanako hanako mama arj

gw@JIT [247] e duijana copta na kanchi to kanchi to hanako hanako mama arj

mph@JIT [701] e dui -jan copt -a na kanchi to kanchi to han -ko han -ko mama arj

mgl@JIT [701] EXCLA two -CL see -IMP PTCL younge DEM.U youngest fem.si DEM.UP 2s -GE 2s -GE mother's. PTCL

lg@JIT [701] C N -N C -C C/N C/N C C/N C C -C C -C N C

id@JIT [99]

ps@JIT [2]

eng@JIT [202] He! Look at the two people. There, Kanchi. Kanchi. There, your uncle, your uncle. Right?

nep@JIT [200] ए! दुई जनालाई हेर न। उ कान्छी। कान्छी। उहाँ तिमरो तिमरो मामा मामा। नि ?

cxt@JIT [1]

comment@JIT [14]

dt@JIT [202] 10/Mar/2010 10/Mar/2010 10/Mar/2010 10/Mar/2010 10/Mar/2010

point@JIT [22] uncodable

object pointe [22] picture on the wall

ref in utt@JIT [22] multiple possible referents

word in utt@ [22] to (DEM.UP), hanako mama (2s-GEN mot,

type of sente [21] declarative sentence

intention@JI [21] statement

connected p [4]



# ACQDIV bank

language	corpus	session_id	utterance_id	word_id	morpheme_id	word	pos_word_stem	morpheme	pos	gloss	morpheme_language	speaker_label	age_in_days	gender	role	macrorole	languages_spoken	wa
Chintang	Chintang	1	1	1	1	habiniŋ	V	hap	V	cry	Chintang	GKR	14600	Female	None	Adult	nep bap ctn	NA
Chintang	Chintang	1	1	1	2	habiniŋ	V	i	sfx	1/2pS/P	Chintang	GKR	14600	Female	None	Adult	nep bap ctn	NA
Chintang	Chintang	1	1	1	3	habiniŋ	V	niriŋ	sfx	NEG	Chintang	GKR	14600	Female	None	Adult	nep bap ctn	NA
Cree	Cree	478	393608	987673	1600151	awân	PRODEM	NA	PRODEM	who	Cree	ADU	12842	Female	Speaker	Adult	crl	NA
Indonesian	Indonesian	503	410983	1023914	1616384	da	stem	da	stem	bye	Indonesian	MOT	NA	Female	Mother	Adult	xmm ind	NA
Indonesian	Indonesian	503	410983	1023916	1616386	???	???	???	???	???	Indonesian	MOT	NA	Female	Mother	Adult	xmm ind	NA
Indonesian	Indonesian	503	410983	1023917	1616387	Tante	stem	Tante	stem	aunt	Indonesian	MOT	NA	Female	Mother	Adult	xmm ind	NA
Inuktitut	Inuktitut	1500	1326526	3516259	4381281	Naa	PTCL	no	PTCL	no	Inuktitut	ALI	916	Male	Target_Child	Target_Child	ike	NA
Japanese	Japanese_Miyata	1577	1373193	3589311	4419047	issho	N	issho	N	together	Japanese	AMO	NA	Female	Speaker	Adult	und	NA
Japanese	Japanese_Miyata	1577	1373196	3589314	4419048	pikapika	IDEOPH	pikapika	IDEOPH	???	Japanese	AMO	NA	Female	Speaker	Adult	und	NA
Japanese	Japanese_Miyata	1577	1373198	3589317	4419050	ne	PTCL	ne	PTCL	TAG	Japanese	AMO	NA	Female	Speaker	Adult	und	NA
Russian	Russian	1939	1810004	4783322	5470915	oj	INTJ	oj	INTJ	INTERJ	Russian	ALJ	625	Unspecified	Target_Child	Target_Child	NA	NA
Russian	Russian	1939	1810005	4783323	5470916	???	???	???	???	???	Russian	ALJ	625	Unspecified	Target_Child	Target_Child	NA	NA
Russian	Russian	1939	1810006	4783324	5470917	Alja	N	alja	N	M:SG:NOM:AN	Russian	LEN	NA	Female	Aunt	Adult	NA	NA
Russian	Russian	1939	1810006	4783325	5470918	ne	PTCL	ne	PTCL	PCL	Russian	LEN	NA	Female	Aunt	Adult	NA	NA
Sesotho	Sesotho	2389	2637993	6816355	7494533	ere	V	er	V	say	Sesotho	MHL	NA	Female	Mother	Adult	sme	NA
Sesotho	Sesotho	2389	2637993	6816355	7494534	ere	V	e	sfx	m^i	Sesotho	MHL	NA	Female	Mother	Adult	sme	NA
Sesotho	Sesotho	2389	2637993	6816356	7494535	mphe	V	m	px	om1s	Sesotho	MHL	NA	Female	Mother	Adult	sme	NA
Sesotho	Sesotho	2389	2637993	6816356	7494536	mphe	V	ph	V	give	Sesotho	MHL	NA	Female	Mother	Adult	sme	NA
Turkish	Turkish	2667	2858439	NA	7824079	NA	NA	değil	PTCL	???	Turkish	NA	NA	NA	NA	NA	NA	NA
Turkish	Turkish	2667	2858439	NA	7824080	NA	NA	o	PRODEM	???	Turkish	NA	NA	NA	NA	NA	NA	NA
Turkish	Turkish	2667	2858440	7504733	7824081	kamera	N	kamera	N	???	Turkish	MOT	NA	Female	Mother	Adult	xmm ind	NA
Yucatec	Yucatec	2891	3108071	8185947	8123413	Lorena-e'	N	Lorena	N	???	Yucatec	MAR	21486	Male	Speaker	Adult	jav ind	NA
Yucatec	Yucatec	2891	3108071	8185947	8123414	Lorena-e'	N	e'	sfx	TOP	Yucatec	MAR	21486	Male	Speaker	Adult	jav ind	NA
Yucatec	Yucatec	2891	3108071	8185948	8123415	táan	???	táan	???	???	Yucatec	MAR	21486	Male	Speaker	Adult	jav ind	NA
Yucatec	Yucatec	2891	3108071	8185949	8123416	u	???	u	???	???	Yucatec	MAR	21486	Male	Speaker	Adult	jav ind	NA



**Steven Moran**



**Nick Lester**

# Study 2: Variation sets

# Hypothesis

**There are patterns in the input that make learning and generalisation possible.**

# Learning in interaction





## Two hypothetical language learning situations.

<b>Alien message #1</b>	<b>Alien message #2</b>
(1a) kedmalburafuloropesai	(1b) <u>kedmalburafuloropesai</u>
(2a) gianaber	(2b) <u>rafuloro</u>
(3a) manadukbiunel	(3b) <u>manaloropesai</u>
(4a) kiciorudanamjeisulcaz	(4b) <u>kedmalbumanalogoropesai</u>



# Variation sets in the input

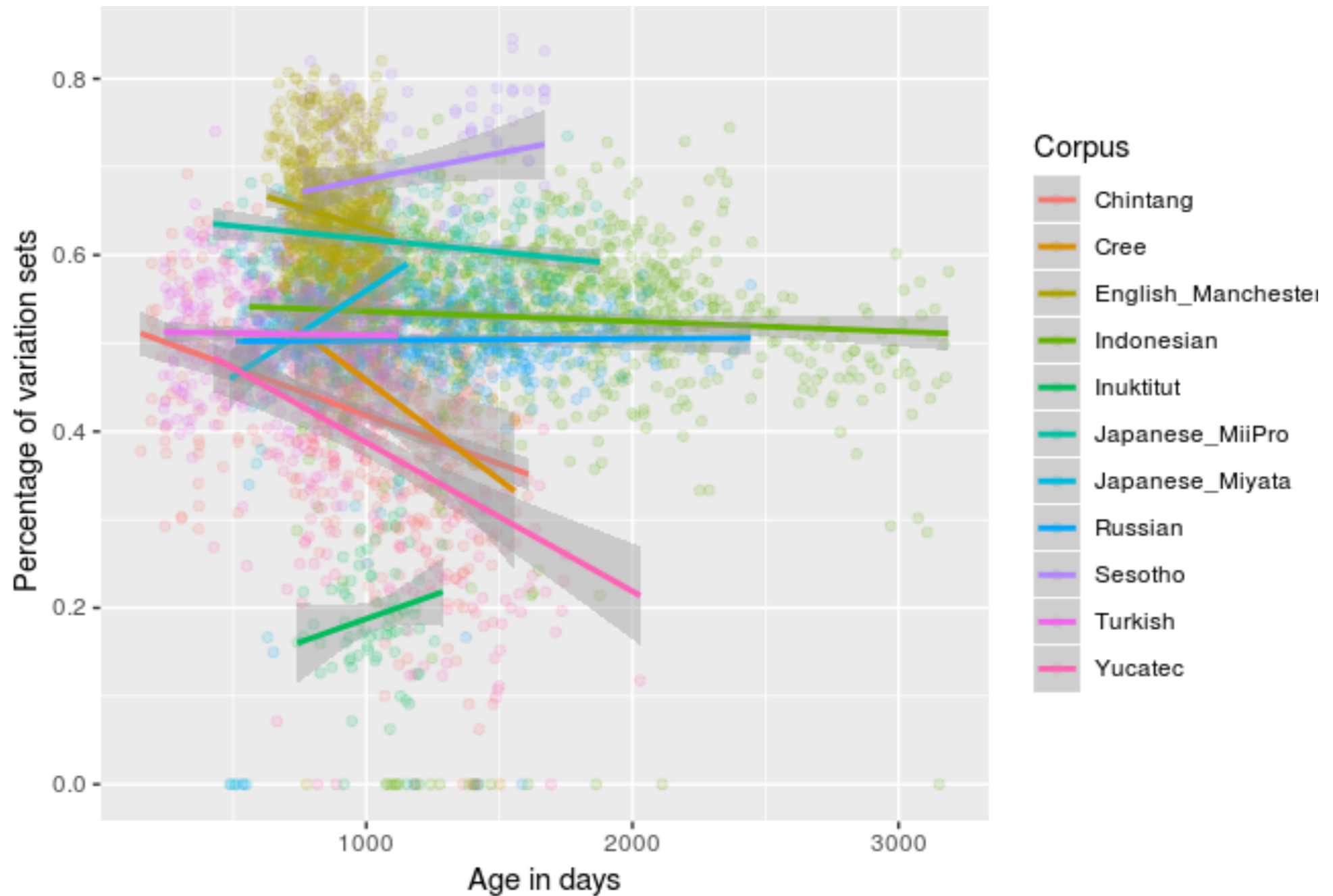
- \*MOT: und jetzt rein in den schuh.
- \*MOT: und die schuhe?
- \*MOT: weg .
- \*MOT: so zumachen.
- \*MOT: eine schleife.
- \*MOT: oh 'n grossen schuh.
- \*MOT: bum.
- \*MOT: warte.
- \*MOT: du kriegst gleich dein'n schuh wieder.
- \*CHI: Mama.
- \*MOT: nein nich?
- \*MOT: wieder zu?
- \*CHI: zu.
- \*MOT: zu?
- \*CHI: &ja schuh &em.
- \*MOT: schuh?
- \*MOT: ein schuh?
- \*MOT: die puppe is zu klein für den grossen schuh?

# Variation sets in the input

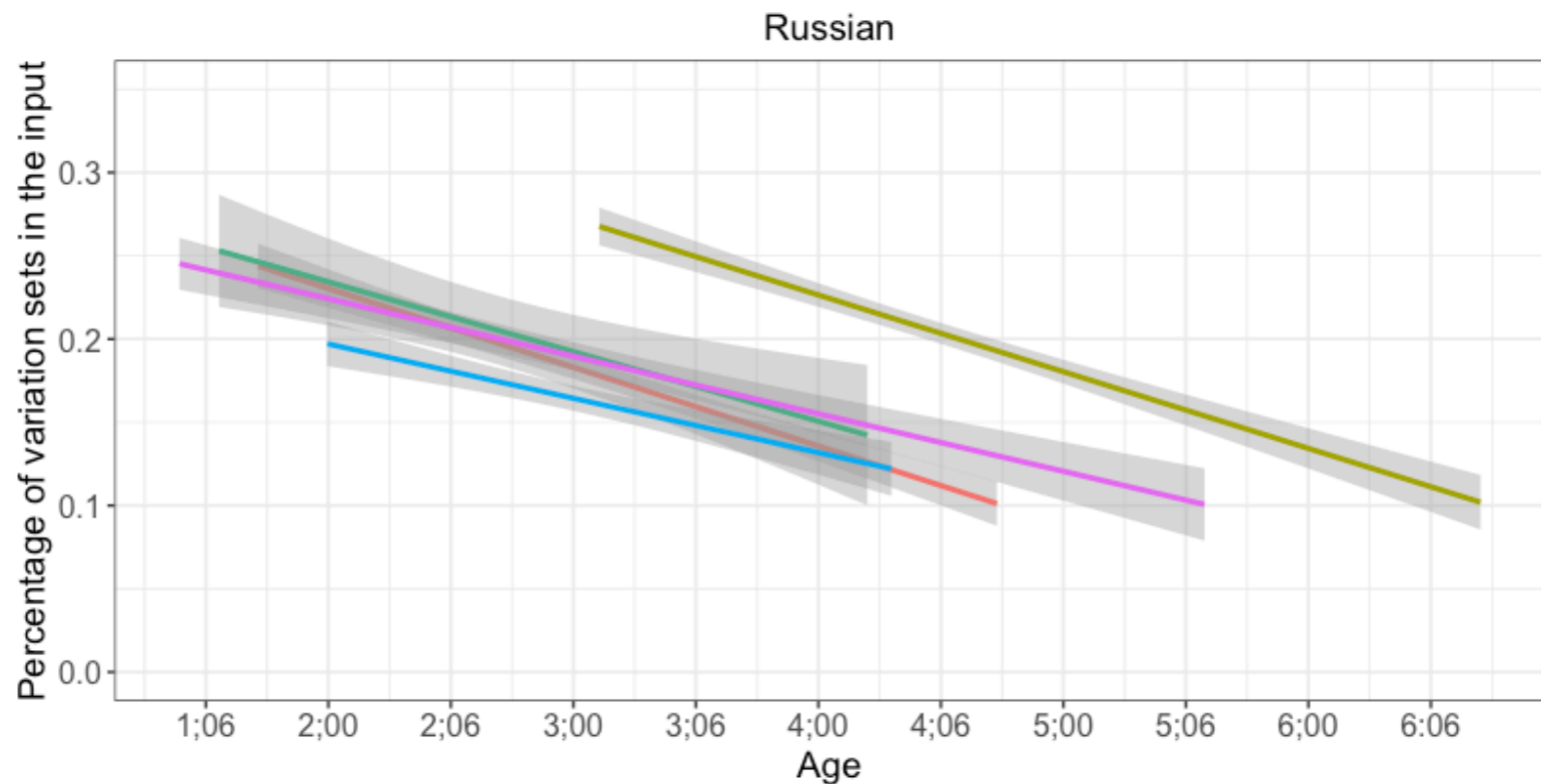
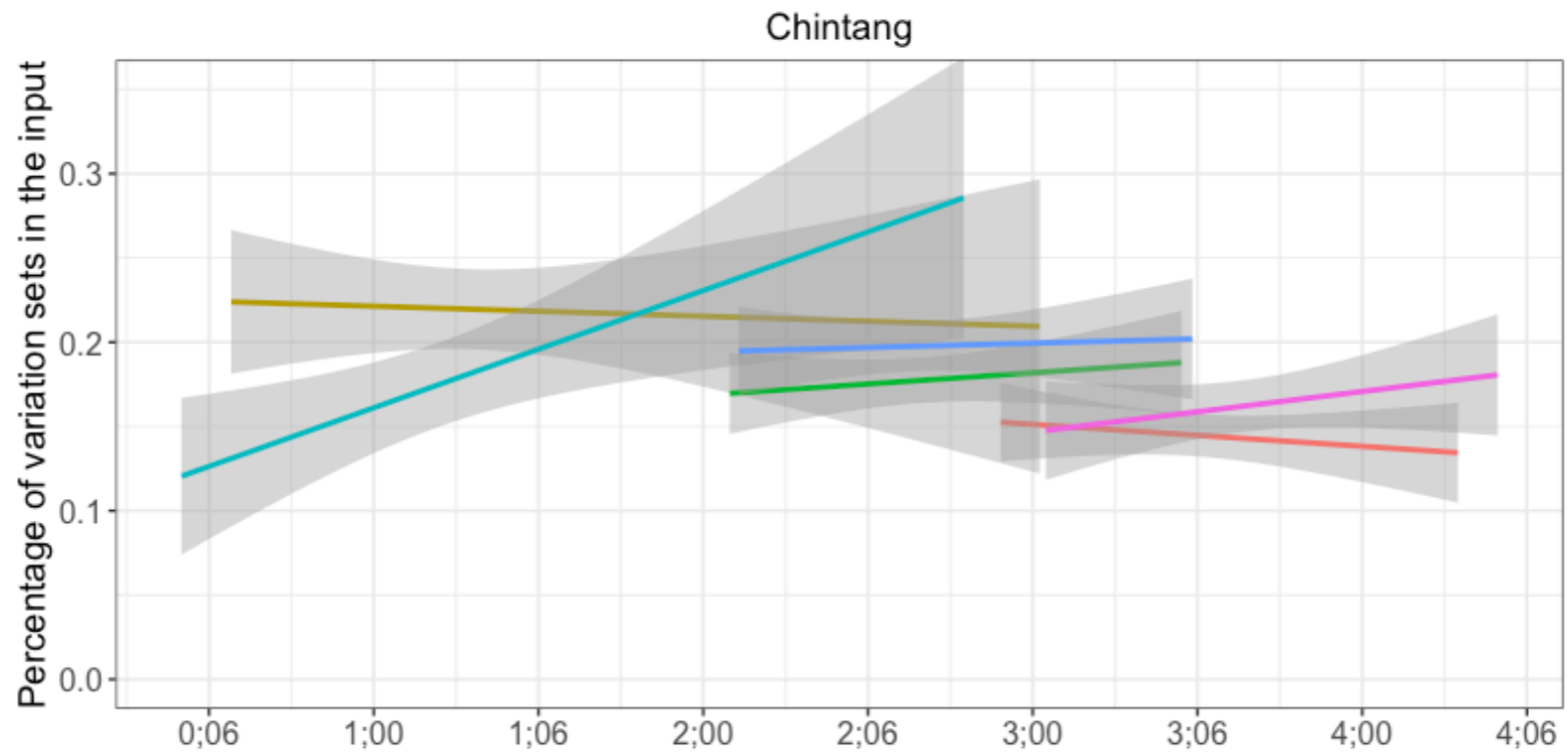
- Repetitive units in interactional units
- 17% - 30% of verbs and nouns in variation sets in child-directed speech

***Are variation sets* in input  
universally reliable patterns and  
how do they develop as a  
function of the age of the child?**

# Variation sets



# Variation sets



# Conclusion variation sets

Found in all of our maximally diverse languages

Language specific patterns in the development over time, not necessarily decreasing





**Steven Moran**

# Study 3: Frames to detect nouns and verbs

# Frames

I  
I  
I  
I  
I  
I  
I

like  
hate  
hear  
love  
and  
not

you  
you  
you  
you  
you  
you  
you

# Frames

## Frame you\_\_it

put (28), want (15), do (10), see (7), take (6), turn (5), taking (5), said (5), sure (4), lost (4), like (4), leave (4), got (4), find (4), throw (3), threw (3), think (3), sing (3), reach (3), picked (3), get (3), dropped (3), seen (2), lose (2), know (2), knocked (2), hold (2), help (2), had (2), gave (2), found (2), fit (2), enjoy (2), eat (2), chose (2), catch (2), with (1), wind (1), wear (1), use (1), took (1), told (1), throwing (1), stick (1), share (1), sang (1), roll (1), ride (1), recognize (1), reading (1), ran (1), pulled (1), pull (1), press (1), pouring (1), pick (1), on (1), need (1), move (1), manage (1), make (1), load (1), liked (1), lift (1), licking (1), let (1), left (1), hit (1), hear (1), give (1), flapped (1), fix (1), finished (1), drop (1), driving (1), done (1), did (1), cut (1), crashed (1), change (1), calling (1), bring (1), break (1), because (1), banged (1)

### Frame *a-* ‘2S/A’ \_\_\_ *-o* ‘3P’

- a. *theke a-khind-o-ko?*  
why 2S/A-pull-3P-IND.NPST  
‘Why do you pull it?’
- b. *hūi kābili a-nept-o*  
DEM third.daughter 2S/A-step.on-3P  
‘Kaīli, you might step on it!’
- c. *a-bid-o-ko?*  
2S/A-watch-3P-IND.NPST  
‘Can you look after her?’

### Frame *mai-* ‘NEG’ \_\_\_ *-th* ‘NEG’

- a. *la mo mai-soṅ-th-a*  
INTERJ DEM.DOWN NEG-move-NEG-IMP  
‘Don’t move down there!’
- b. *e mai-khaṅ-th-o-s-e ni*  
OK NEG-see-NEG-3P-PRF-IND.PST EMPH  
‘Yeah, s/he didn’t see it.’
- c. *mo sotaṅma-ce phidaṅ u-mai-let-th-a-ṅs-a-kha*  
DEM.DOWN S.-ns ginger 3nsS/A-NEG-plant-NEG-PST-PRF-PST-NMLZ  
‘The Sotangma people have not yet planted ginger down in the field.’

**Are *frames* in the input universally reliable patterns that could help in categorizing parts of speech?**

# Frames in ACQDIV bank

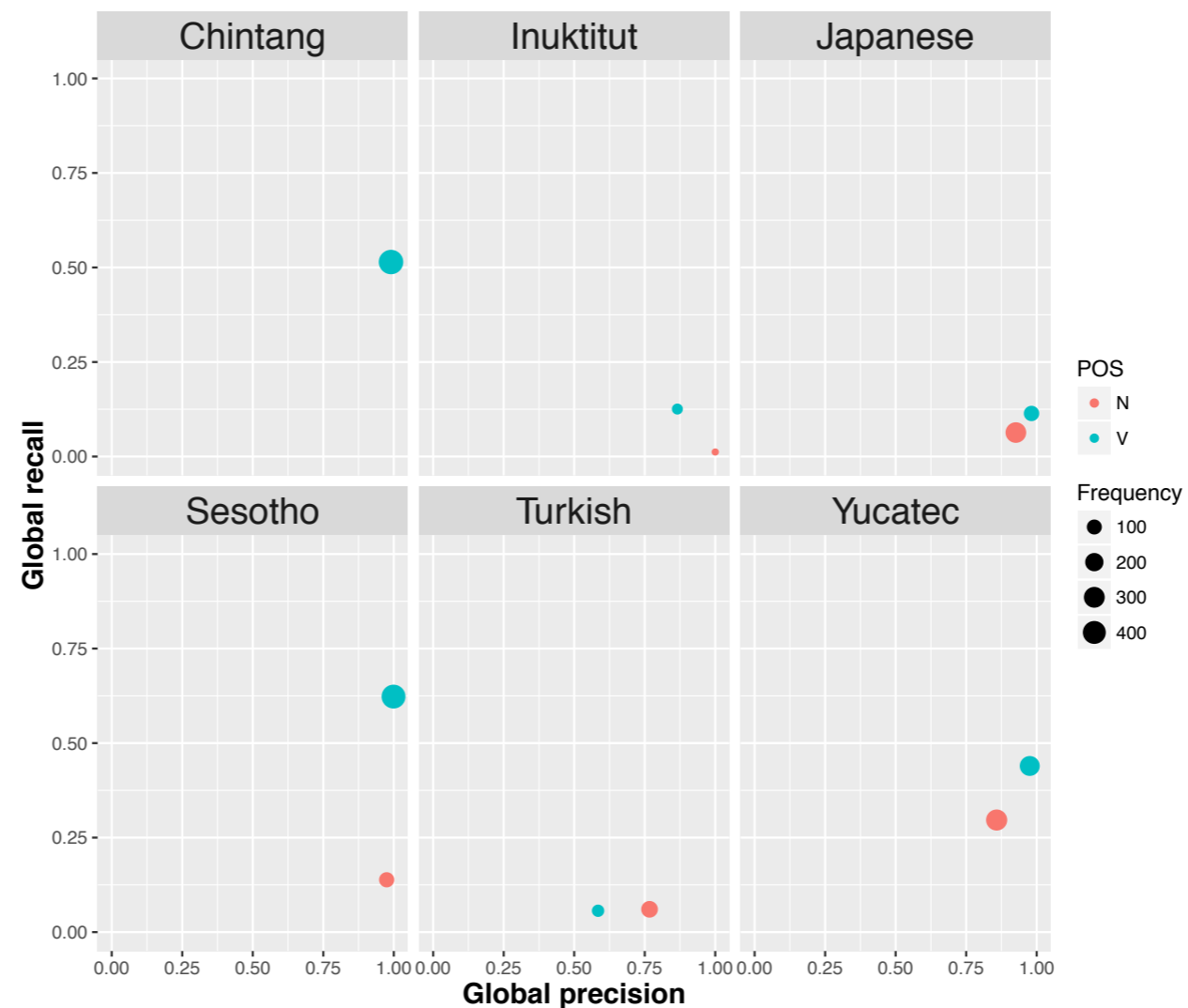
Word frames: no

	Accuracy	SD	Completeness	SD	Frames	Min	Max
Chintang	0.57	0.24	0.04	0.02	33	90	2720
Inuktitut	0.98	0.11	0.03	0.01	37	2	3
Japanese	0.82	0.21	0.02	0.02	97	67	915
Russian	0.44	0.22	0.04	0.03	48	234	1485
Sesotho	0.83	0.23	0.01	0.01	107	8	163
Turkish	0.62	0.20	0.08	0.08	15	34	318
Yucatec	0.78	0.28	0.01	0.01	133	3	41

Morpheme frames: yes

	Accuracy	SD	Completeness	SD	Frames	Min	Max
Chintang	0.95	0.09	0.08	0.07	60	517	7940
Inuktitut	0.93	0.16	0.02	0.01	100	5	43
Japanese	0.98	0.04	0.02	0.03	187	83	1943
Sesotho	0.97	0.12	0.04	0.04	88	66	1358
Turkish	0.88	0.17	0.01	0.01	835	21	1000
Yucatec	0.90	0.18	0.01	0.02	153	20	584

# Categorization of nouns and verbs via frames





# **Conclusion frames:**

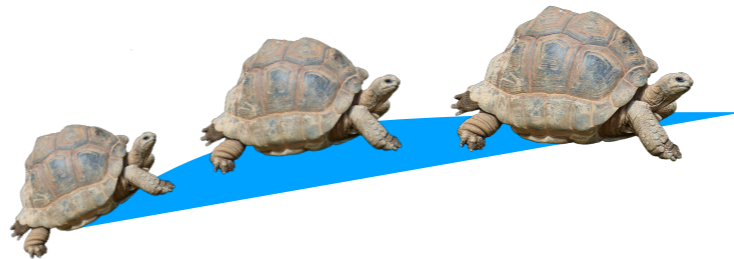
**Nouns and verbs can be  
categorized via repetitive  
patterns in their morphological  
structure**



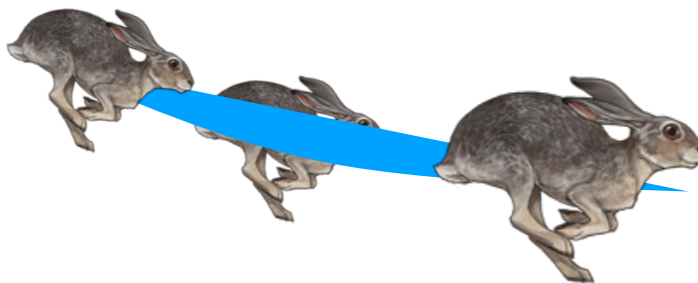
**Nick Lester**

# Study 4: Temporal cues to detect nouns and verbs

# Temporal cues before nouns and verbs

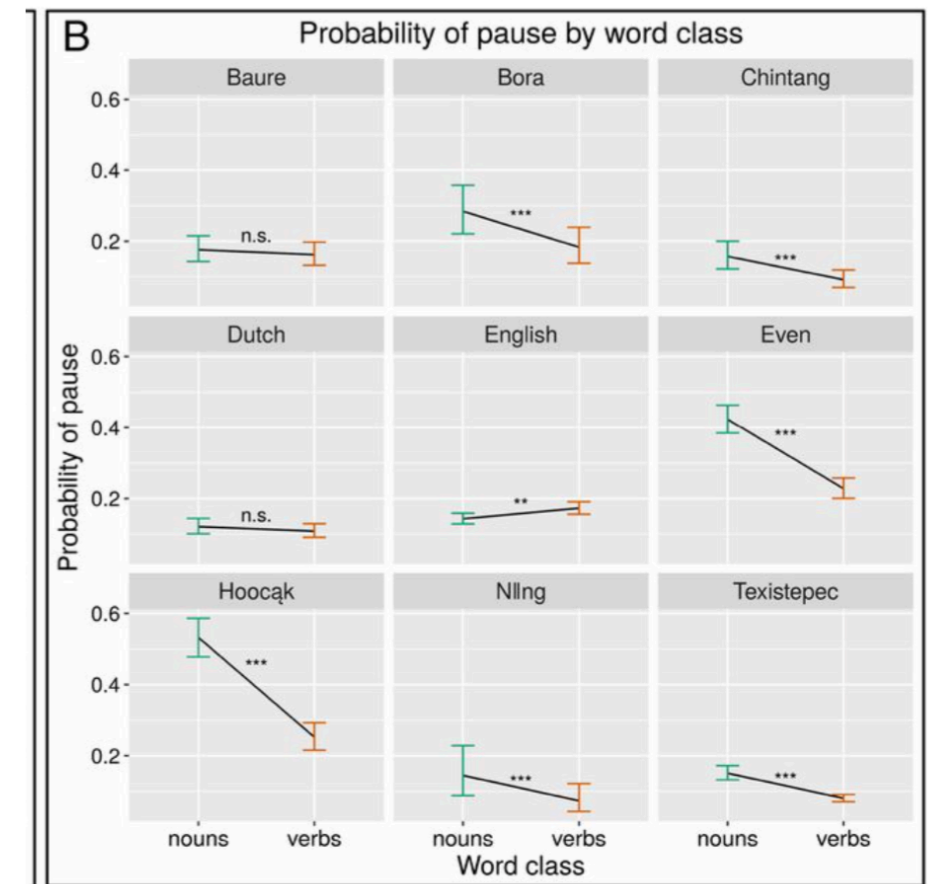
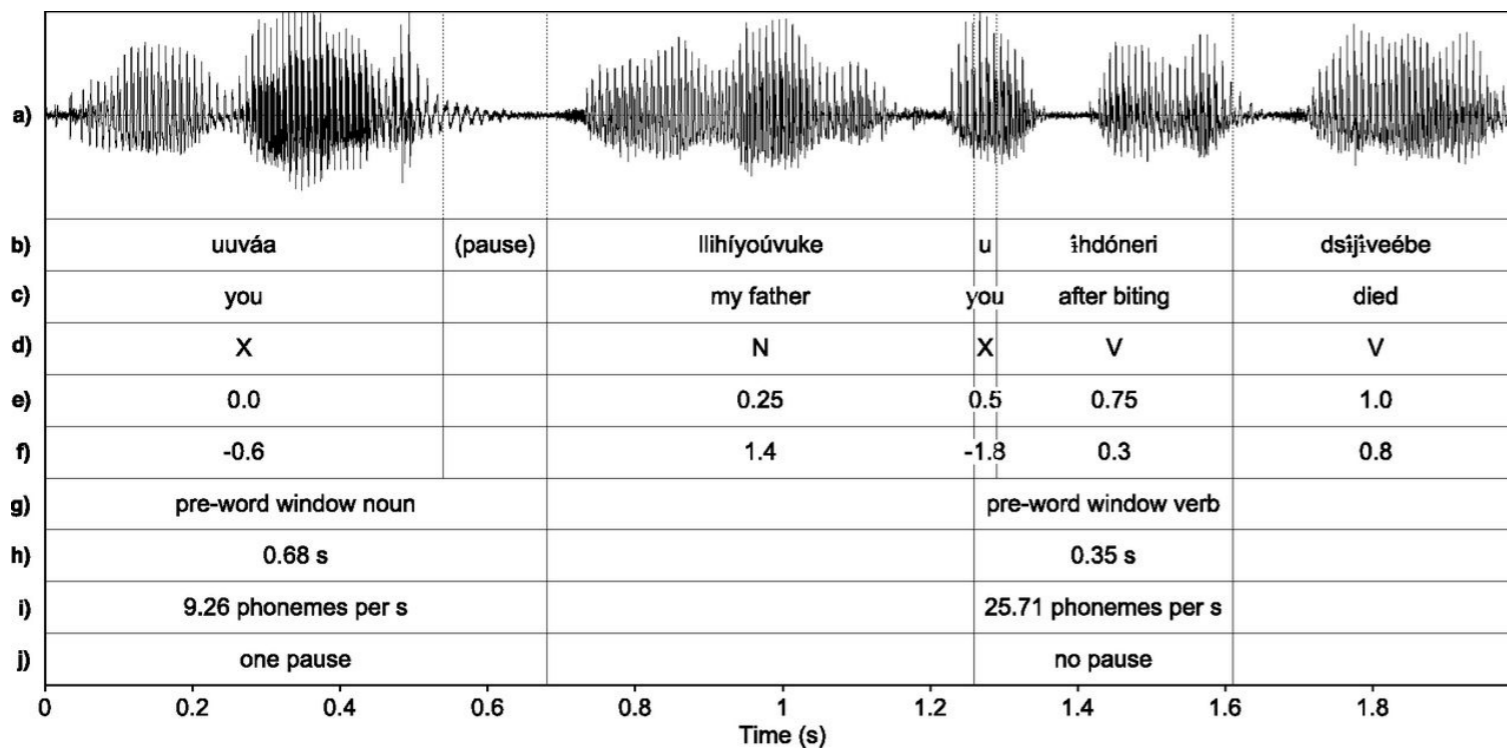


**nouns**

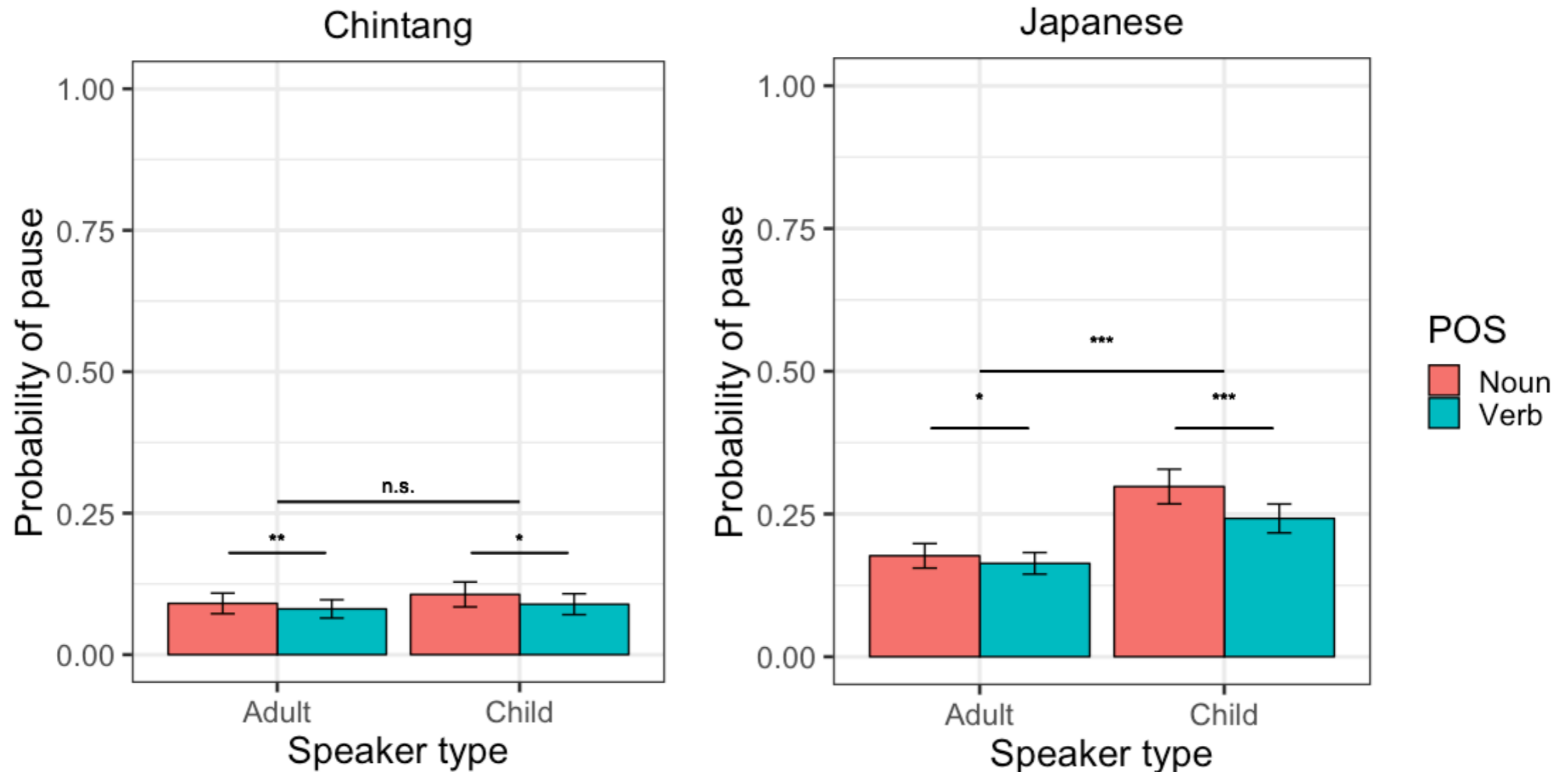


**verbs**

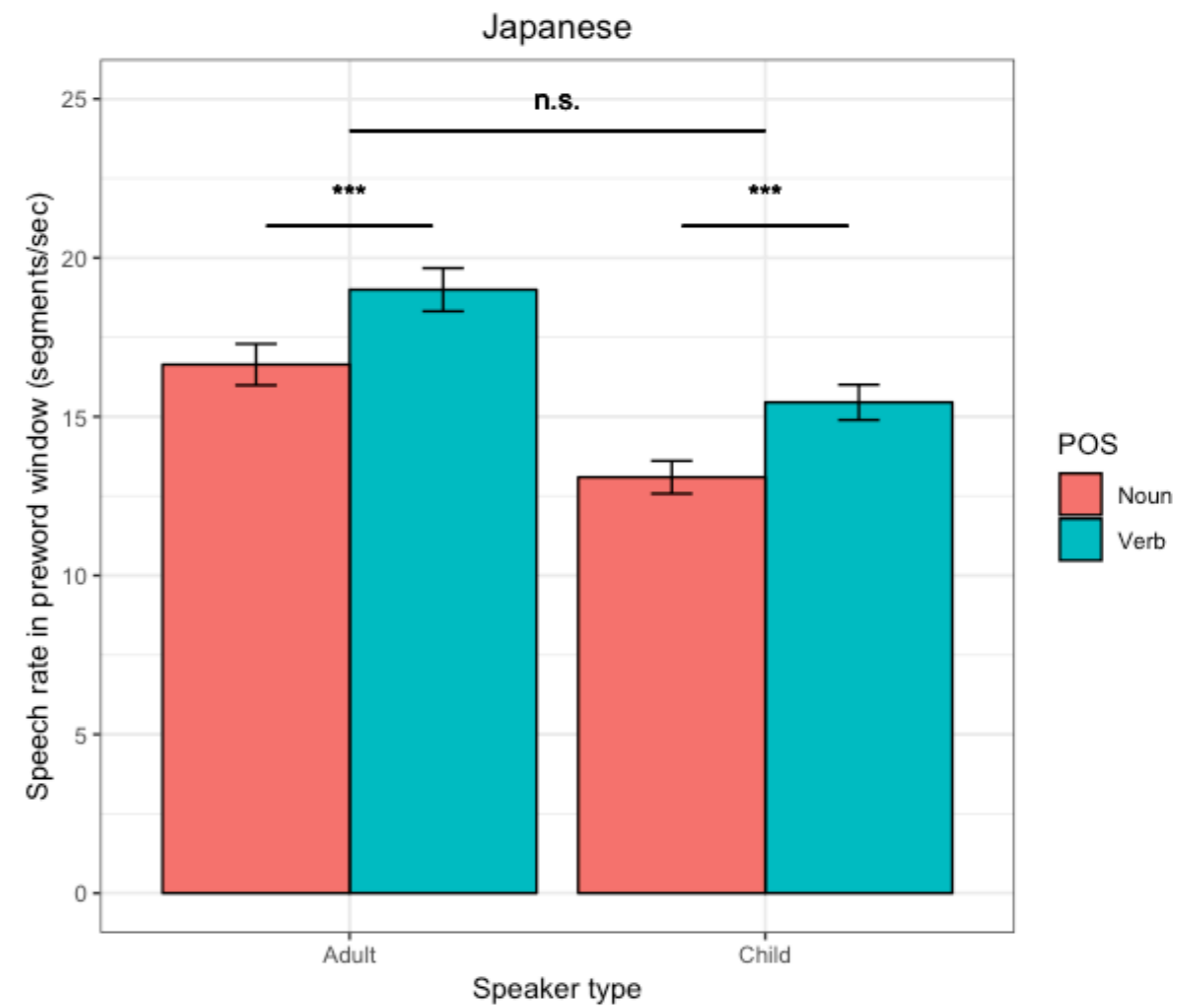
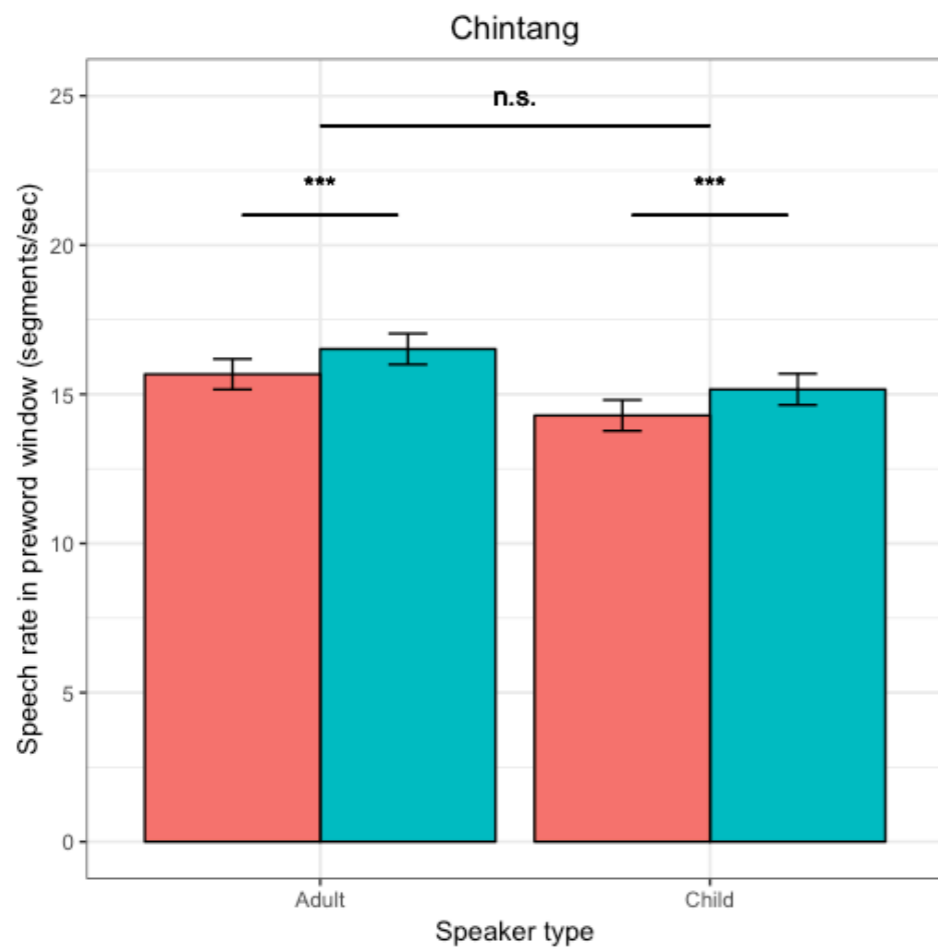
# Temporal cues before nouns and verbs: adults



# Temporal cues for nouns and verbs: pauses in child-surrounding speech

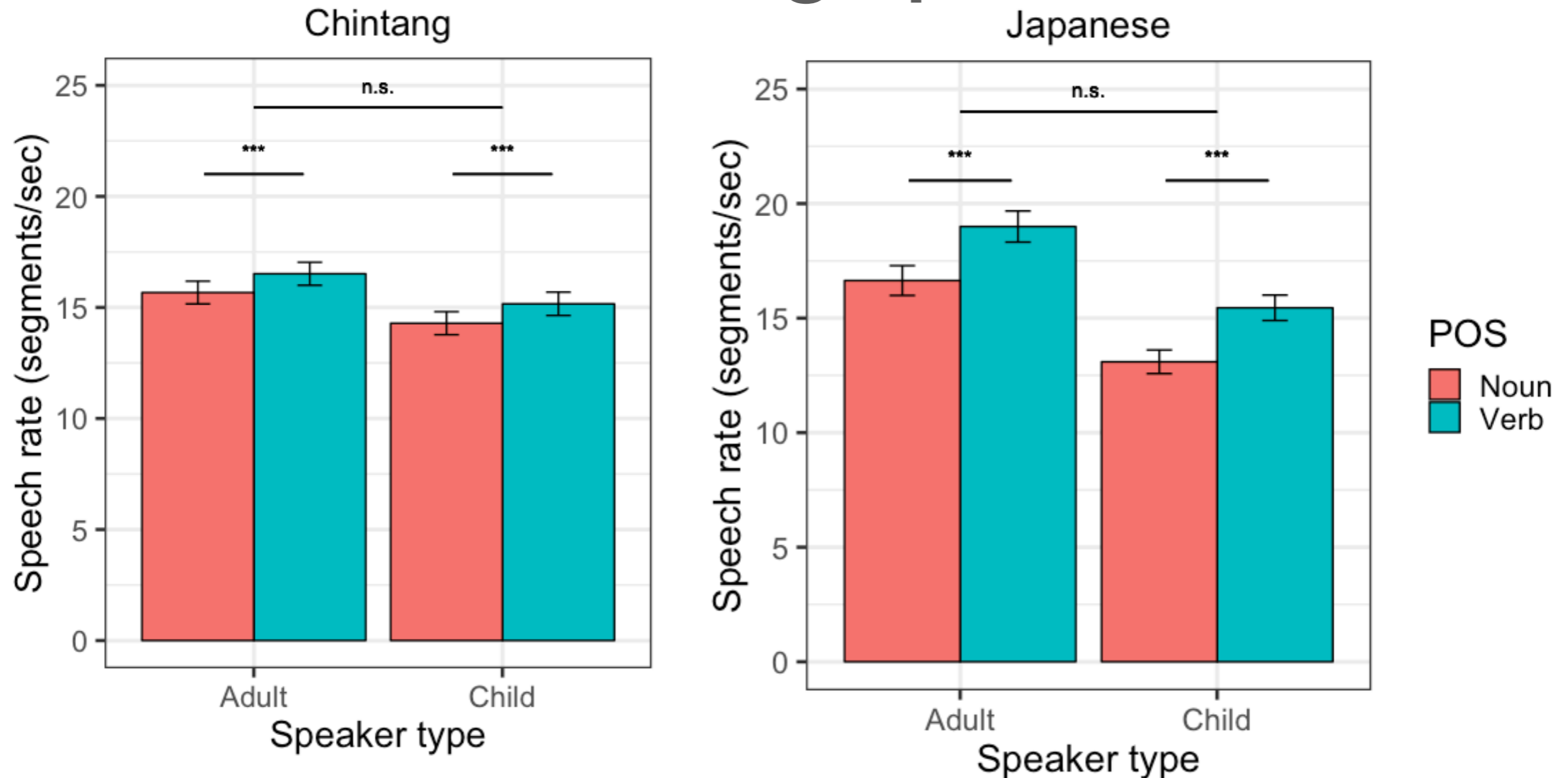


# Sprechgeschwindigkeit vor Nomen oder Verben





# Temporal cues for nouns and verbs: speech rate in child-surrounding speech



Lester, Bickel & Stoll in prep.

# **Conclusion**

## **temporal cues:**

**Both pauses before nouns and verbs and the temporal duration of these units can serve as cues for syntactic categorization**

# To conclude:

- The input is not chaotic and there are quite a number of patterns learners can rely on
- There are striking similarities in the information distribution in child-directed speech in maximally diverse languages
- Some cross-linguistic differences in the patterns
- We are only at the beginning at understanding the computational power underlying language development
- Children can apply distributional learning and make use of statistical patterns provided by the input

**Thank you very much  
for your interest!**