



The effect of topography on language and cognition in Isthmus Zapotec

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Geographic Grounding: Place, direction and landscape in the grammars of the v

Copenhagen, Denmark

Synopsis

2

Introduction

MesoSpace

Topography in MesoSpace

Space and topography in Diidxa za

Cultural mediation: ethnophysiology

Conclusions

Introduction

3

- ▶ does topography influence language & cognition
 - ▶ test case: spatial frames of reference
 - ▶ previous qualitative work: Wassman & Dasen 1999; Polian & Bohnemeyer 2011; Palmer 2015

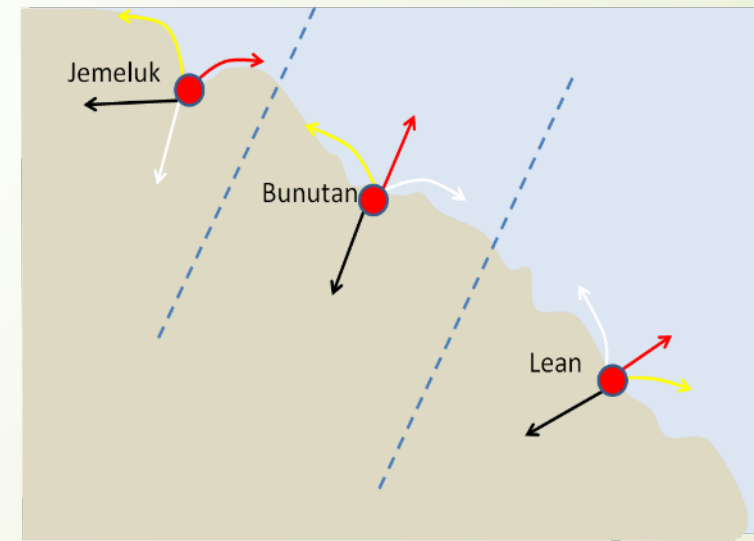
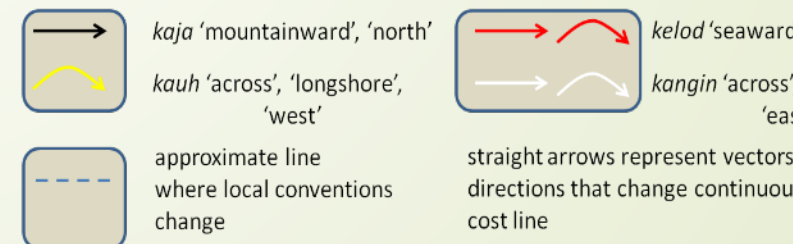


Figure 1. Three neighboring villages on the North-East peninsula of Bali using the same set of geocentric terms each based on different local conventions (Bohnemeyer et al. ms, based on a detail from Wassman & Dasen 1998: 698)



- ▶ does topography influence cognition? (cont.)
 - ▶ preliminary quantitative evidence: MesoSpace
 - ▶ (Bohnemeyer et al 2014, 2015, ms.)
 - ▶ second part of the talk: the role of culture
 - ▶ inter-community variation in the Isthmus of Tehuantepec

Synopsis

5

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- ▶ Cultural mediation: ethnophysiology
- ▶ Conclusions

MesoSpace

6

➤ NSF award #BCS-0723694
“Spatial language and cognition in Mesoamerica”

➤ 15 field workers

➤ 13 MA languages

➤ Mayan

- Chol (J.-J. Vázquez)
- Q’ anjob’ al (E. Mateo Toledo)
- Tzeltal (G. Polian)
- Yucatec (J. Bohnemeyer)

➤ Mixe-Zoquean

- Ayutla Mixe (R. Romero Méndez)
- Soteapanec (S. Gutierrez Morales)
- Tecpatán Zoque (R. Zavala Maldonado)

➤ Oto-Manguean

- Otomí (E. Palancar; N. H. Green; S. Hernández-Gómez)
- Juchitán Zapotec (G. Pérez Báez)



➤ Tarascan

➤ Purepecha (A. Capistrán)

➤ Totonacan

➤ Huehuetla Tepehua (S. Smythe Kung)

➤ Uto-Aztecan

➤ Cora (V. Vázquez)

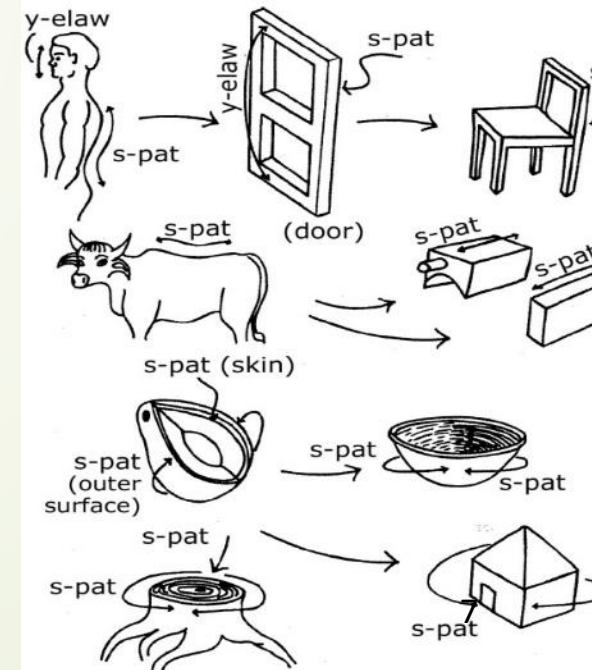
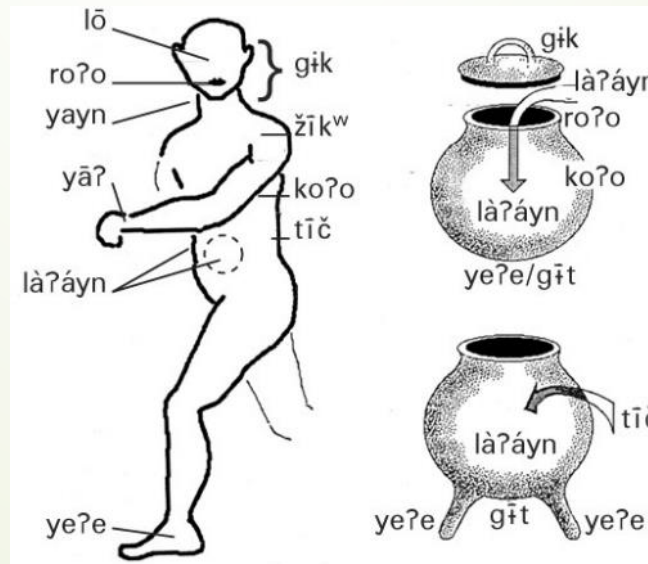
➤ Pajapan Nawat (V. Perc...

7

- 3 non-MA “controls”
 - Seri (C. O’ Meara)
 - Mayangna (E. Benedicto, A. Eggleston in collaboration with the Mayangna Yulbarangyang Balna)
 - Mexican Spanish (R. Romero Méndez)
- 2 (interrelated) domains
 - **frames of reference** and **meronyms** (labels for entity parts)



Figure 4. Meronyms in Ayoquesco Zapotec (left) and Tenejapa Tzeltal (adapted from MacLaury 1989 and Levinson 1994)



➤ MesoSpace Ib

8

➤ *Spatial language and cognition beyond Mesoamerica*

- NSF award #BCS 1053123, 2011 – 2016

➤ new languages

- Jahai (Mon-Khmer; N. Burenhult)
- Japanese (isolate; J. Olstad)
- Mandarin (Sino-Tibetan; H. Hsiao)
- Taiwanese (Sino-Tibetan; H. Hsiao)
- Vietnamese (Mon-Khmer; J. Lovegren)
- Wan (Mande; T. Nikitina)
- Yurakaré (isolate, Bolivia; R. van Gijn and V. Hirtzel)

➤ continuing languages

- additional data collected from speakers of
 - Isthmus Zapotec; Tzeltal, and Yucatec Maya; Mayangna and Spanish

➤ objectives

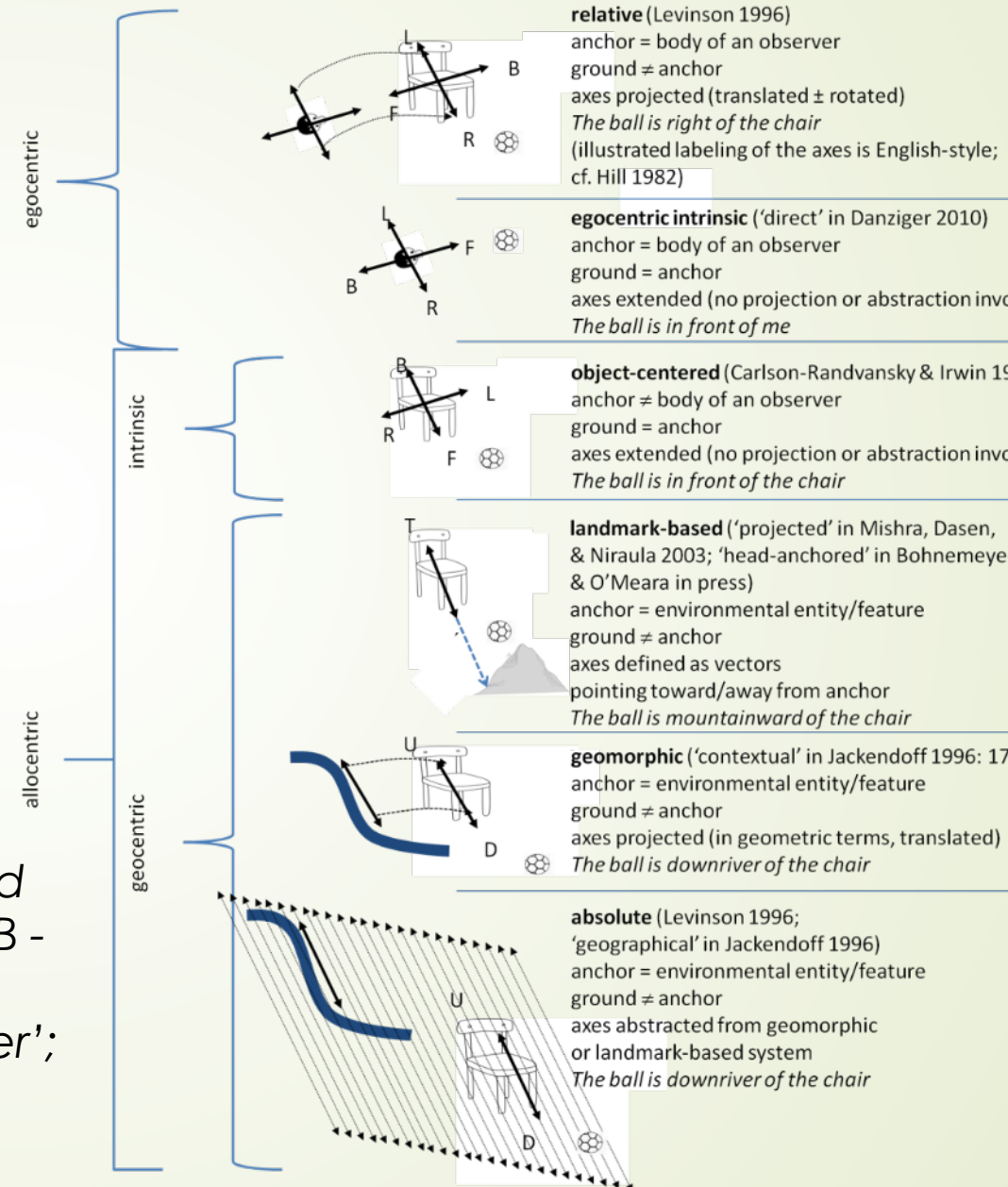
- collect further data on linguistic vs. environmental determinants of reference use

frames of reference:
alternative classifications
and subtypes

Figure 5. Reference frame types and their classification (A - 'away from', B - 'back', D - 'downriver', F - 'front', L - 'left', R - 'right', T - 'toward', U - 'upriver'; Bohnemeyer & Levinson ms.)

classification by anchor alone
(e.g., Carlson-Radvansky & Irwin 1993; Wassmann & Dasen 1998; Li & Gleitman 2002; *inter alia*)

classification generation (L Pederson 1998; Bohnemeyer *inter alia*)



MesoSpace (cont.)

- Tasks used by MesoSpace to study reference frames
 - Ball & Chair (photo stimuli, referential communication task)
 - Talking Animals (3-D toy animal stimuli, referential communication task)
 - New Animals (3-D toy animal stimuli, recall and recreate array of animals)

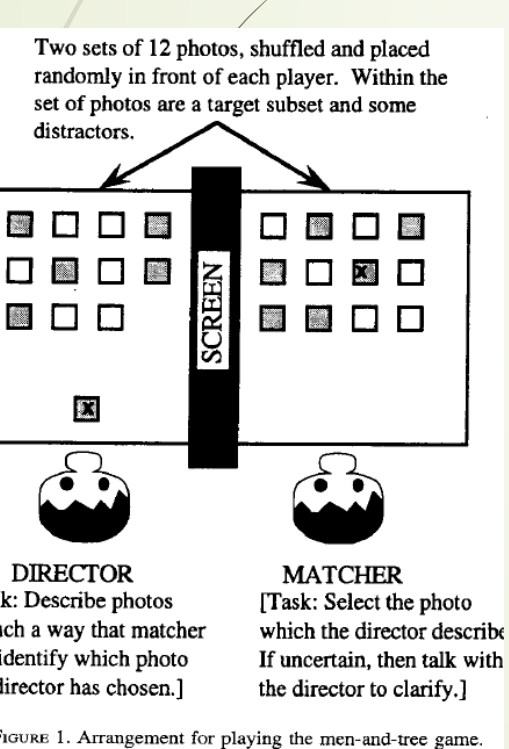


Figure 7. One of four Talking Animals trials

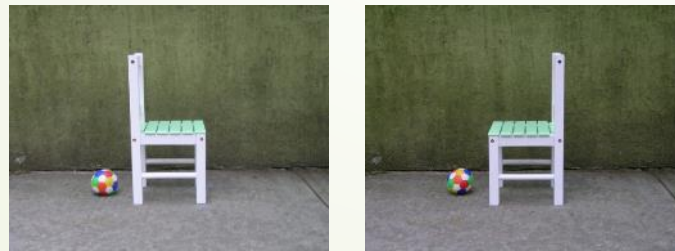


Figure 8. Two of the Ball & Chair photos, featuring an intrinsic contrast

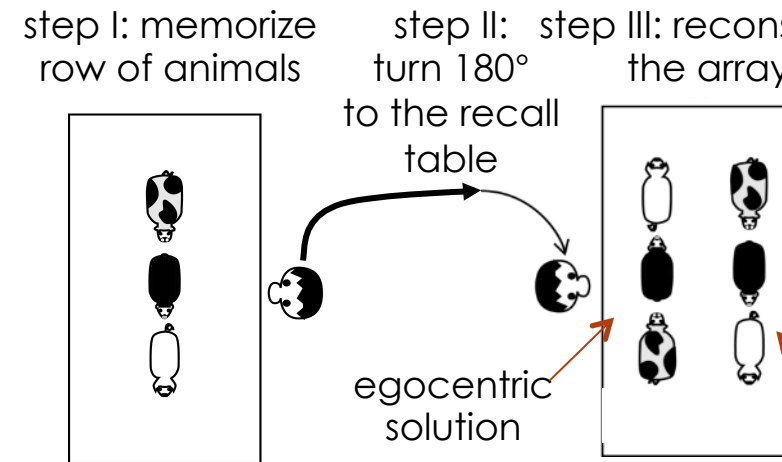


Figure 9. Animals-in-a-Row: description (Levinson 2003)

- Evaluating possible contributing factors - independent variables in mixed effect regression models (GLMMs)
 - Participants' age, education, L2 use, reading and writing frequency
 - Self-reported via demographic survey
 - Population Geography
 - Population density based on Google Earth area of municipality and population from census (INEGI 2010)
 - Local Topography
 - In Mexico: Classification of geomorphological regions (Hernández Santana et al. 2007)
 - World-wide: Improved Hammond Classification of landforms (ESRI, 2011; ArcGIS software)

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2

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Topography in MesoSpace

- ▶ Two classifications have been used to represent local topography in multivariate statistical analyses
 - ▶ Geomorphological regions (New Atlas of Mexico, Hernández Santana et al. 2007)
 - ▶ Bohnemeyer et al. 2014, 2015, ms.
 - ▶ Improved Hammond classification of landforms (ESRI 2011 ArcGIS)
 - ▶ Moore et al ms.

Topography in MesoSpace (cont.)

4

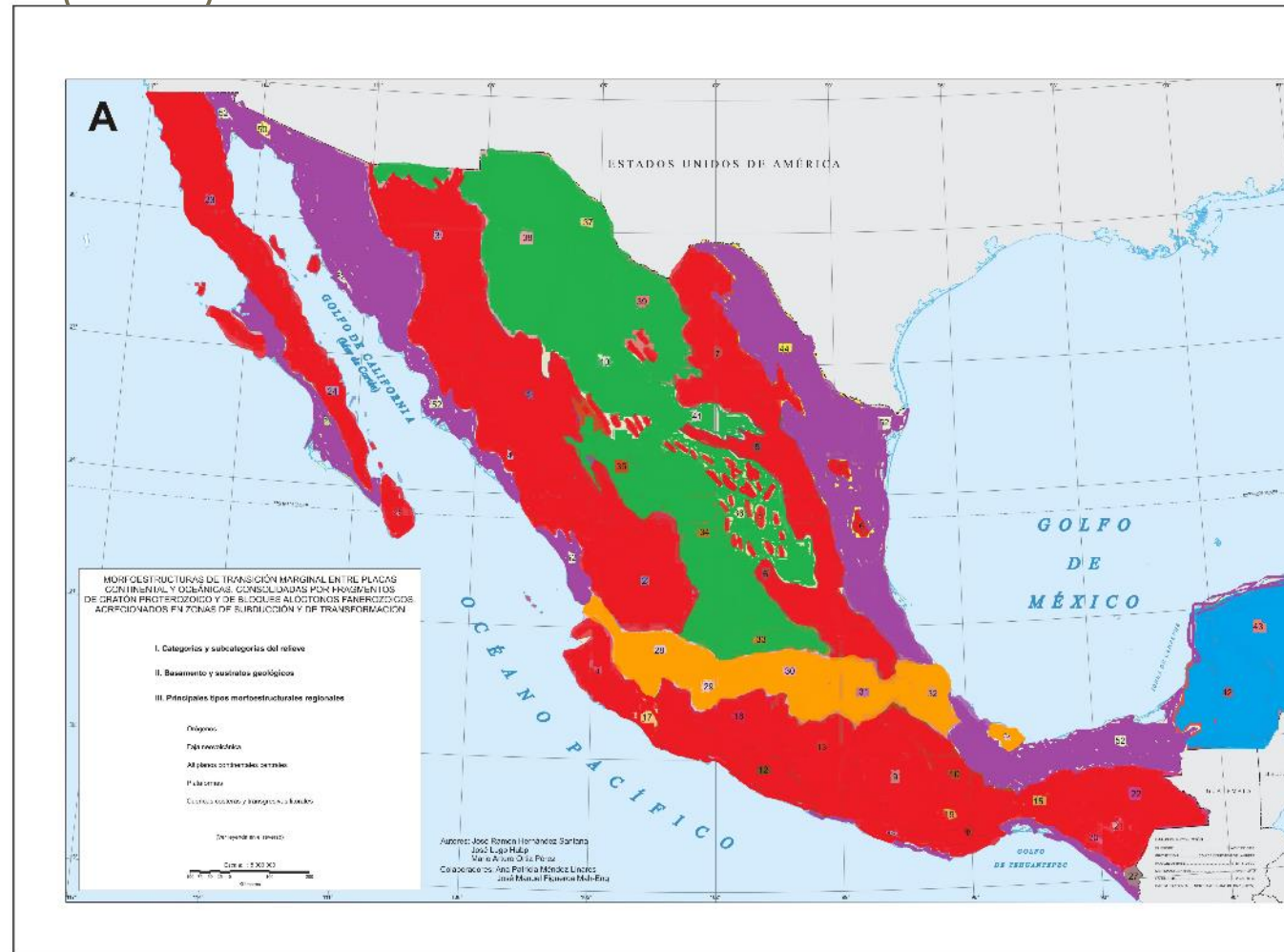
- orogenic belts
- volcanic belts
- central high plateaus
- continental shelf
- coastal basins and littoral transgressions

MORFOESTRUCTURAS REGIONALES

NA III 1



MORFOESTRUCTURAS REGIONALES



Geomorphological regions (New Atlas of Mexico, Hernández S et al. 2007)

Figure 10. (left) Mapa NA III_1. From Hernández Santana et al. **Figure 11.** (above) Modified 5-way classification used in previous MesoSpace classifications

Topography in MesoSpace (cont.)

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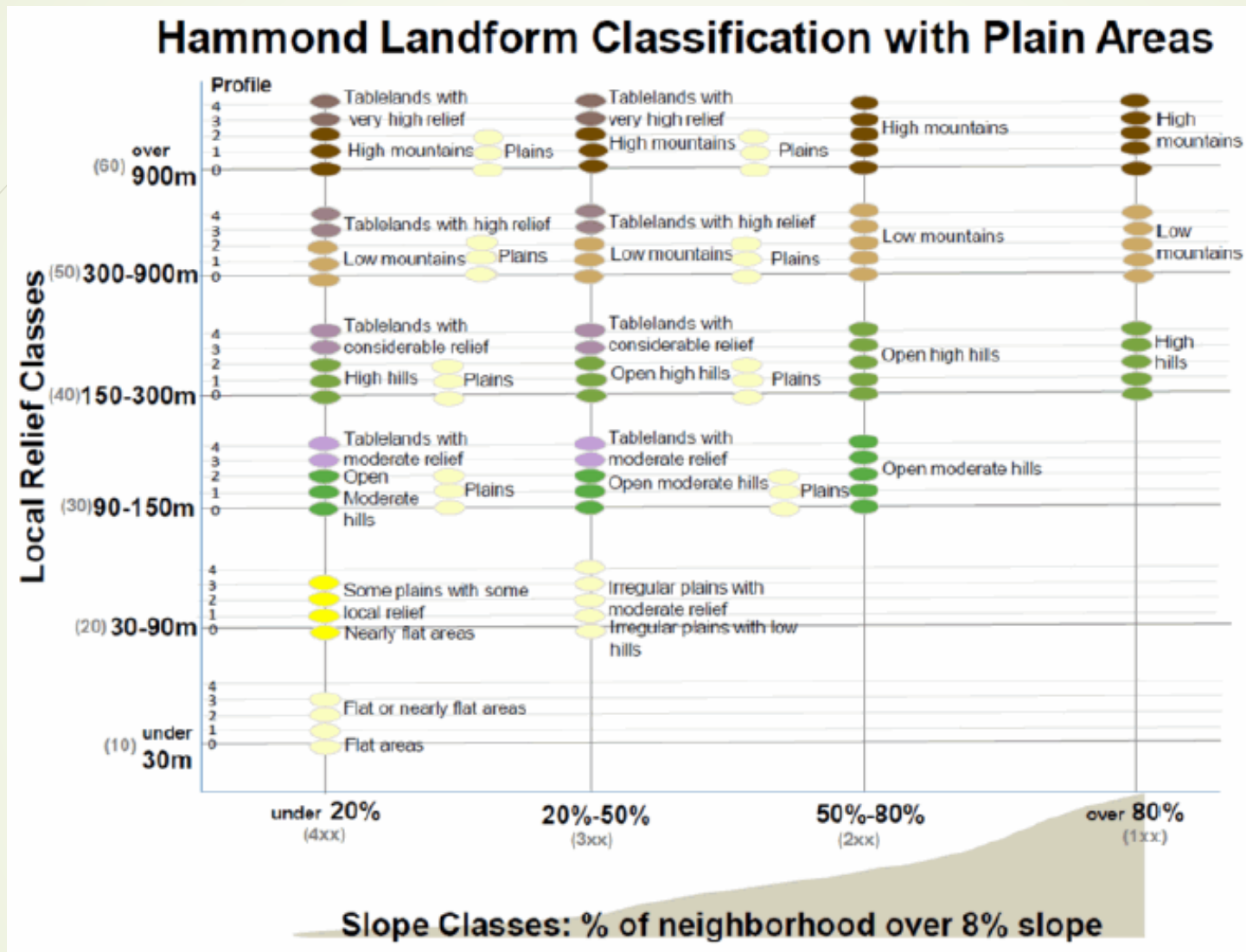


Figure 12. World Landforms – Improved Hammond Method (ESRI 2011)
 (<http://www.arcgis.com/home/item.html?id=cd817a746aa7437cbd72a6d39cdb4559>)

6

findings I: Ball & Chair

1. Geomorphological classification of field sites
B&C study (N = 53 x 2 participants)

Age	Locality	Geomorphological Classification
	Chacoma, Chiapas	orogenic belt
	Yaxley, Quintana Roo	continental shelf
	Felipe Carrillo Puerto, Quintana Roo	continental shelf
	San Pedro y San Pablo Ayutla, Oaxaca	orogenic belt
	San Ildefonso Tultepec, Queretaro	volcanic belt
o	La Ventosa, Oaxaca	coastal basin
na	Santa Fe de la Laguna, Michoacan	volcanic belt
	El Desemboque (de los seris), Sonora	coastal basin
	Rosita, Región Autónoma del Atlántico Norte	coastal basin
)	San Miguel Balderas, Mexico State	volcanic belt
)	Chimalacatlan, Tlaquiltenango, Morelos	orogenic belt
ua)	Rosita, Región Autónoma del Atlántico Norte	coastal basin
na)	Barcelona, Spain	coastal basin



Figure 13. Two of the Ball & Chair photos, featuring an intrinsic

Table 2. Summary of the four regression models for B&C responses using reading frequency to explain literacy. Models that include L1-Spanish speakers exclude L2 use as a predictor variable. (Significance codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.')

		Models		
		1	2	3
Sample	L1-SPANISH SPEAKERS INCLUDED	Yes	Yes	No
Dependent variable	GEOCENTRIC	Yes	No	Yes
	RELATIVE	No	Yes	No
Effects	LANGUAGE GROUP	***	***	
	L2-SPANISH USE			
	LITERACY			
	TOPOGRAPHY	*		
	POPULATION DENSITY			*

Bottom line

- More relative frame use in coastal basins
- More geocentric frame use in volcanic belts
- Population density positively correlated with geocentrism, negatively w/ relative

MesoSpace (cont.)

7 findings II: Talking Animals

3. Field sites of the TA study by pop. density and geomorphology (N = 343 x 2)

Site	Locality	Country	Density	Density Log Scale	Topographic Classification
Setagaya	Setagaya	Japan (Mainland)	15551	4.19	flat
Southern Min Chinese	Taipei	Taiwan	9949	4.00	flat
Southern Min Chinese	Taipei	Taiwan	9949	4.00	flat
Naha	Naha	Japan (Okinawa)	8244	3.92	hills
Buffalo	Buffalo	United States	2569	3.41	flat
Yomitan	Yomitan	Japan (Okinawa)	1200	3.08	hills
Southern Min Chinese	Tainan	Taiwan	855	2.93	flat
Long Mỹ	Long Mỹ	Vietnam	406	2.61	flat
Japan (Mainland/Okinawa)	Fujinomiya	Japan (Mainland)	339	2.53	low mountains
	Aizuwakamatsu	Japan (Mainland)	321	2.51	low mountains
	Nago	Japan (Okinawa)	293	2.47	low mountains
	Miyakojima	Japan (Okinawa)	268	2.43	hills
	Yonaguni	Japan (Okinawa)	58	1.76	hills
	Shisho	Japan (Mainland)	49	1.69	low mountains
Apotec	La Ventosa	Mexico	5	0.70	flat
	Juchitán de Zaragoza	Mexico	5	0.70	flat
Mexico	Yaxley	Mexico	2	0.30	flat
	Felipe Carillo Puerto	Mexico	2	0.30	flat



Figure 14. C
four Talking
trials

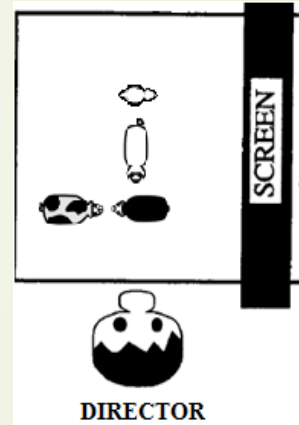


Figure 15. D
the Talking A
task (adapted
Pederson et
562)

MesoSpace (cont.)

8 findings II: Talking Animals (cont.)

Figure 4. Regression models of the Talking Animals data: summary of effects
 Sig. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1)

Independent variable	Literacy variable		Independent variables (fixed effects)				
	Writing	Reading	L1	L2 use	Literacy	Topography	Pop. density
Centric	Yes	No	***		**		***
	No	Yes	**	•	***		***
De-centric	Yes	No	***		***	•	**
	No	Yes	***		**	•	**



Figure 14. Four Talking Animals trials

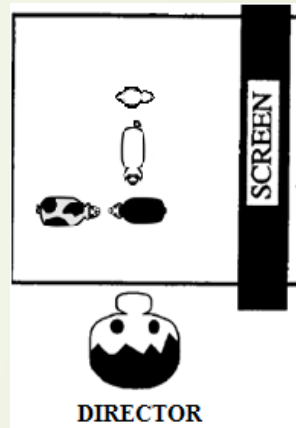


Figure 15. Director interface for the Talking Animals task (adapted from Pederson et al., 2007, p. 562)

MesoSpace (cont.)

9

findings III: New Animals

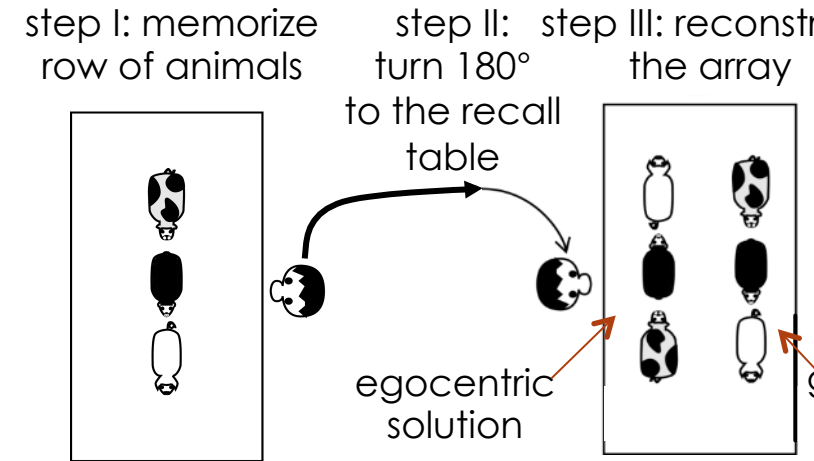


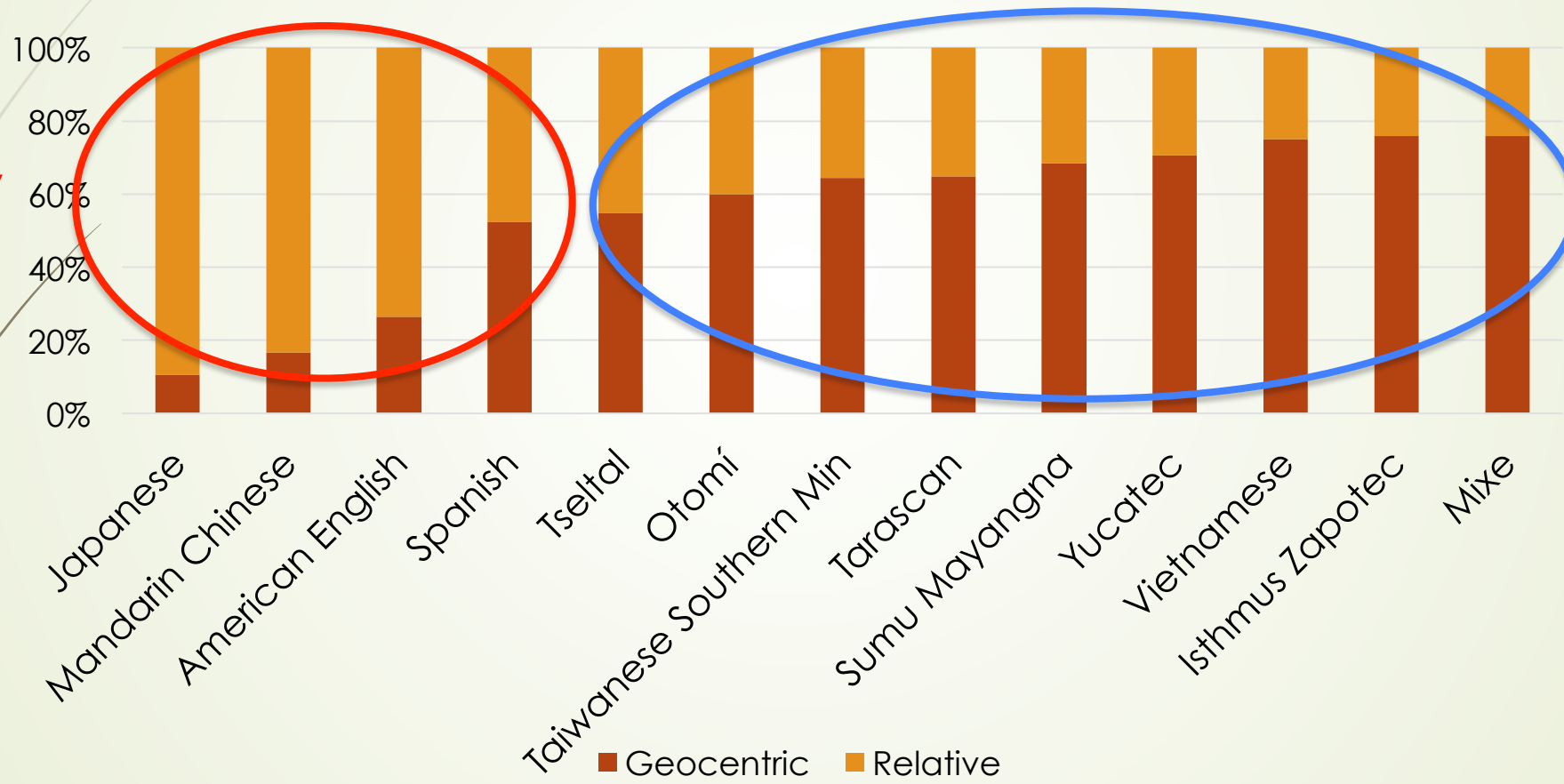
Figure 16. *Animals-in-a-Row*: description (Levinson 2003)

Table 5. Participants whose responses were included in the analysis, by language, age, and sex

	American English	Japanese	Mandarin Chinese	Mixe	Otomí	Spanish	Sumu Mayangnga	Taiwanese Southern Min	Tarascan	Tseltal	Vietnamese	Yucatec	Ist'at Zapotec
male	12	33	2	3	1	11	4	2	8	9	4	7	
female	8	15	7	9	4	21	6	19	6	5	16	10	
<30	2	17	0	8	5	18	6	19	8	6	8	11	
30	19	31	9	4	0	14	4	2	6	8	12	6	
Total	21	48	9	12	5	32	10	21	14	14	20	17	

findings III: New Animals (cont.)

New Animals Responses - Facing Direction



istically centric

linguistic geocentric

Figure 17. Response type frequency by L1

1 findings III: New Animals (cont.)

Dependent variable	Literacy variable			Independent variables (fixed effects)		
	Writing	Reading	L1	Literacy	Topography	Pop. density
Egocentric	Yes	No	***		**	***
	No	Yes	***		*	***

Table 6. Regression models of the New Animals data: summary of effects
 (Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1)

Bottom line

- Field sites featuring hilly topography show sig. less egocentric u than flat regions
- Population density positively correlated w/ egocentrism

MesoSpace (cont.)

2

➤ discussion

- both population density and topography confirmed
 - as independent factors influencing reference frame use in both discourse and recall memory
- by hypothesis, the effect of population density is primarily mediated by infrastructure
 - egocentrism more efficient for navigating urban roadways
- the effect of topography has been hypothesized to be mediated by the availability of salient potential 'anchors'
 - such as physiogeographic gradients and natural landmarks
 - cf. Polian & Bohnemeyer 2011; Li & Gleitman 2002
- questions
 - what is the role of culture in these geographic effects?
 - at what level of granularity do such effects begin to matter?

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3

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Space and topography in Diidxa za

- Diidxa za (Isthmus Zapotec)
 - Otomanguenan, VSO, tonal, ~100,000 speakers (INEGI 2010 census)
- Reference frame use in Isthmus Zapotec
 - Pérez Báez (2011) reference frame use in recall and discourse
 - 2-D stimuli in La Ventosa
 - Strong geocentric preference, based on prevailing winds
 - Moore (2016) frame use in recall and discourse
 - 3-D stimuli in La Ventosa and Juchitán
 - Confirmed geocentric preference
 - Significant variation exists between communities
 - Variation also exists in degree of preference for geocentric over egocentric encoding in memory

Space and topography in Diidxa za (cont.)

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► Isthmus Zapotec field sites

- La Ventosa
- Juchitán de Zaragoza
- Santa María Xadani

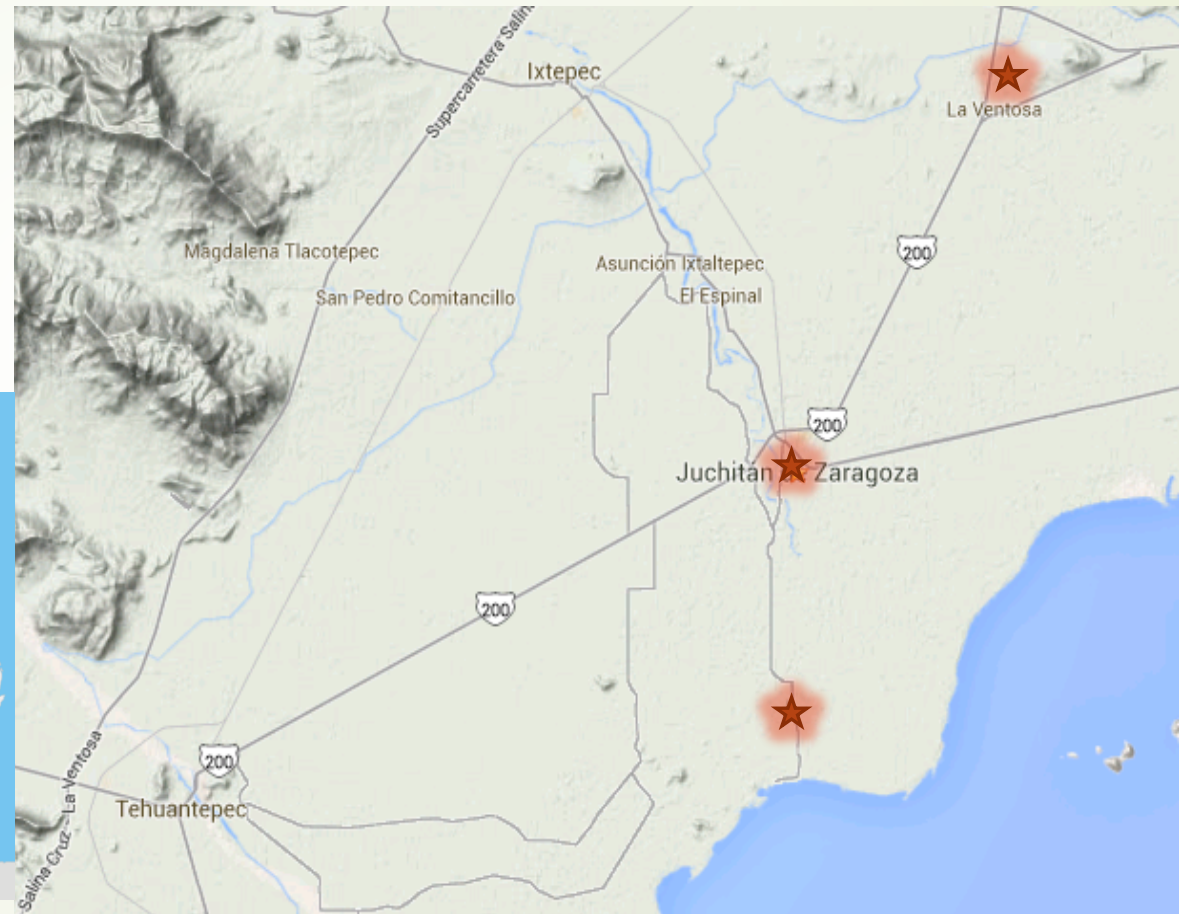


Figure 18. Oaxaca, Mexico, and three field sites in the Isthmus of Tehuantepec

Space and topography in Diidxa za (cont.)

Distribution of frame type responses in discourse by Community

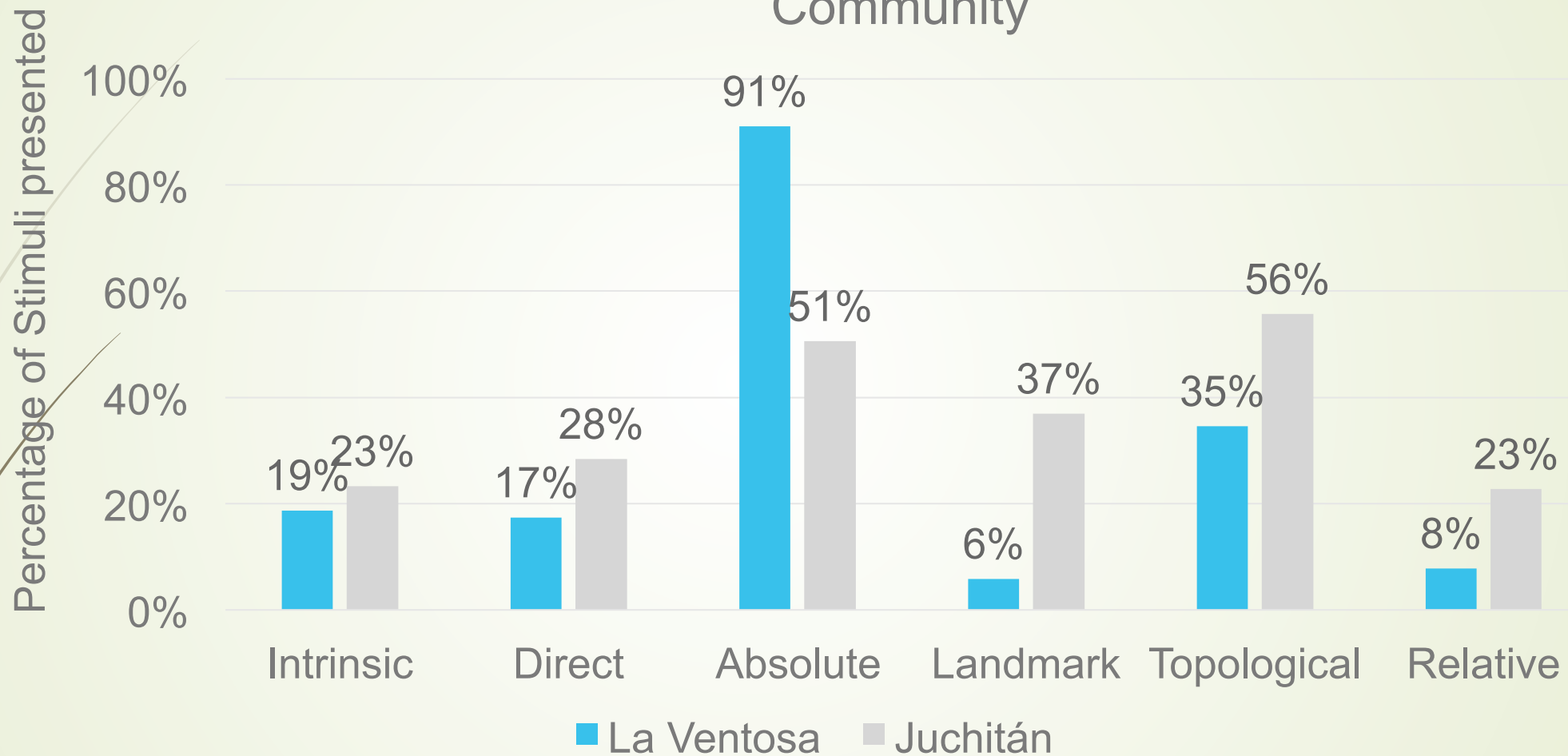


Figure 19. Frame use in discourse. Responses to Talking Animals (Moore 2016)
(N = 23 x 2 participants)

Space and topography in Diidxa za (cont.)

7

- Generalized linear mixed-effects models (Lme4 package in R)
- Factors included in Models
 - Education level
 - Reading and writing frequency
 - Frequency of L2 use (Spanish)
 - Community membership
- Dependent Variables
 - L2, Education, Reading, Writing, Community membership
- Independent Variables
 - Reference frame types
- Community membership only significant factor in predicting reference frame use

Space and topography in Diidxa za (cont.)

8

Dependent variable	Literacy variable		Independent variables (fixed effects)			
	Writing	Reading	Community	L2 Use	Education	Literacy
Geocentric	Yes	No	***			
	No	Yes	***			
Relative	Yes	No	**			
	No	Yes	**			
Absolute	Yes	No	***			
	No	Yes	***			
Direct	Yes	No	**			
	No	Yes	**			*
'Landmark-based'	No	Yes	***			

Table 8. Regression models of the Talking Animals Zapotec data: summary of effects (Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1)

(model details in Appendices)

Space and topography in Diidxa za (cont.)

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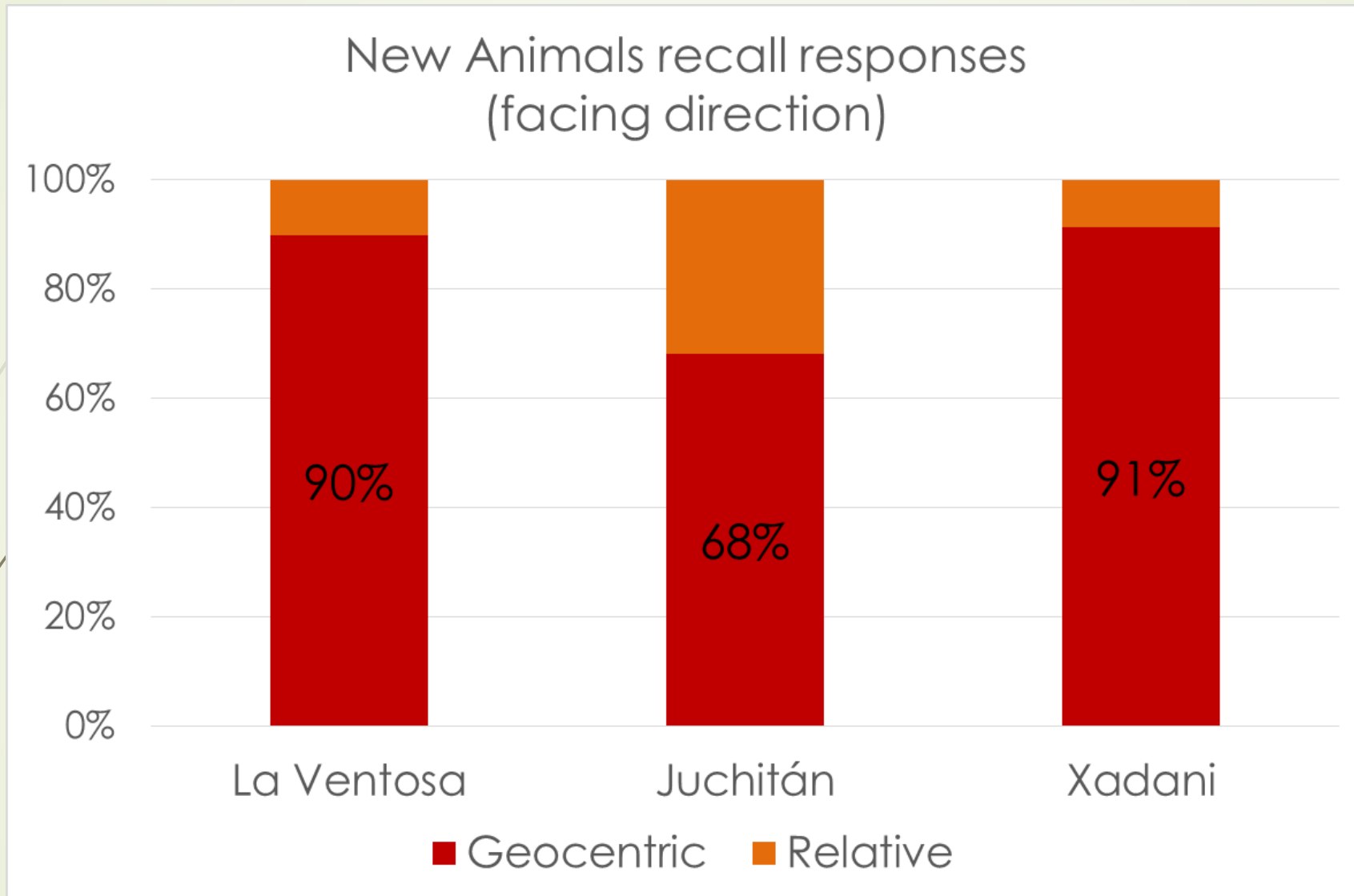


Figure 20. Frame use in recall memory. Responses to *New Animals*. (N = 28 participants)

Space and topography in Diidxa za (cont.)

0

- ▶ Large-scale topographic classifications don't capture the variation in local landscape throughout the Isthmus
 - ▶ All three are coastal (Hernandez et al) or flat (ESRI)
- ▶ Yet, variation in frame use in discourse and memory exists between communities

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1

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2

- the impact of the environment on cognition is mediated by culture
 - community-specific practices evolve around salient environmental gradients
 - inter-community variation such as witnessed in the Isthmus is the result of this
 - another example: Bali (Wassm

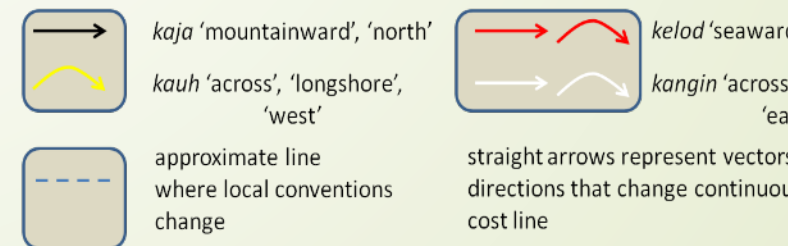
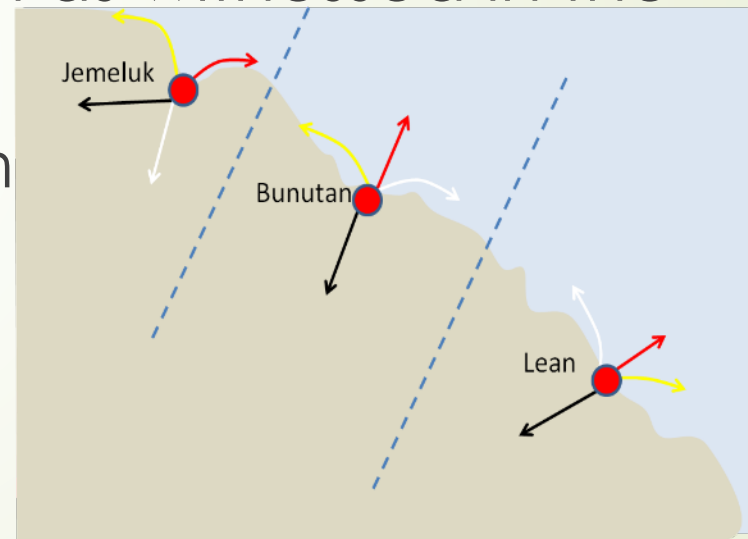


Figure 21. Three neighboring villages on the North-East peninsula of Bali using the same set of geocentric terms each based on different local conventions (Bohnmeyer et al. ms, based on a detail from Wassman & Dasen 1998: 698)

3

▶ studying the cultural mediation of environmental forces: ethnophysiology

- ▶ cf. Bohnermeyer 2002; Burenhult & Levinson 2008; Johnson & Hur 2010; Mark & Turk 2003; Mark et al 1999; O'Meara 2010; Smith & Mark 2003; *inter alia*

Cultural mediation: ethnophysiology (cont.)

4

- ➔ one salient environmental feature that's not directly captured in a topographic classification: prevailing winds
- ➔ the Isthmus of Tehuantepec has year-round prevailing northwesterly winds
- ➔ these are indirectly shaped by the relief, in that a gap in the North American cordillera creates a giant wind tunnel

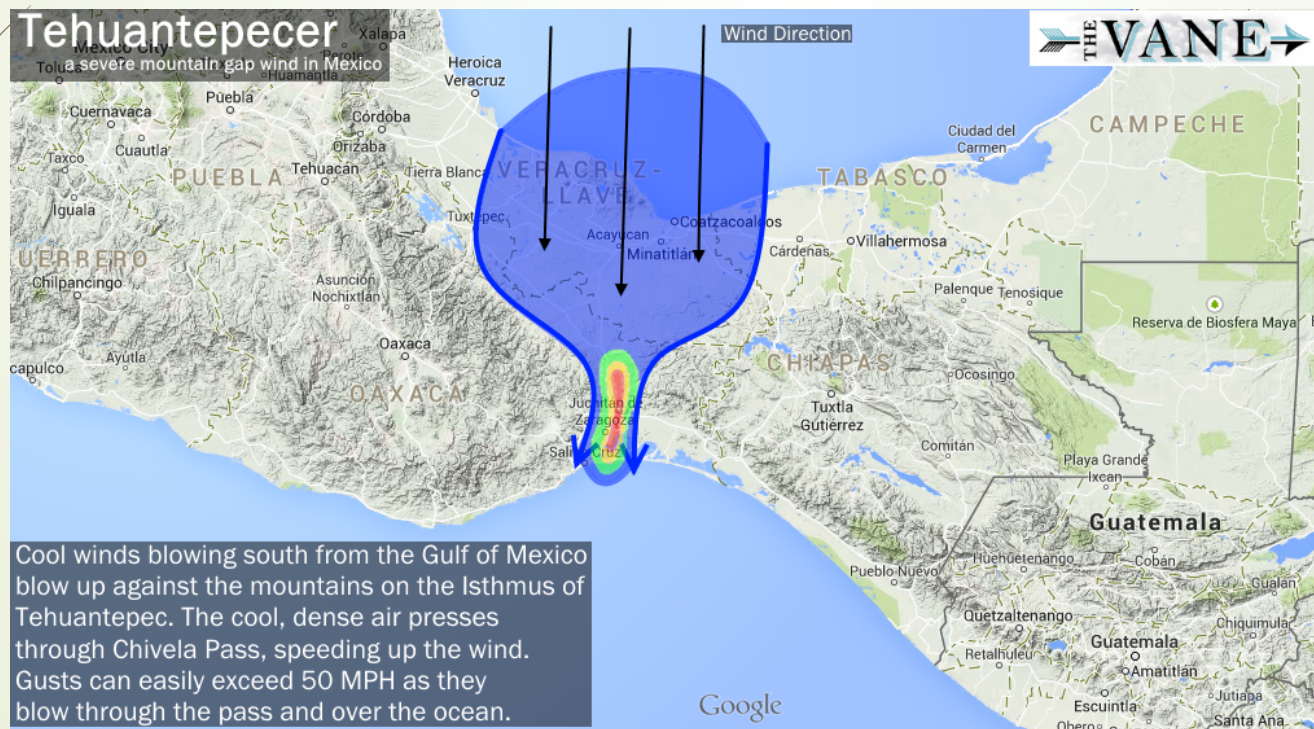


Figure 22. Diagram of Tehuantepecer winds (<https://i.kinja-img.com/gaw/image/upload/hvttkme469ebgy6l1mhb.PNG>)

5

- these winds provide the Isthmus with a salient geocentric cue that is readily accessible outdoors
- although this cue is available throughout the region, it appears to play a more prominent cultural role in La Ventosa
 - as reflected in the name ('the windy one') and the ubiquity of wind farms, which are a source of great public controversy



Figure 23. La Ventosa windfarms, taken from the Sierra Sur foothills

▶ evidence from ethnophysiological elicitation confirms

▶ data sources

▶ Lexical inventory/salience of features

- ▶ Considerable variation exists between communities (Listing task: 10 speakers, 3 communities, avg. 12 terms/person)

▶ Landmarks in direction-giving

- ▶ Route description task: 5 pairs per community

▶ Landmarks/environmental features used in descriptions of small scale space (Talking Animals)

- ▶ Is there significant variation between communities in

- ▶ (i) extent of Geocentric use

- ▶ (ii) type of Geo use (local/manmade landmark, environmental landmark, absolute/cardinal system)??

- ▶ If so, this variable in a statistical model could independently predict use (in discourse & recall)

- ▶ vs. community membership, or other factors

Findings

- Most frequent responses to a landscape term listing task (it that occurred 5+ times)
 - Prompt words were: *dani*, *guiigu*, *guiixhi*

La Ventosa	Juchitán de Zaragoza	Santa María Xadani
Yaga 'tree' (13)	Yaga 'tree' (12)	Nisa do' 'sea' (10)
Dani 'hill' (13)	Guiigu 'river' (9)	Dani 'hill' (9)
Bi 'wind' (12)	Dani 'hill' (9)	Guiixhi 'forest/jungle' (7)
Mani' 'animal' (9)	Mani 'animal' (8)	Ranya 'milpa' (6)
Nisa 'water' (7)	Yuu 'house' (5)	Guiigu 'river' (6)
Guiigu 'river' (7)	Nisa 'water' (5)	Bize* 'well' (6)
Nisa do' 'sea' (5)	Guixi 'trash' (5)	Esteru 'marsh/swamp' (5)
Guie 'rock/soil' (5)		

Table 9.
task resp
by comr

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8

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Conclusions

confirmed: geography influences spatial language & cogn

- ▶ MesoSpace has found evidence of effects of population density and topography
 - ▶ as independent factors influencing reference frame use in both discourse and recall memory
- ▶ by hypothesis, these effects are primarily mediated by infrastructure and the local availability of potential salient anchors

Conclusions (cont.)

- the challenge of topographic classification
 - there is as yet no universally agreed upon system of variables
 - that jointly capture the morphology of the Earth' crust everywhere
 - MesoSpace has successfully pioneered the application of the Improved Hammond Classification
 - for the search of cognitive effects of geography

Conclusions (cont.)

- ▶ studying cultural mediation between environment and cognition/behavior: ethnophysiology
- ▶ community-specific practices evolve around salient environmental gradients

Thank you!

Xquixe pe' laatu!

Special thanks to...

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Appendices:

Model print outs for Zapotec Talking Anima

Geocentric, Write -> Community Membersh

9

Generalized linear mixed model fit by maximum likelihood (Laplace Approximation) ['glmerMod']

Family: binomial (logit)

Formula: GEOJ ~ CD + L2D + EdD + WriteD + (1 | ID)

Data: mydata

AIC	BIC	logLik	deviance	df.resid
444.1	471.3	-216.0	432.1	682

Scaled residuals:

Min	1Q	Median	3Q	Max
-3.9080	0.1834	0.2125	0.3154	0.9009

Random effects:

Groups Name	Variance	Std.Dev.
ID (Intercept)	0.9783	0.9891

Number of obs: 688, groups: ID, 89

Fixed effects:

	Estimate	Std. Error	z value	Pr(> z)	
(Intercept)	1.54690	0.51832	2.984	0.002841	**
CDLAV	1.61751	0.42981	3.763	0.000168	***
L2D	-0.20254	0.30456	-0.665	0.506043	
EdD	0.16145	0.28222	0.572	0.567270	
WriteD	0.04705	0.20213	0.233	0.815957	

Correlation of Fixed Effects:

	(Intr)	CDLAV	L2D	EdD
CDLAV	-0.211			
L2D	-0.573	-0.429		
EdD	0.110	0.063	-0.250	
WriteD	-0.467	0.205	-0.042	-0.457



Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Geocentric, Read-> Community Members

0

```
Generalized linear mixed model fit by maximum likelihood (Laplace  
Approximation) ['glmerMod']
```

```
Family: binomial ( logit )
```

```
Formula: GEOJ ~ CD + L2D + EdD + ReadD + (1 | ID)
```

```
Data: mydata
```

AIC	BIC	logLik	deviance	df.resid
443.9	471.1	-215.9	431.9	682

```
Scaled residuals:
```

Min	1Q	Median	3Q	Max
-4.0174	0.1844	0.2050	0.3090	0.9058

```
Random effects:
```

Groups Name	Variance	Std.Dev.
ID (Intercept)	1.001	1.001

```
Number of obs: 688, groups: ID, 89
```

```
Fixed effects:
```

	Estimate	Std. Error	z value	Pr(> z)	
(Intercept)	1.7078	0.5047	3.384	0.000715	***
CDLAV	1.5518	0.4295	3.613	0.000303	***
L2D	-0.1581	0.3154	-0.501	0.616193	
EdD	0.2426	0.2712	0.894	0.371115	
ReadD	-0.1133	0.2172	-0.522	0.602014	

```
---
```

```
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
Correlation of Fixed Effects:
```

	(Intr)	CDLAV	L2D	EdD
CDLAV	-0.199			
L2D	-0.493	-0.450		
EdD	0.035	0.097	-0.184	
ReadD	-0.408	0.184	-0.248	-0.363

Relative, Write -> Community Membership

1

```
Generalized linear mixed model fit by maximum likelihood (Laplace
approximation) ['glmerMod']
Family: binomial ( logit )
Formula: RELJ ~ CD + L2D + EdD + WriteD + (1 | ID)
Data: mydata
```

```
      AIC      BIC    logLik deviance df.resid
458.3    485.5   -223.2    446.3     682
```

Standardized residuals:

```
      Min       1Q   Median       3Q      Max
-3.4189 -0.2983 -0.2002 -0.1796  3.7679
```

Random effects:

```
Groups Name          Variance Std.Dev.
ID (Intercept)  1.313      1.146
Number of obs: 688, groups: ID, 89
```

Fixed effects:

```
              Estimate Std. Error z value Pr(>|z|)
(Intercept) -1.86458    0.57958  -3.217  0.00129 **
CDLAV        -1.33916    0.44367  -3.018  0.00254 **
L2D          -0.03556    0.32140  -0.111  0.91189
EdD           0.06220    0.30505   0.204  0.83843
WriteD       0.16911    0.22380   0.756  0.44987
```

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Correlation of Fixed Effects:

```
              (Intr) CDLAV  L2D    EdD
CDLAV        -0.309
L2D          -0.569 -0.324
EdD           0.127  0.022 -0.232
WriteD       -0.494  0.206 -0.002 -0.508
```

Relative, Read-> Community

```

Generalized linear mixed model fit by maximum likelihood (Laplace
approximation) ['glmerMod']
Family: binomial ( logit )
Formula: RELJ ~ CD + L2D + EdD + ReadD + (1 | ID)
Data: mydata

```

```

      AIC      BIC    logLik deviance df.resid
458.9    486.1   -223.4    446.9     682

```

Standardized residuals:

```

      Min       1Q   Median       3Q      Max
-0.4583 -0.2931 -0.1996 -0.1780  3.6227

```

Random effects:

```

Groups Name          Variance Std.Dev.
ID          (Intercept) 1.356    1.164
Number of obs: 688, groups: ID, 89

```

Fixed effects:

```

              Estimate Std. Error z value Pr(>|z|)
(Intercept) -1.63179    0.56022  -2.913  0.00358 **
CDLAV        -1.42691    0.45228  -3.155  0.00161 **
L2D          -0.02981    0.32952  -0.090  0.92792
EdD          0.18849    0.28482   0.662  0.50811
ReadD       -0.02197    0.23043  -0.095  0.92404

```

```

Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

Correlation of Fixed Effects:

```

              (Intr) CDLAV  L2D    EdD
CDLAV  -0.316
L2D    -0.503 -0.362
EdD    0.018  0.042 -0.178
ReadD -0.425  0.244 -0.185 -0.371

```

Absolute, Read -> Community Membership

3

```
Generalized linear mixed model fit by maximum likelihood (Laplace
approximation) ['glmerMod']
Family: binomial ( logit )
Formula: ABSJ ~ CD + L2D + EdD + ReadD + (1 | ID)
Data: mydata
```

```
      AIC      BIC    logLik deviance df.resid
483.2    510.4   -235.6    471.2     682
```

Standardized residuals:

```
      Min       1Q   Median       3Q      Max
-0.7336  0.1129  0.1596  0.2980  2.1919
```

Random effects:

```
Groups Name          Variance Std.Dev.
ID (Intercept)  3.058    1.749
Number of obs: 688, groups: ID, 89
```

Fixed effects:

```
              Estimate Std. Error z value Pr(>|z|)
(Intercept) -0.23793    0.70073  -0.340    0.734
CDLAV        3.47602    0.62308   5.579 2.42e-08 ***
L2D         -0.17488    0.43581  -0.401    0.688
EdD          0.50912    0.37601   1.354    0.176
ReadD        0.04957    0.29669   0.167    0.867
```

```
---
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Correlation of Fixed Effects:

```
          (Intr) CDLAV  L2D    EdD
CDLAV  -0.270
L2D    -0.505 -0.403
EdD     0.011  0.075 -0.169
ReadD  -0.398  0.182 -0.234 -0.352
```

Absolute, Read-> Community Membership

4

```
Generalized linear mixed model fit by maximum likelihood (Laplace
approximation) ['glmerMod']
Family: binomial ( logit )
Formula: ABSJ ~ CD + L2D + EdD + WriteD + (1 | ID)
Data: mydata
```

```
      AIC      BIC    logLik deviance df.resid
481.4    508.6   -234.7   469.4     682
```

Sampled residuals:

```
      Min      1Q   Median      3Q      Max
9977  0.1081  0.1667   0.2916  2.2895
```

Random effects:

```
Groups Name          Variance Std.Dev.
(Intercept)  2.902      1.703
Number of obs: 688, groups: ID, 89
```

Fixed effects:

```
      Estimate Std. Error z value Pr(>|z|)
(Intercept)  -0.6605     0.7201  -0.917   0.359
CDLAV         3.6022     0.6204   5.806 6.4e-09 ***
(Intercept)  -0.1787     0.4164  -0.429   0.668
(Intercept)   0.3026     0.3824   0.791   0.429
WriteD        0.3723     0.2744   1.357   0.175
```

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Correlation of Fixed Effects:

```
          (Intr) CDLAV  L2D    EdD
CDLAV    -0.304
L2D      -0.564 -0.379
EdD       0.085  0.032 -0.228
WriteD   -0.480  0.246 -0.048 -0.428
```

Direct, Write -> Community Membership

5

```
Generalized linear mixed model fit by maximum likelihood (Laplace
approximation) ['glmerMod']
Family: binomial ( logit )
Formula: DIRJ ~ CD + L2D + EdD + Writed + (1 | ID)
Data: mydata
```

```
AIC      BIC      logLik deviance df.resid
885.1    712.3    -336.5   673.1     682
```

Standardized residuals:

```
Min      1Q  Median      3Q      Max
-2.898 -0.5081 -0.3954 -0.3161  2.6862
```

Random effects:

```
Groups Name          Variance Std.Dev.
(Intercept) 0.3159    0.562
Number of obs: 688, groups: ID, 89
```

Fixed effects:

```
              Estimate Std. Error z value Pr(>|z|)
(Intercept) -1.5203      0.3634  -4.184 2.86e-05 ***
CDLAV        -0.7802      0.2792  -2.794  0.0052 **
L2D          0.3203      0.2046   1.566  0.1174
EdD         -0.2393      0.1820  -1.315  0.1887
Writed       0.1923      0.1323   1.453  0.1461
```

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Correlation of Fixed Effects:

```
(Intr) CDLAV  L2D    EdD
CDLAV  -0.277
L2D    -0.595 -0.386
EdD    0.110  0.049 -0.231
Writed -0.476  0.157  0.003 -0.471
```

Direct, Read-> Community Membership, Reading

```
Generalized linear mixed model fit by maximum likelihood (Laplace
approximation) ['glmerMod']
Family: binomial ( logit )
Formula: DIRJ ~ CD + L2D + EdD + ReadD + (1 | ID)
Data: mydata
```

```
          AIC      BIC    logLik deviance df.resid
1  583.0    710.2   -335.5    671.0     682
```

Standardized residuals:

```
      Min       1Q   Median       3Q      Max
1 3442 -0.5047 -0.3930 -0.3201  2.7717
```

Random effects:

```
Groups Name              Variance Std.Dev.
1  (Intercept)  0.3119    0.5585
Number of obs: 688, groups: ID, 89
```

Fixed effects:

```
              Estimate Std. Error z value Pr(>|z|)
1 (Intercept)  -1.5412     0.3464  -4.449 8.64e-06 ***
2 CDLAV        -0.7364     0.2784  -2.645 0.00816 **
3 L2D           0.2316     0.2078   1.115 0.26498
4 EdD          -0.2491     0.1736  -1.435 0.15143
5 ReadD        0.2806     0.1380   2.032 0.04210 *
```

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Correlation of Fixed Effects:

```
          (Intr) CDLAV  L2D    EdD
CDLAV  -0.285
L2D    -0.524 -0.404
EdD    0.028  0.065 -0.160
ReadD -0.396  0.167 -0.205 -0.381
```


Landmark, Read -> Community Membersh

7

```
Generalized linear mixed model fit by maximum likelihood (Laplace
approximation) ['glmerMod']
Family: binomial ( logit )
Formula: LANEJ ~ CD + L2D + EdD + ReadD + (1 | ID)
Data: mydata
```

```
AIC      BIC      logLik deviance df.resid
388.9    416.1    -188.5   376.9     682
```

Standardized residuals:

```
Min      1Q    Median      3Q      Max
-5.398 -0.1700 -0.0926 -0.0536  4.2369
```

Random effects:

```
Groups Name          Variance Std.Dev.
(Intercept) 4.899      2.213
Number of obs: 688, groups: ID, 89
```

Fixed effects:

```
Estimate Std. Error z value Pr(>|z|)
(Intercept) -0.4116    0.8938  -0.461    0.645
LAV          -4.0363    0.8656 -4.663 3.12e-06 ***
              0.5749    0.5807  0.990    0.322
              -0.7611    0.5287 -1.440    0.150
ID           -0.5531    0.3895 -1.420    0.156
```

```
Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

Correlation of Fixed Effects:

```
(Intr) CDLAV  L2D    EdD
CDLAV -0.164
L2D   -0.521 -0.490
EdD   0.023  0.113 -0.196
ReadD -0.379  0.279 -0.216 -0.329
```