Predictors of Native-like Lexical Categorization in Chinese Learners of English

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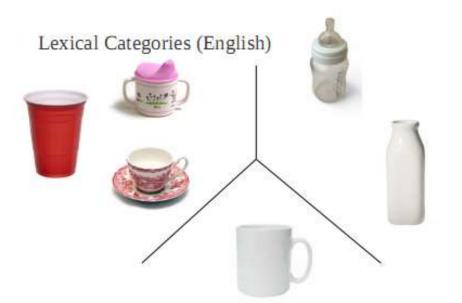


The Center for Language Science

Lexical Categorization

Languages carve up the world of objects and events into named categories that suggest some similarity between their diverse constituents

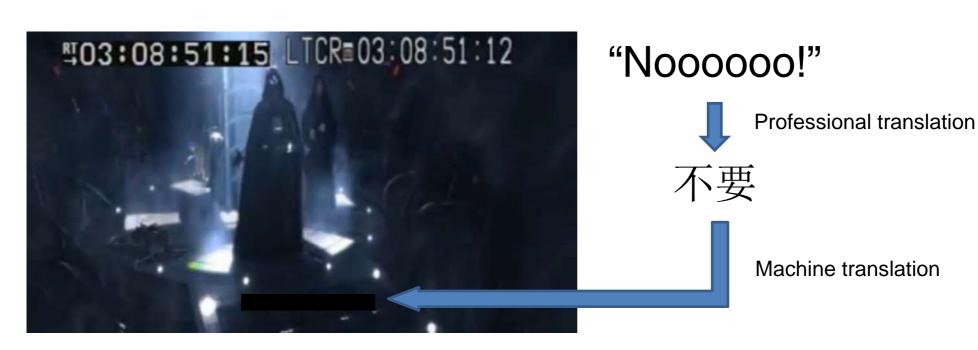
- -Color (Landar, Ervin, Horowitz, 1960; Caskey-Sirmons & Hickerson, 1977; Kay & Regier, 2007)
- -Furniture (Graham & Belnap, 1986, Malt, Sloman, & Gennari, 2003; Li, 2012)
- -Containers (Malt et al, 1999; Ameel et al, 2005)
- -Causation (Wolff & Ventura, 2009)



These categories can disagree!

Non-native-like categories are an example of semantic accent, or the transfer of norms from one language to another.

Semantic Accent









Convergence & Transfer

Lexico-semantic conflicts appear to be solved through compromise between languages

(Ameel et al, 2005)

- Using L1- or L2-specific patterns: transfer
- Similarity between L1 & L2 production: convergence



English-to-Chinese transfer



bottle

Chinese-to-English transfer

Goals of the Present Study

New conditions: How does bilingual interaction manifest

- ...in a more distant language pair?
- ...outside bilingual community?
- ...in L2 learners?

Parameterize native-like L2 categorization:

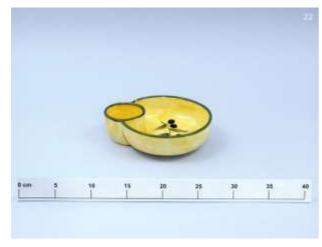
- Age of L2 onset (Li et al., 2007; Pavlenko & Malt, 2011)
- Immersion (Linck et al., 2009; Schmitt, 2010)
- L2 proficiency (Dong et al., 2005; Zinszer & Li, 2010)
- Code-Switching frequency (Wolff & Ventura, 2009)

Methods

Participants: Drawn from larger sample of Chinese-English bilingual undergraduate & graduate students, based on sufficient English proficiency:

- 15 students from Penn State University (USA)
- 20 students from Beijing Normal University (China)

Sample	Age	AOEE	LOR	English Prof	O-Span
PSU	18 - 23 y M = 20 y	5 - 15 y M = 9 y	0 - 9 y M = 3.3 y	2.8 - 7.0 M = $5.1 / 7$	43 - 57 M = 49
BNU	18 - 26 y M = 22 y	5 - 15 y M = 11y	~	1.3 - 5.5 $M = 4.0 / 7$	34 - 57 M = 49



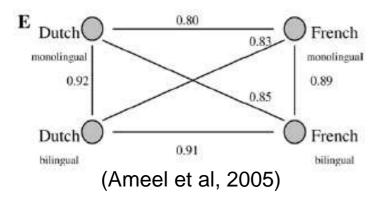
What is this?



这是什么?

Group Correlations

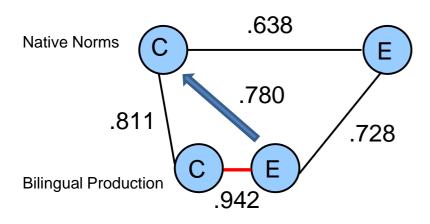
Hypothesis



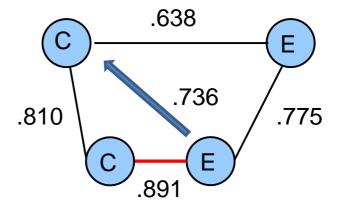
Results

Convergence (correlation within bilinguals): greater than baseline in both groups

Cross-language transfer: groups are relying on L1 to differing extent



Beijing Normal



Penn State

Predictors of Native-like English

Binomial logistic regression for P(native-English dominant name):

Object-wise variables:

- Naming Agreement by monolinguals in Chinese & English
- Number of proposed names by monolinguals in Chinese & English

Participant-wise variables:

- Age of earliest L2 (English) exposure
- Overall English proficiency
- Frequency of code-switching
- All interactions, non-significant interactions later removed

Predictors of Native-like English

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Object Variable log(Odds Ratio), p<0.05 Object Variable log(Odds Ratio), p<0.05

Chi. Norm Agreement (-1.56)

Eng. Norm Agreement (2.90)

Eng. Norm Agreement (4.32)

Chinese Norm Names (-0.19)

English Norm Names (-0.11)

- For un-immersed learners, L1 naming patterns are a significant detriment to L2 native-likeness.
- Immersed learners gain a dramatic advantage in native-likeness from native-speaker agreement.

Predictors of Native-like English

Beijing Normal

Penn State

Participant Variable log(Odds Ratio), p<0.05

Participant Variable log(Odds Ratio), p<0.05

Age of Exposure (-0.12)

English Proficiency (0.09), p=0.12

Code-Switch Frequency (-0.84)

AoEE x CS-Freq (0.07)

English Proficiency (0.56)

Code-Switch Frequency (0.45), p=0.08

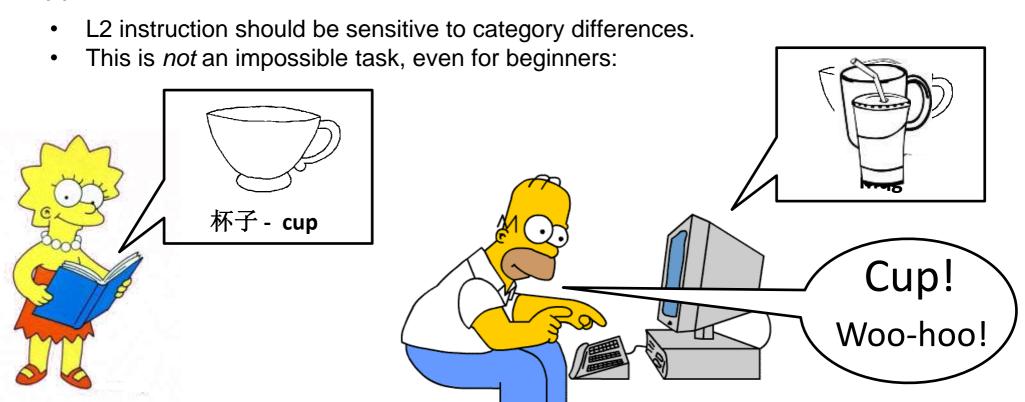
Profic x CS-Freq (-0.08), p=0.14

- Only in un-immersed learners does L2 native-likeness decline with increased age of earliest exposure and code-switching
- Self-rated L2 proficiency is only predictive of lexical category proficiency in immersed learners.
- Code-switching appears to have differential effects vs. immersion

Discussion

- Agreement among native speakers is highly predictive of L2 learners' success in acquiring native-like categorization
- Individual differences in language history can differentially affect immersed and un-immersed learners.

Application



Thank You & 谢谢

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