

# Cross-linguistic differences in lexical access and spoken word recognition

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## 0.1 Acknowledgments

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## 1 Background

### 1.1 Research Goals

- Learn more about the role of morphology in the mental lexicon
- That is, are morphemes stored separately in the lexicon and then combined to form words during lexical access, or are words stored whole in the lexicon?
- Extend previous research using open response spoken word recognition to bisyllabic words
- Compare context effects across two phonologically similar, yet morphologically diverse languages

### 1.2 Morphology

- The study of Lexical Access seeks to determine how the mental lexicon affects language processing.
- The role of morphology in the lexicon is studied widely in lexical access research
- Results from cross-linguistic research suggest that morphology plays different roles in lexical access based on the type of morphological system of the language
- Two classes of models differ in their predictions of how morphologically complex words are stored in the lexicon and accessed.
- Associative Models
  - Claim that words are stored whole in the lexicon

– Examples: TRACE, MERGE

### • Combinatorial Models

- Claim that morphemes are stored separately and combined during lexical access
- Also known as morphological decomposition models
- Examples: Taft (1988); Taft and Forster (1975)

## 1.3 Previous Research

- Using a Lexical Decision task, and a Cross-modal Priming task, Clahsen et al. (2001) found a difference in processing of German inflected adjectives.

**Table 1** Example from Clahsen et al. (2001)

-m dominant adjectives		-s dominant adjectives	
Stem form	-m -s	Stem form	-m -s
ruhig	838 51 13	rein	783 14 38

## 1.4 Qualitative Predictions

- A highly inflectional language (German) will show a greater effect of morphological complexity than a language with little inflectional morphology (English)
- Other context effects such as lexical frequency and neighborhood density will have a smaller effect on non-native listeners than native listeners, given that their lexicons are not as developed

## 2 Method

### 2.1 Task / Subjects

- Open Response Speech-In-Noise Task
  - Participants respond via keyboard input
- 2 different Signal to Noise Ratios (SNRs) used for each experiment
- signal dependent (but uncorrelated) noise ( see Schroeder, 1968)
- Two separate experiments
  - Experiment 1 — 30 native speakers of English
  - Experiment 2 — 32 native speakers of German

### 2.2 English Materials

- 150 CVCCVC words
  - 74 monomorphemic *basket* /bæskɪt/ *compass* /kəmˈpæs/ *random* /rændəm/
  - 76 bimorphemic *mending* /mɛndɪŋ/ *painted* /peɪntɪd/ *senses* /sɛnsɪz/

- 150 CVCCVC nonwords *nutvit* /nʊtvɪt/ *nisren* /nɪsrɪn/ *tulsid* /tʊlsɪd/
- single male talker

### 2.3 German Materials

- 150 CVCCVC words
  - 75 monomorphemic *dunkel* /dʊŋkəl/ *selten* /zɛltən/ *hektik* /hɛktɪk/
  - 75 bimorphemic *Feindes* /fɛɪndəs/ *bestem* /bɛstəm/ *derber* /dɛrbɛr/
- 150 CVCCVC nonwords *nemschen* /nɛmfən/ *mofkem* /mɔfkəm/ *bomgech* /bɔmgəx/
- single male talker

## 3 Analysis

### 3.1 Confusion

1. Convert spelling to phonemes
2. For each SNR, Block (word or nonword), and position (C1, C2 etc.) make a confusion matrix
3. For each subject, calculate the mean word score ( $p_w$ ) and phoneme score ( $p_p$ )

### 3.2 J-factor

- The j-factor model provides a measure of context effects.
- The j-factor model assumes that phonemes are the basic unit of speech, and that phonemes are perceived independently (which has been shown to hold true most of the time; see Fletcher, 1953; Allen, 1994).
- The probability of correctly identifying a given word (or nonword) can be calculated as the product of the probabilities of its constituent phonemes, as shown in equation 1.

$$p_w = p_{C1}p_{V1}p_{C2}p_{C3}p_{V2}p_{C4} \quad (1)$$

where  $p_w$  is the probability of correctly identifying a word (or nonword). Assuming that phonemes are perceived independently, (1) can be rewritten as:

$$p_w = p_p^j \quad (2)$$

where  $j$  is the number of phonemes, and  $p_p$  is the geometric mean of the probabilities of each constituent phoneme. Rewriting (2), the quantity  $j$  can be empirically determined from confusion matrices by:

$$j = \frac{\log(p_w)}{\log(p_p)} \quad (3)$$

### 3.2.1 Previous J-factor results

- 3 studies have used the j-factor model with CVC English stimuli (Boothroyd and Nittrouer, 1988; Olsen et al., 1997; Benkí, 2003)
- All have found  $j_{nonword} \approx 3$  and  $j_{word} \approx 2.5$
- 1 study using CVC Mandarin stimuli (Benkí et al., in preparation) did not find a difference between words and nonwords

### 3.3 Quantitative Predictions

- Nonwords —  $j = 6$ ; interpretation is that phonemes are being predicted independently of one another
- Words —  $j < 6$ ; interpretation is that lexical status is affecting perception.
- Morphology —  $j_{bi} > j_{mono}$ ; interpretation is that monomorphemes have more context than bimorphemes
- Frequency —  $j_{word} \propto \frac{1}{\text{frequency}}$ ; interpretation is that frequency provides a facilitatory effect
- Neighborhood density —  $j_{word} \propto \text{density}$ ; interpretation is that density provides an inhibitory effect

## 4 Experiment One Results — English listeners

### 4.1 English — Lexical Status

- As expected, there is a significant difference in  $j$  between words and nonwords (see Figure 1, page 5)
- $j$  for nonwords is slightly smaller than expected

### 4.2 English — Morphology

- After removing confounds with lexical frequency and neighborhood density, no significant difference was found between monomorphemes and bimorphemes

### 4.3 English — Lexical Frequency

- Words were grouped into low and high frequency groups via median splits
- As predicted, high frequency words have a lower  $j$ , indicating a facilitatory effect of frequency

### 4.4 English — Neighborhood Density

- Words were also grouped into sparse and dense neighborhoods via median splits
- As predicted, an increase in density causes an inhibitory effect

## 5 Experiment Two Results — German listeners

### 5.1 Item Exclusion

- Initial results for German had much lower than expected  $j$ -scores
- Additional analysis revealed that this was due to stimuli containing post-vocalic /R/ which frequently does not behave as an independent phoneme
- Results for lexical status and morphology shown here have excluded words containing post-vocalic /R/  
94 nonwords and 79 words (36 monomorphemic and 43 bimorphemic)
- Lexical frequency and neighborhood density effects did not seem to be affected by this, so they are shown with the full set of stimuli

### 5.2 German — Lexical Status

- As predicted,  $j_{word}$  is significantly lower than  $j_{nonword}$  (see Figure 2, page 5)
- $j$  for nonwords is slightly smaller than expected

### 5.3 German — Morphology

- As predicted,  $j_{mono}$  was significantly lower than  $j_{bi}$
- This indicates a greater context effect for monomorphemes than bimorphemes

### 5.4 German — Lexical Frequency

- Effects of lexical frequency were also significant
- However, the effect is opposite of that predicted — we find an inhibitory effect

### 5.5 German — Neighborhood Density

- Neighborhood density is also significant
- As predicted, an increase in density causes an inhibitory effect

## 6 Discussion

### 6.1 Summary of Results

**Table 2** J-factor analysis summary

	Lexical Status	Morphology	Log wordform frequency	Log lemma frequency	phonological neighborhood density	phonetic neighborhood density
English	2.07***	0.09	0.51**	0.47**	-0.47**	-0.90***
German	1.45***	0.78***	-0.69***	-0.98***	-0.29*	-1.02***

\*\*\* $p < .001$ , \*\* $p < .01$ , \* $p < .05$

### 6.2 Cross-linguistic effects

- One of the major differences found between the English and German results is the effect of morphology
- The interpretation for this is that German has a much richer inflectional morphology, and therefore morphology plays a larger role in the structure of the lexicon
- Similar cross-linguistic differences have been reported by Marslen-Wilson (2001).
- In comparing Polish, Arabic, English, and Chinese they have obtained different results in terms of how morphology is processed and represented in the lexicon.

Marslen-Wilson (2001) find that:

- In English, complex words such as *darkness* are represented by their constituent morphemes, and are combined during lexical access. English also exhibits stem-priming, e.g. the stem in *darkness* and *darkly* prime *dark*. This is not the case for semantically opaque words such as *department*, which does not prime *depart*.
- Polish also exhibits affix priming, e.g. *kotek/ogródek* ‘a little cat’ / ‘a little garden’ – the diminutive affix in the prime facilitates perception of the target and suffix interference (e.g. *pis-anie/pis-arz* ‘writing’/‘writer’ – no facilitation is found in such pairs, despite facilitation of inflectional endings).
- Morphology seems to play an even larger role in Arabic, which has root priming even for semantically opaque words.
- Chinese has virtually no inflectional or derivational morphology
- Compounding is very active in Mandarin Chinese, and bimorphemic compounds account for up to 70% of all word forms in the language.
- However Marslen-Wilson and colleagues find no evidence for morphological decomposition in Mandarin compounds.
- Vannest et al. (2002) also find similarly various results in a comparison of English and Finnish derivational morphology.
- Research on Finnish inflectional morphology has shown support for combinatorial-like processing (e.g. Laine et al., 1999), Vannest et al.
- But they find less evidence for morphological decomposition with derivational morphology than for English.
- They hypothesize that words with derivational affixes are stored separately in Finnish in order to decrease the amount of morphological processing that the Finnish speaker must perform.

### 6.3 Interaction of Phonetics and Morphology

- It is possible that differences in mono- and bimorphemic stimuli could be partially due to acoustics or response bias.
- The final consonants in the bimorphemic stimuli were restricted to the phonemes /R s m n/, which, along with /ə/ constitute all of the possible inflectional endings for nouns and adjectives in German.
- /m/ and /n/ are known to be highly confusable with one another.
- In addition, /n/ occurs as an inflectional ending much more frequently than /m/.
- In order to investigate this further, a Signal Detection Theory (SDT) analysis was carried out.
- SDT measures the sensitivity of distinguishing two stimuli, using the metric,  $d'$ .
- SDT also provides a measure of bias,  $c$ , which indicates whether one is more or less likely to respond with a particular phoneme.
  - Positive values of  $c$  indicate a bias towards a response;
  - negative values indicate a bias against a response.
- To carry out the SDT analysis, the original confusion matrices for each S/N were transformed into 2x2 submatrices. An SDT analysis was then applied to each submatrix.

**Table 3** Signal Detection Theory analysis of /m/ and /n/ submatrix in final position. For this analysis /m/ is considered to be the target stimulus. Positive  $c'$  indicate a bias towards /n/.

1. in the absence of lexical context effects (non-word condition), /m/ and /n/ are highly confusable, with a small bias towards /n/
2. /m/ and /n/ are perceived as most distinct in the monomorphemic condition,
3. bias towards /n/ is greatest in the bimorphemic case.

	$d'$	$c$
Nonwords		
lower S/N (2 dB)	-0.182	0.555
higher S/N (7 dB)	0.664	0.743
Bimorphemes		
lower S/N (2 dB)	1.616	0.984
higher S/N (7 dB)	1.913	0.556
Monomorphemes		
lower S/N (2 dB)	3.514	0.239
higher S/N (7 dB)	4.733	-0.060

### 6.4 Conclusions

- The  $j$ -factor results for CVCCVC words are mostly consistent with the previous results using CVC stimuli (Boothroyd and Nittrouer, 1988; Olsen et al., 1997; Benkí, 2003)
- One striking new result is that  $j_{word}$  does not scale linearly with word length
- The influence of morphology on spoken word recognition is language dependent

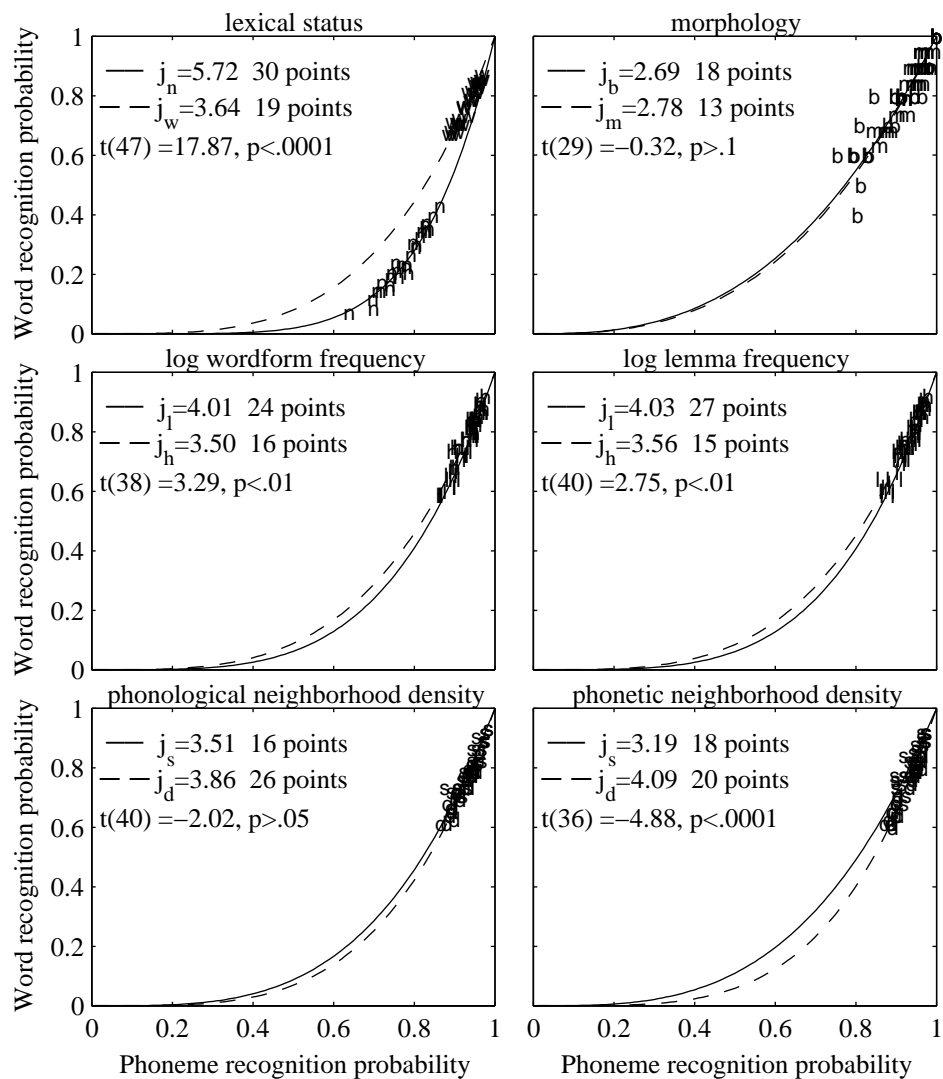
- The processing differences between mono- and bimorphemic found in this study present a challenge to theories of lexical access which assume whole word storage.
- Listeners are particularly sensitive to lexico-statistical information when presented with highly confusable stimuli

### 6.5 Future Research

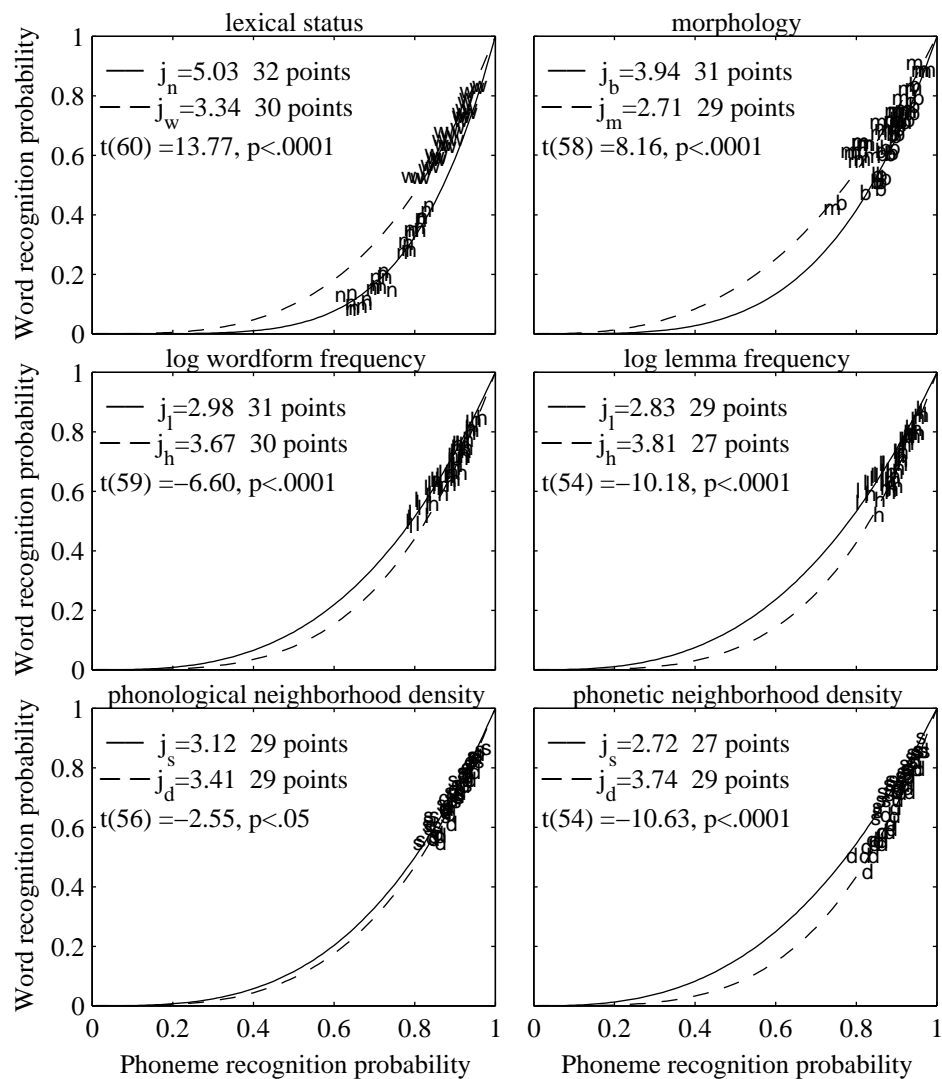
- Further investigate effects of word length on spoken word recognition using stimuli of a variety of lengths
- Determine the time course of these effects using speech-in-noise tasks which also incorporate a measure of time course (either behavioral or neurological)

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**Figure 1:** Experiment 1 (English listeners) J-factor Analysis by subject—Each plot compares two subsets of results from the subject analysis. Curves represent  $y = x^j$ . Frequency and neighborhood density plots show only word results. P-values given are from 2-sample t-tests; before computing the statistics, all points lying in the floor or ceiling ranges ( $> .95$  or  $< .05$ ) were removed, but are still shown on the plot.



**Figure 2:** Experiment 2 (German listeners) J-factor Analysis by subject—Each plot compares two subsets of results from the subject analysis. Curves represent  $y = x^j$ . The frequency and density plots display only word results. P-values given are from 2-sample t-tests; before computing the statistics, all points lying in the floor or ceiling ranges ( $> .95$  or  $< .05$ ) were removed, but are still shown on the plot.

## 7 Sample Confusion Matrices

**Table 4** English — V1 nonwords S/N = -5 dB. Numbers given are percentages. Total number of presentations for each phoneme is given in the final column.

	i	ɪ	eɪ	ɛ	æ	oʊ	ɑ	ə	ɚ	ɔɪ	aʊ	aɪ	null	other	Total
i	<b>79</b>	7	7										7		14
ɪ	5	<b>79</b>									3		3		392
eɪ	9	<b>64</b>	11	9				4					2	2	56
ɛ	8		<b>85</b>	2				2					3		518
æ			27	<b>62</b>			4	1			3		3		350
oʊ	1	1			<b>34</b>	25	23	1	1	3			4	7	182
ɑ				16	2	<b>46</b>	25				2		1	8	252
ə	1		8		12	10	<b>64</b>				1		3	2	196
ɔɪ										<b>64</b>	14	7	14		14
aʊ				20	17	2	2	6			<b>46</b>		5	1	84
aɪ	12		14	12			2				5	<b>52</b>		2	42
															mean $p_p$ = 62
															min $p(oʊ)$ = 34
															max $p(ɛ)$ = 85

**Table 6** English — C4 nonwords S/N = -5 dB. Numbers given are percentages. Total number of presentations for each phoneme is given in the final column.

	d	g	ɕ	t	k	tʃ	f	s	ʃ	v	z	l	m	n	ŋ	nd	rd	null	other	Total
d	<b>93</b>			2												2		4		308
ɕ	1	<b>90</b>																8		84
t	17		<b>77</b>	3			2											2		294
k			4	<b>94</b>														2		308
s							<b>97</b>			2								1		266
ʃ									<b>93</b>									7		14
v	7		4				7	<b>39</b>	25									14	4	28
z	1	2					37		<b>54</b>									6		224
m													<b>73</b>	25				2		238
n	1												6	<b>85</b>	1	5		3		182
ŋ	1	1											3	7	<b>82</b>			5		154
																				mean $p_p$ = 80
																				min $p(v)$ = 39
																				max $p(s)$ = 97

**Table 5** German — V1 nonwords S/N = 2 dB. Numbers given are percentages. Total number of presentations for each phoneme is given in the final column.

	i	ɪ	y	ʏ	u	ʊ	e	ɛ	œ	ɔ	a	ɔɪ	aɪ	ɔl	null	other	Total
ɪ	7	<b>64</b>			5	1	4	7	6	1	3				1	2	560
ʊ	5	1	4	2	<b>73</b>	2	1	1	7						1	3	480
ɛ	4				2	1	<b>79</b>	4	4	2	1				1	3	400
œ	3	5			2	1	43	<b>22</b>	19		3				1	2	176
ɔ							2	1	<b>85</b>	3	6	1			1	3	384
a							1	1	5	<b>87</b>	1	4				1	288
ɔɪ		1					9	9	16	24	<b>33</b>	4	1			4	80
aɪ										28		<b>72</b>					32
																	mean $p_p$ = 64
																	min $p(œ)$ = 22
																	max $p(a)$ = 87

**Table 7** German — C4 nonwords S/N = 2 dB. Numbers given are percentages. Total number of presentations for each phoneme is given in the final column.

	p	t	k	s	x	l	r	m	n	ŋ	null	other	Total
k	3	4	<b>86</b>		5						1		240
s	1			<b>97</b>							1	1	496
x			4		<b>91</b>						1	4	128
l	1					<b>72</b>	6				19		432
r	2					1	<b>78</b>				12	7	624
m	3					3	2	<b>21</b>	60	5	3	3	240
n	1					2	2	27	<b>57</b>	3	3	5	240
													mean $p_p$ = 72
													min $p(m)$ = 21
													max $p(s)$ = 97