

Words as Anchors

Known Words Facilitate Statistical Learning

Toni Cunillera,^{1,2} Estela Càmara,^{3,4} Matti Laine,²
and Antoni Rodríguez-Fornells^{4,5}

¹Department of Basic Psychology, Faculty of Psychology, University of Barcelona, Barcelona, Spain

²Department of Psychology, Åbo Akademi University, Åbo, Finland

³Department of Neuropsychology, Otto-von Guericke University, Magdeburg, Germany

⁴Department of Physiology, Faculty of Medicine, Campus de Bellvitge – IDIBELL, University of Barcelona, Barcelona, Spain

⁵Institució Catalana de Recerca i Estudis Avançats (ICREA), Barcelona, Spain

Abstract. Can even a handful of newly learned words help to find further word candidates in a novel spoken language? This study shows that the statistical segmentation of words from speech stream by adults is facilitated by the presence of known words in the stream. This facilitatory effect is immediate as the known words were acquired only minutes before the onset of the speech stream. Our results demonstrate an interplay between top-down lexical segmentation and bottom-up statistical learning, in line with infant research suggesting that integration of multiple cues facilitates early language learning. The ability to simultaneously benefit from both types of word segmentation cues appears to be present through adulthood and can thus contribute to second language learning.

To learn a new language, listeners must first attain a basic vocabulary. This begins with identification of word candidates in the new language through segmentation of the speech stream. This is not a trivial task as speech represents a continuous signal with no clear pauses indicating word boundaries within a sentence. The difficulty of the segmentation task can also be highlighted by the comparison of spoken and written language: in the latter case, blank spaces clearly mark word boundaries. The acoustic signal, however, contains some reliable cues that can help to segment words in spoken input (e.g., Jusczyk, 1999; Kuhl, 2004).

Among other cues, the distributional properties of speech can be exploited for segmenting the speech stream into words. For example, words can be detected by computing the transitional probabilities of syllables, a process that is part of a more general learning mechanism coined as statistical learning (Saffran, Aslin, & Newport, 1996; Saffran, Newport, & Aslin, 1996). Statistical learning appears to be a domain-general mechanism (but see Conway & Christiansen, 2006) documented in different sensory modalities such as audition, vision, and touch (Conway & Christiansen, 2006; Fiser & Aslin, 2002; Kirkham, Slemmer, & Johnson, 2002). It has been utilized in a diverse set of learning paradigms in which statistical regularities can be exploited, such as in learning an artificial syntax (Gomez & Gerken, 1999) and segmenting out a tone or a speech sequence (Abla, Katahira, & Okanoya, 2008; Saffran, Aslin, et al., 1996; Saffran, Johnson, Aslin, & Newport, 1999; Saffran, Newport, et al., 1996). Concerning word segmenta-

tion, the statistical learning view posits that words can be detected by computing the likelihood of appearance of syllables along the speech sequence (Saffran, Aslin, et al., 1996; Saffran, Newport, et al., 1996). Here high transitional probabilities would indicate that the presence of the syllable X strongly predicts the occurrence of the syllable Y , and it is most likely to happen within a word than between words (i.e., XY could be a word or a part of it) because the pairs of contiguous syllables within a word are constrained by the lexicon and the phonotactics of the language. In contrast, low transitional probabilities (signaling a weak contingency between X and Y) are indicative of a word boundary.

Intuitively, one could assume that an interaction between known words and words to be learned may facilitate speech segmentation through statistical learning. In other words, one would expect that when the very first words are learned, further word segmentation could be facilitated by these words. It is a common experience that when traveling abroad, people may know a few words of the language of the foreign country. One may recognize such words when listening to a conversation between native speakers, although most of the speech message may sound as an interrupted stream of nonsense speech. But even in that situation new words may be detected when being adjacent to a known word that provides a boundary for segmentation. Thus far, the statistical learning approach (Aslin, Saffran, & Newport, 1998; Saffran, Aslin, et al., 1996; Saffran, Newport, et al., 1996) has not considered as to how recently segmented words (based on the output of statistical learning)

48
49
50
51
52
53
54
55
56
57
58
59
60
61
62
63
64
65
66
67
68
69
70
71
72
73
74
75
76

77	could aid to isolate the remaining words from the speech	Welsh, 1978; McClelland & Elman, 1986; Norris, 1994).	137
78	stream.	Statistical learning (based on word frequency and distribu-	138
79	The idea that isolated or familiar words can play a signif-	tical regularities) is relegated to a secondary tier in the	139
80	icant role in speech segmentation is not new (see, e.g.,	model. The model predicts that when known words are first	140
81	Peters, 1983; Pinker, 1994), although this hypothesis has	recognized in an utterance, the subsequent contiguous string	141
82	been evaluated only in infant language learning. For	of syllables is immediately inferred as a new word. It seems,	142
83	example, Brent and Siskind (2001) explored the role of iso-	therefore, that the inference of a new word is most likely to	143
84	lated words in the formation of infants' vocabulary. These	occur when it appears between a familiar word and a phrasal	144
85	authors demonstrated that isolated words are abundant in	pause (see Dahan & Brent, 1999). Hence, the model predicts	145
86	infant-directed speech. They found that about 30–50 words	less accurate segmenting in utterances where the familiar	146
87	used by infants in their study (~ 44% of the production	word is found in the middle of the utterance.	147
88	from infants recorded at the age between 9 and 15 months)	Statistical learning (Saffran, Aslin, et al., 1996) predicts	148
89	were words spoken to the infants in isolation by their moth-	that word boundaries are expected when low transitional	149
90	ers before infants used them. The authors suggested that ini-	probabilities are detected in the speech sequence, irrespec-	150
91	tial words that compose the small vocabulary of infants may	tive whether they appear in the middle or at the edge of a	151
92	provide reliable cues that help infants with their vocabulary	sentence. However, as we mentioned above, it does not take	152
93	expansion observed during the second year of life (e.g.,	into account the role of recently isolated words during the	153
94	Bates, Bretherton, & Snyder, 1988; Fenson et al., 1994).	online segmentation process (as the INCDROP model does)	154
95	Thus, new words may be segmented from multiwords utter-	ances by recognizing adjacent known words. This strategy	155
96	may be adopted by children with the first words that they	or the role of acoustical cues as word-stress or allophonic	156
97	master, and could happen early in development as Mandel,	variations. This apparently relates to the scope of the origi-	157
98	Jusczyk, and Pisoni (1995) demonstrated that infants recog-	nal statistical learning studies (to test the role of a single	158
99	nize their own names already at the age of 4.5 months.	specific cue in segmentation) rather than to a theoretical	159
100	In another study, Bortfeld, Morgan, Golinkoff, and	stance. For example, it is interesting to note that Dahan and	160
101	Rathbun (2005) exposed 6-month-old infants to a series of	Brent (1999, p. 183) suggested that “transitional probabili-	161
102	short utterances in which a familiar word (infant's own name	ties might work together with lexically driven segmentation”.	162
103	or mother's name) or an unfamiliar one was followed by a	Although this hypothesis is plausible, no empirical evidence	163
104	new word (an object name unknown for the child). The	has been provided as to how these two processes, segmen-	164
105	results of this experiment proved that infants segmented	tation based on known words (anchors) and statistical learn-	165
106	new words from fluent speech only when new words were	ing (transitional probabilities), might interact with each	166
107	followed by a familiar name. This study demonstrated that	other.	167
108	the first words infants recognize become useful segmen-	In this study, we examined whether familiar words can	168
109	tation cues, probably acting as anchor points that indicate	act as anchor words that aid adults to segment unknown	169
110	which sound sequence next in the sentence is a wordlike	words in a new language. In other words, we studied	170
111	unit.	whether the presence of familiar words in a speech segmen-	171
112	Segmenting fluent speech into words is a very different	tation task facilitates statistical learning of novel words.	172
113	task in an already mastered language and in a new language	Participants were exposed to a continuous speech stream	173
114	to be learned. For a language already mastered a variety of	of an artificial language that could be parsed into wordlike	174
115	lexical cues can be used, whereas for a new language only	units only through statistical learning, that is, by computing	175
116	acoustical, perceptual, and distributional cues are available	transitional probabilities between syllables. Prior to the	176
117	in the speech signal. While empirical evidence for an inter-	presentation of the speech stream, participants learned two	177
118	play between statistical learning and lexical knowledge in	novel words (anchor words) which did or did not belong	178
119	speech segmentation is lacking, there is at least one compu-	to the subsequent speech stream. We hypothesized that these	179
120	tational model directly relevant to this issue, namely the	recently acquired words, when recognized in the language	180
121	INCDROP (incremental distributional regularity optimiza-	stream, would improve speech segmentation.	
122	tion; Brent, 1997; Brent & Cartwright, 1996; Dahan &		
123	Brent, 1999). Its main assumption is that speech segmen-		
124	tation might rely on the experience acquired with the words	Experiment 1	181
125	of a language. Another important feature of the model is that		
126	it brings together two properties of language at the earliest	Method	182
127	stages of language acquisition and lexicon formation,		
128	namely the distributional regularities and the knowledge of	Participants	183
129	familiar words. As posited by Brent (1997), it was proposed		
130	as a top-down lexically driven segmentation model that is	Fifty-six (mean age 20.8 ± 2.23 SD) undergraduate	184
131	able to discover new words. It is worth noting that the INC-	psychology students at the University of Barcelona received	185
132	DROP model differs from other proposals in asserting that	extra course credits for their participation in the experiment.	186
133	experience with words of a language is the main determinant	Participants were randomly assigned to one of the two	187
134	of segmentation at the earliest stages of language acquisition	experimental conditions (anchor word condition or nonan-	188
135	(e.g., Cole, Jakimik, & Cooper, 1980; Marslen-Wilson &	chor word condition, see below)	189
136			

190	Stimuli		
191	The Artificial Language Stream		
192	Forty-eight different consonant-vowel syllables were combined to create two language streams which followed the same structure as those created by Saffran, Aslin, et al. (1996); Saffran, Newport, et al. (1996). We decided to use two different language streams to control for possible arbitrary listening preferences. For each stream, eight trisyllabic nonsense words were concatenated to form a nonstop speech stream by using the text-to-speech synthesizer MBROLA with a Spanish male diphone database at 16 kHz (Dutoit, Pagel, Pierret, Bataille, & van der Vreken, 1996). Importantly, words were combined in a way that each word in the stream was followed by each of the other words the same number of times.		242
205	The use of this artificial language learning methodology enables us to rule out such potential segmentation cues as word-stress or coarticulation. Thus, all phonemes had the same duration (116 ms) and pitch (200 Hz; equal pitch rise and fall, with pitch maximum at 50% of the phoneme) in the language streams. The only reliable cue for word boundaries was the statistical structure of the language. In all streams the transitional probability of the syllables forming a word was 1.0, while for syllables spanning word boundaries it was 0.14. Each word was repeated 28 times along the stream with the constraint that the same word never occurred twice in a row. The duration of each word was 696 ms, yielding a total stream duration of 2 min 35 s and 904 ms. A written excerpt from the speech stream is as follows: “demuri/senige/somapo/kotusa/tokuda/piruta/furake/bagoli/senige/tokuda/demuri. . .”. Here the three-syllable wordlike units are separated by slashes. In the anchor word condition, the two words that were taught prior to stream exposure were included in the speech stream.		243
224	In addition, eight nonwords were created for each language by recombining the syllables of the eight words composing the stream. Nonwords were sequences of three syllables that never formed a string in the language stream (transitional probability = 0). Finally, for the two languages 112 part-words were created by recombining the syllables of the eight words from each language. Fifty-six part-words were made by concatenating the last two syllables of a word and the first one of another (part-words 2-3-1), and the other 56 were made by concatenating the last syllable of a word and the first two syllables of another (part-words 3-1-2).		244
235	Word Learning Phase		245
236	The participants were taught two words of the new language by showing pictures together with an auditorily presented narration in Spanish. The presentation lasted for ~ 3 min. The synopsis was as follows: “A space traveler stops by an unknown planet looking for water and food. After he lands he decides to move to a nearby city. There he meets		246
	a local inhabitant who speaks an unknown strange language. The alien provides the traveler with water and apples and at the same time teaches him the words in his language that refer to water and apple”. Each time the two new “alien” words were presented, the female narrator’s voice was replaced by the synthesized speech used in the subsequent artificial language stream.		247
	The two novel words were repeated three times during the presentation. Each word was associated either to water or apple. We decided to provide the novel word with an associated meaning to simulate a more natural learning process: the first words learned in a foreign language are usually concrete, familiar, and frequent objects.		248
	Procedure		249
	Twenty-eight participants were randomly assigned to the anchor word condition and the other 28 to the nonanchor word condition. In the word learning phase, the participants were instructed to pay attention to the slide show and to learn the new words that would be presented. Immediately after the slide show, they heard the new words separately and were asked to write down the corresponding meaning (Spanish translation equivalent, i.e., <i>agua</i> and <i>manzana</i>). Each participant saw the same slide show but with different words so that the word presentation became counterbalanced across participants. The segmentation task began not until the participant had identified the meaning of the new words. They were allowed to write the response up to three times, and when the fourth erroneous response was recorded, the slide show was replayed ¹ . In the anchor word condition, the participants learned two of the eight words composing the novel language. In contrast, in the nonanchor word condition the participants learned two trisyllabic sequences that were not presented in the language stream.		250
	Immediately after successful completion of the word learning phase, the participants were requested to listen carefully to the language stream and to discover the words of the novel language. They were informed that a final test would be presented at the end of the language stream. Importantly, they were not informed about the presence of the two recently learned words in the language stream. For each condition, the participants were randomly assigned to one of the two language streams (Language A or B). The two language streams were counterbalanced across participants in correspondence with the preceding slide show.		251
	Immediately after the language stream, a standard auditory two-alternative-forced-choice (2AFC) test was presented. Test items comprised the eight words of each stream (for the anchor word condition, the two previously learned words were included in the set of eight words) and eight part-words randomly selected from the pool of 112 part-words of the same stream (four part-words corresponding to the syllable structure 2-3-1 and four to the syllable structure 3-1-2; see the Stimuli section). Words and		252

¹ We had to repeat the presentation of the slide show only in a very small number of cases.

part-words were combined so that each word was paired with four different part-words but each of the eight part-words appeared equally often. This procedure rendered a total of 32 pairs that were presented in random order. After hearing each pair of test items, the participants were asked to decide by pressing a button whether the first or the second item of the pair was a word of the new language. Presentation of the items of a pair was separated by a 400 ms pause.

Results and Discussion

We began comparing the segmentation rates between the participants who listened to the different language streams (Language A and B) for the two experimental conditions. The results revealed no significant stream differences in either the anchor word condition or the nonanchor word condition (in both cases $t(26) < 1$). Consequently, in all subsequent analyses the data were collapsed across the two languages.

For the anchor word condition, the mean percentage of correctly segmented six novel words (excluding the two anchor words that were taught prior to the segmentation task) was $71.4 \pm 14.8\%$ (percentage different from chance level (50%), $t(27) = 7.7, p < .001$). For the nonanchor word condition, the mean percentage of correctly segmented eight novel words was $63.6 \pm 9.8\%$. This percentage was also different from chance (50%), $t(27) = 7.3, p < .001$. The segmentation performance was significantly better for the anchor word condition than for the nonanchor word condition ($t(54) = 2.33, p = .02, d = .63$). When including the anchor words in the analysis, the difference in segmentation performance between the anchor word condition and the nonanchor word conditions remained essentially the same ($t(54) = 2.49, p = .01, \text{effect size: } d = .66$).

Even though the brief training phase ensured that all participants in the anchor word condition had learned to associate the isolated anchor words to their corresponding meanings, it was of interest to explore as to what extent the anchor words were explicitly segmented in the speech stream. For each participant, the criterion of at least three out of four correct responses in the test phase for each of the two anchor words was employed. Almost half of the participants (13/28) failed to reach this criterion. In order to examine whether or not their lack of explicit segmentation of the anchor words affected the overall segmentation performance of the study group, data from these 13 participants were substituted by data collected from new participants who fulfilled this criterion. The pattern of results remained very similar to the one reported above for the original anchor condition group: the mean percentage of correctly segmented words in the anchor condition was $71.5 \pm 14.8\%$, a percentage that was significantly different from chance (50%), $p < .001$. A comparison of the segmentation results across conditions revealed significant differences (anchor words vs. nonanchor words: $t(54) = 2.33, p < .03, d = .62$).

The present results corroborate the hypothesis that the presence of anchor words facilitates the segmentation of novel words in a language stream. They also indicate that

an explicit segmentation of the anchor words themselves is not necessary for this facilitation to occur. However, there are two potential confounds that need to be ruled out. First, it might be that the difference in segmentation performance between the anchor versus the nonanchor condition is due to interference created by the nonanchor words rather than facilitation by the anchor words in their corresponding conditions. Second, the overall segmentation load may have favored the anchor word condition. The participants in the nonanchor word condition had to segment eight totally novel words, whereas in the anchor word condition, two of the eight novel words had been shown in the training phase. In order to clarify these issues, we conducted two new experiments.

Experiment 2

A possible explanation for the significantly lower segmentation rate for the nonanchor word condition in Experiment 1 would be that recently learned words caused participants to use a detrimental mis-segmentation strategy, as the syllables that composed of these words were also present in the subsequent language stream. In order to rule out this alternative, we ran an experiment where the learned words were composed of syllables that were not present in the subsequent speech segmentation task.

Method

Participants

Twenty-eight (mean age $20.1 \pm 1.42 SD$) undergraduate psychology students at the University of Barcelona participated for extra course credits. None of them took part in Experiment 1. Participants were randomly assigned to one of the two language streams (Language A or B).

Stimuli and Procedure

The language streams, words, part-words, the slide show, and the whole procedure were the same as in Experiment 1, with the exception of different words being taught in the slide show. For the present experiment, words from Language A were used in the slide show for the Language B and vice versa. Thus, in contrast with Experiment 1, the learned words consisted of syllables that were not present in the subsequent language stream.

Results and Discussion

No significant differences in segmentation performance were encountered between languages ($t(26) < 1$) and therefore the data were collapsed across the two languages for all subsequent analyses. The mean percentage of correctly

segmented words was $63.8 \pm 13.9\%$, being significantly different from chance (50%), $t(27) = 5.3, p < .001$. A comparison of the segmentation results of Experiment 2 and the nonanchor word condition of Experiment 1 showed no differences ($t(54) = 0.7, p > .9, d = .02$). This indicates that it was irrelevant for the speech segmentation performance whether or not “nonanchors” consisted of syllables that were present in the language stream.

Moreover, when comparing the nonanchor word condition in this experiment with the anchor word condition in Experiment 1, we observed better segmentation for the anchor word condition ($t(54) = 1.98, p = .05, d = .53$).

Experiment 3

We ran another experiment in order to compare the segmentation rate in the anchor word condition with a nonanchor condition of a language composed of only six words. An intrinsic property of the anchor word condition in Experiment 1 was that although the streams consisted of eight words, only six of them were totally novel for the participants. Consequently, it could be argued that the significantly lower segmentation performances observed for the nonanchor word conditions in Experiments 1 and 2 were due to participants facing a more demanding task (segmenting eight words) in comparison with the anchor word condition (segmenting six words and recognizing the other two words either explicitly or implicitly).

In order to equate the number of words that needed to be segmented, in the present experiment we reduced the words composing the nonanchor word condition from eight to six. If task difficulty was responsible for the differences reported in the previous experiments, we should observe a better segmentation rate in this new nonanchor word condition than in the previous nonanchor word conditions.

Method

Participants

Another 28 (mean age 20.8 ± 2.29 *SD*) undergraduate psychology students at the University of Barcelona who did not take part in the previous experiments were recruited for the present experiment and received extra course credits for their participation. They were randomly assigned to one of the two language streams (Language A or B).

Stimuli

Two new languages were created by recombining six of the eight words from the previously used languages. Consequently, the stream duration was reduced to 1 min 56 s and 928 ms. The structure of the languages was the same as in the previous experiments (see the Stimuli section of Experiment 1). In the two streams the transitional

probability of the syllables forming a word was 1.0, while for syllables spanning word boundaries it was 0.2. The number of part-words was reduced to 64 in this experiment. In addition, six new nonwords for each language were created by recombining the syllables of the six words composing the language, yielding six syllable sequences with transitional probability equal to zero in the language stream. The two to-be-learned words used in the first phase of the experiment were taken from these nonwords. The slide show and the overall setup were the same as in Experiments 1 and 2.

Procedure

The procedure was the same as in Experiments 1 and 2. The same 2AFC speech segmentation test was administered to the participants as in Experiments 1 and 2 but the number of item pairs was 36 for the present experiment. The six words composing the stream were exhaustively combined with six part-words (three part-words corresponding to the syllable structure 2-3-1 and three to the syllable structure 3-1-2) rendering 36 pair items.

Results and Discussion

No differences were observed between the two languages ($t(26) < 1$) and thus the data were collapsed. The mean percentage of correctly segmented words was $63.1 \pm 13.8\%$ (see Figure 1), being different from chance level (50%), $t(27) = 5.0, p < .001$. When comparing the segmentation performance between Experiments 2 and 3, no difference was found ($t(54) < .3$). This indicates that anchor word facilitation effect observed in Experiment 1 was not due to a difference in segmentation load between the anchor versus nonanchor condition. When comparing the nonanchor word conditions between Experiments 1 and 3, no significant difference was observed either ($t(54) < .2$).

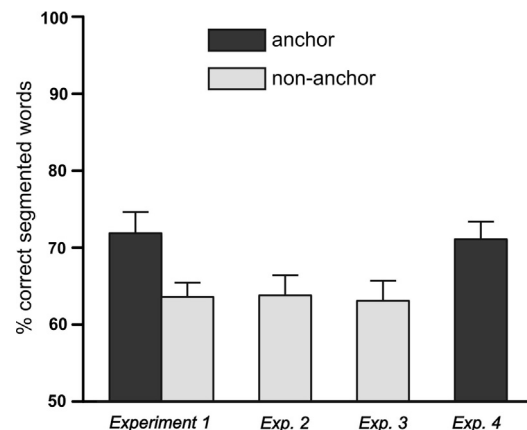


Figure 1. Mean percentage (\pm SE) of correctly segmented words in the auditory 2AFC test performed at the end of Experiments 1–4.

478	We then compared the present results with those of	(Exp. 1 nonanchor condition vs. Exp. 4: $t(54) = 2.14$,	521
479	Experiment 1 and again observed a larger rate of segmented	$p < .04$, $d = .57$; Exp. 3 vs. Exp. 4: $t(54) = 1.98$, $p =$	522
480	words for the anchor word condition (Exp. 3 vs. Exp. 1	$.053$, $d = .53$).	523
481	anchor word condition: $t(54) = 2.19$, $p < .04$, $d = .58$).	As in Experiment 1, we also explored how well the	524
482	In summary, the present findings help to rule out the possi-	anchor words were explicitly segmented in the speech	525
483	bility that the lower segmentation performance in the non-	stream. Given the added anchor word training in the present	526
484	anchor condition in Experiment 1 was due to a higher	experiment, the number of participants failing to explicitly	527
485	overall segmentation load as compared to the anchor word	segment the anchor words was expected to be lower. Indeed,	528
486	condition.	only 5 of the 28 participants failed to fulfill the criterion of at	529
		least three out of four correct responses in the test phase for	530
		each of the two anchor words. When their results were	531
487	Experiment 4	replaced by data from five new participants who fulfilled	532
		this criterion, the results remained again essentially the	533
		same mean percentage of correctly segmented words:	534
488	Finally, we wanted to ensure that the facilitatory effect of	$71.1 \pm 12.1\%$, t tests against chance (50%) $p < .001$; Exp.	535
489	anchor words on speech segmentation observed in Experim-	1 (nonanchor condition) versus Exp. 4: $t(54) = 2.55$,	536
490	ent 1 is reliable enough to be replicated. The present	$p < .02$, $d = .68$; Exp. 3 versus Exp. 4: $t(54) = 2.32$,	537
491	experiment thus attempted to replicate the anchor word con-	$p < .03$, $d = .62$).	538
492	dition in Experiment 1 with an identical setup except for	In sum, the present results replicate those obtained for	539
493	some more training for the anchor words in the learning	the anchor word condition in Experiment 1. This gives fur-	540
494	phase.	ther support to the hypothesis that recently learned words	541
		can facilitate speech segmentation.	542
495	Method		
496	Participants	General Discussion	543
497	An additional set of 28 (mean age 20.3 ± 2.48 <i>SD</i>) under-	In this study we explored how recently learned words affect	544
498	graduate psychology students at the University of Barcelona	statistical learning in a speech segmentation task. The results	545
499	took part in the experiment and received extra course credits	from Experiments 1 and 4 demonstrated that speech seg-	546
500	for their participation. They were randomly assigned to one	mentation performance was increased when recently learned	547
501	of the two language streams (Language A or B).	words were embedded in the language stream. Experiments	548
		2 and 3 showed that the observed advantage was not due to	549
		interference caused by miscuing in the control condition or	550
502	Stimuli	due to the different number of words to be segmented.	551
503	All the stimuli (language streams, words, part-words, and	The present findings suggest that the very first learned	552
504	the slide show) were the same as in Experiment 1.	words help to isolate and discover novel words of a new lan-	553
		guage. Thus the first learned words appear to aid the under-	554
		lying statistical learning process when segmenting new	555
505	Procedure	words. Additional analyses performed in Experiment 1 indi-	556
		cated that this facilitatory effect is presented irrespective	557
		whether all subjects explicitly segmented the anchor words	558
506	The whole procedure was the same as in Experiment 1, with	or not. In other words, even for subjects who did not any-	559
507	the exception of a modified training setup used for the <i>word</i>	more consciously recognize the anchor words, these words	560
508	<i>learning phase</i> . For the present experiment, we doubled the	still appeared to boost segmentation performance.	561
509	number of repetitions of the "to-be-learned words" in the	Our results indicate that lexically driven segmentation, as	562
510	learning phase. Each anchor word appeared thus six times	proposed by the INCDROP model (Dahan & Brent, 1999),	563
511	in the slide show (the last three times simultaneously in spo-	can work in concert with computation of transitional proba-	564
512	ken and written form), as compared with the three auditory	bilities between syllables. There is, however, an alternative	565
513	exposures used in Experiment 1.	explanation on how speech segmentation is achieved by dis-	566
		tributional cues in the speech input, and it has been success-	567
		fully implemented in a computer model called PARSER	568
514	Results and Discussion	(Perruchet & Vinter, 1998). Rather than computing transi-	569
		tional probabilities, PARSER is based on the formation of	570
515	With this new training setup, the results were similar to	chunks (Anderson & Lebiere, 1998) positing it as the core	571
516	those reported in Experiment 1 ($70.1 \pm 12.6\%$, a percentage	principle for statistical learning in speech segmentation	572
517	significantly above chance levels (50%), $p < .001$). Import-	(see Perruchet & Pacton, 2006 for an interesting discussion	573
518	antly, when comparing the performance in this new exper-	of this controversy). The model makes the strong claim that	574
519	iment with the nonanchor conditions in Experiments 1 and	for segmenting speech there is no need for computations. In-	575
520	3, we observed the same advantage for the anchor condition	stead, chunks are formed and shaped over time following	576

the laws governing associative memory and the attentional capacity constraints that limit the processing of incoming information. In the same vein, Pacton and Perruchet (2008) have recently proposed a general associative learning model that asserts attention as the necessary and sufficient condition for associative learning and language chunking to occur. However, the interplay of different segmentation cues has not yet been implemented in the PARSER.

The present results thus reflect an interplay between a top-down process (lexical segmentation) and a bottom-up process (computation of transitional probabilities). Bortfeld et al. (2005) suggested a similar process to explain how 6-month-old infants succeeded in segmenting out new words from utterances after recognizing familiar words in them. However, an important difference with our study is that the familiar names used by Bortfeld et al. (“mommy/mama” or the infants’ name) were probably well consolidated in their infants’ memory, as they heard these words every day. Our participants were able to benefit from the learned words although their experience with these words was minimal and even when not all of them could explicitly recognize the anchor words anymore at the final segmentation task. This demonstrates that lexical items can contribute to speech segmentation immediately after they are learned.

The present results show that adult listeners can combine statistical learning with other segmentation cues available in speech. Infant research has suggested that integration of multimodal cues facilitates language learning (Bahrick & Lickliter, 2000; Hollich, Newman, & Jusczyk, 2005; Hollich et al., 2000). A recent speech segmentation study also found a positive effect of combining intrasensory statistical regularities in speech and music (Schön et al., 2008). Therefore, it is plausible that the coalition of multiple cues, as far as they do not collide (see, e.g., Johnson & Jusczyk, 2001; Thiessen & Saffran, 2003), can facilitate speech segmentation. The cue-specific weights in a multi-cue context during second language acquisition are not yet clear (but see Christiansen, Allen, & Seidenberg, 1998).

Another critical issue concerns the use of top-down lexical segmentation and bottom-up computation of transitional probabilities at different ages. Our data suggest that both of these mechanisms remain active after childhood (see Braine et al., 1990; Gillette, Gleitman, Gleitman, & Lederer, 1999). In line with this, statistical learning has been demonstrated in both infants and adults when learning an artificial mini-language (Saffran, Aslin, et al., 1996; Saffran, Newport, et al., 1996). Likewise, it appears that infants benefit from isolated and familiar words at the initial stages of language comprehension (Bortfeld et al., 2005; Mandel et al., 1995) and at the beginning of their vocabulary expansion (Brent & Siskind, 2001), and such an effect is present also in adults with their initial contact with a new language (Dahan & Brent, 1999).

Further evidence for similarities of adults’ and infants’ language learning systems comes from a word learning experiment where adults were exposed to infant-directed speech (Golinkoff & Alioto, 1995). English-speaking adults were exposed to short sentences spoken in Chinese while watching pictures corresponding to target object names embedded in the sentences. One group heard sentences

pronounced in infant-directed speech, whereas the other group heard sentences pronounced in adult-directed speech. Only those exposed to infant-directed speech could segment the target words. It is important to note that some of the properties of the infant-direct speech are found in a variety of languages like English, Italian, French, German, Japanese, and Chinese (Fernald et al., 1989). However, it is impossible to say as to which cue or cues contributed most to speech segmentation, as infant-directed speech has many characteristic features (slower speech rate, extended frequency range, higher fundamental frequency, repeated pitch contours, marked intensity shifts, longer pauses, simplified vocabulary, and vowel lengthening; Hoff-Ginsberg & Shatz, 1982). Interestingly, some properties of infant-directed speech are observable in “foreigner talk”, that is, in native speakers interacting with nonnatives (Snow, Vaneeden, & Muysken, 1981).

It thus seems plausible that when infants and adults are exposed to a new language, they both rely on the same top-down and bottom-up strategies to isolate new words. In fact Bortfeld et al. (2005) argued that there is no reason to believe that infants cannot use top-down lexical strategies for segmenting speech. While both strategies appear to be in use throughout the life span, further studies are needed to clarify the relative weight of these strategies in children versus adults.

In summary, we show that very recently acquired words facilitate word segmentation in a new language when the learned words appear in the speech stream. This indicates a possible interplay between lexical top-down processing and bottom-up segmentation based on transitional probabilities of syllables. More generally, our results highlight the employment of multiple cues in vocabulary acquisition.

Acknowledgments

This project was supported by the Academy of Finland NEURO research program grant to ML, the Spanish Government grant to ARF (SEJ2005-06067/PSIC), and the Generalitat de Catalunya predoctoral stage grant to TC. We thank Irene Nogué for helping in data collection and Salvador Soto-Faraco for his comments on the earlier drafts of this paper.

References

- Abla, D., Katahira, K., & Okanoya, K. (2008). On-line assessment of statistical learning by event-related potentials. *Journal of Cognitive Neuroscience*, 20, 952–964.
- Anderson, J. R., & Lebiere, C. (1998). *The atomic components of thought*. Mahwah, NJ: Lawrence Erlbaum Associates.
- Aslin, R. N., Saffran, J. R., & Newport, E. L. (1998). Computation of conditional probability statistics by 8-month-old infants. *Psychological Science*, 9, 321–324.
- Bahrick, L. E., & Lickliter, R. (2000). Intersensory redundancy guides attentional selectivity and perceptual learning in infancy. *Developmental Psychology*, 36, 190–201.
- Bates, E., Bretherton, I., & Snyder, L. S. (1988). *From first words to grammar*. New York: Cambridge University Press.

- 692 Bortfeld, H., Morgan, J. L., Golinkoff, R. M., & Rathbun, K.
693 (2005). Mommy and me – Familiar names help launch babies
694 into speech-stream segmentation. *Psychological Science*, 16,
695 298–304.
- 696 Braine, M. D. S., Brody, R. E., Brooks, P. J., Sudhalter, V., Ross, J.
697 A., Catalano, L., et al. (1990). Exploring language-acquisition
698 in children with a miniature artificial language – Effects of item
699 and pattern frequency, arbitrary subclasses, and correction.
700 *Journal of Memory and Language*, 29, 591–610.
- 701 Brent, M. R. (1997). Toward a unified model of lexical
702 acquisition and lexical access. *Journal of Psycholinguistic*
703 *Research*, 26, 363–375.
- 704 Brent, M. R., & Cartwright, T. A. (1996). Distributional
705 regularity and phonotactic constraints are useful for segmen-
706 tation. *Cognition*, 61, 93–125.
- 707 Brent, M. R., & Siskind, J. M. (2001). The role of exposure to
708 isolated words in early vocabulary development. *Cognition*,
709 81, B33–B44.
- 710 Christiansen, M. H., Allen, J., & Seidenberg, M. S. (1998). Learning
711 to segment speech using multiple cues: A connectionist model.
712 *Language and Cognitive Processes*, 13, 221–268.
- 713 Cole, R. A., Jakimik, J., & Cooper, W. E. (1980). Segmenting
714 speech into words. *Journal of the Acoustical Society of*
715 *America*, 67, 1323–1332.
- 716 Conway, C. M., & Christiansen, M. H. (2006). Statistical
717 learning within and between modalities: Pitting abstract
718 against stimulus-specific representations. *Psychological Sci-*
719 *ence*, 17, 905–912.
- 720 Dahan, D., & Brent, M. R. (1999). On the discovery of novel
721 wordlike units from utterances: An artificial-language study
722 with implications for native-language acquisition. *Journal of*
723 *Experimental Psychology-General*, 128, 165–185.
- 724 Dutoit, T., Pagel, N., Pierret, F., Bataille, O., & van der Vreken,
725 O. (1996). The MBROLA project: Towards a set of high-
726 quality speech synthesizers free of use for non-commercial
727 purposes. In *Conference Proceeding ICSLP'96, Philadelphia*
728 (pp. 1393–1396).
- 729 Fenson, L., et al. (1994). Variability in early communicative
730 development. In *Monographs of the Society of Research in*
731 *Child Development*, 59 (5, Serial No. 242).
- 732 Fernald, A., Taeschner, T., Dunn, J., Papousek, M., de Boysson-
733 Bardies, B., & Fukui, I. A. (1989). Cross-language study of
734 prosodic modifications in mothers' and fathers' speech to
735 preverbal infants. *Journal of Child Language*, 16, 477–501.
- 736 Fiser, J., & Aslin, R. N. (2002). Statistical learning of higher-
737 order temporal structure from visual shape sequences.
738 *Journal of Experimental Psychology: Learning, Memory,*
739 *and Cognition*, 28, 458–467.
- 740 Gillette, J., Gleitman, H., Gleitman, L., & Lederer, A. (1999).
741 Human simulations of vocabulary learning. *Cognition*, 73,
742 135–176.
- 743 Golinkoff, R. M., & Alioto, A. (1995). Infant-directed speech
744 facilitates lexical learning in adults hearing Chinese: Impli-
745 cations for language acquisition. *Journal of Child Language*,
746 22, 703–726.
- 747 Gomez, R. L., & Gerken, L. (1999). Artificial grammar learning
748 by 1-year-olds leads to specific and abstract knowledge.
749 *Cognition*, 70, 109–135.
- 750 Hoff-Ginsberg, E., & Shatz, M. (1982). Linguistic input and the
751 child's acquisition of language. *Psychological Bulletin*, 92,
752 3–26.
- 753 Hollich, G. J., Hirsh-Pasek, K., Golinkoff, R. M., Brand, R. J.,
754 Brown, E., Chung, H. L., et al. (2000). Breaking the
755 language barrier: An emergentist coalition model for the
756 origins of word learning. *Monographs of the Society for*
757 *Research in Child Development*, 65, i-123.
- 758 Hollich, G. J., Newman, R. S., & Jusczyk, P. W. (2005). Infants'
759 use of synchronized visual information to separate streams of
760 speech. *Child Development*, 76, 598–613.
- Johnson, E. K., & Jusczyk, P. W. (2001). Word segmentation by
761 8-month-olds: When speech cues count more than statistics.
762 *Journal of Memory and Language*, 44, 548–567.
- Jusczyk, P. W. (1999). How infants begin to extract words from
764 speech. *Trends in Cognitive Sciences*, 3, 323–328.
- Kirkham, N. Z., Slemmer, J. A., & Johnson, S. P. (2002). Visual
766 statistical learning in infancy: Evidence for a domain general
767 learning mechanism. *Cognition*, 83, B35–B42.
- Kuhl, P. K. (2004). Early language acquisition: Cracking the
769 speech code. *Nature Reviews Neuroscience*, 5, 831–843.
- Mandel, D. R., Jusczyk, P. W., & Pisoni, D. B. (1995). Infants
771 recognition of the sound patterns of their own names.
772 *Psychological Science*, 6, 314–317.
- Marslen-Wilson, W. D., & Welsh, A. (1978). Processing
774 interactions and lexical access during word recognition in
775 continuous speech. *Cognitive Psychology*, 10, 29–63.
- McClelland, J. L., & Elman, J. L. (1986). The trace model of
777 speech-perception. *Cognitive Psychology*, 18, 1–86.
- Norris, D. (1994). Shortlist – A connectionist model of
779 continuous speech recognition. *Cognition*, 52, 189–234.
- Pacton, S., & Perruchet, P. (2008). An attention-based associative
781 account of adjacent and nonadjacent dependency learning.
782 *Journal of Experimental Psychology: Learning, Memory, and*
783 *Cognition*, 34, 80–96.
- Perruchet, P., & Pacton, S. (2006). Implicit learning and
785 statistical learning: one phenomenon, two approaches. *Trends*
786 *in Cognitive Sciences*, 10, 233–238.
- Perruchet, P., & Vinter, A. (1998). PARSER: A model for word
788 segmentation. *Journal of Memory and Language*, 39, 246–263.
- Peters, A. (1983). *The units of language acquisition*. New York:
790 Cambridge University Press.
- Pinker, S. (1994). *Language learnability and language develop-*
792 *ment*. Cambridge, MA: Harvard University Press.
- Saffran, J. R., Aslin, R. N., & Newport, E. L. (1996). Statistical
794 learning by 8-month-old infants. *Science*, 274, 1926–1928.
- Saffran, J. R., Johnson, E. K., Aslin, R. N., & Newport, E. L.
796 (1999). Statistical learning of tone sequences by human
797 infants and adults. *Cognition*, 70, 27–52.
- Saffran, J. R., Newport, E. L., & Aslin, R. N. (1996). Word
799 segmentation: The role of distributional cues. *Journal of*
800 *Memory and Language*, 35, 606–621.
- Schön, D., Boyer, M., Moreno, S., Besson, M., Peretz, I., &
802 Kolinsky, R. (2008). Songs as an aid for language acquisi-
803 tion. *Cognition*, 106, 975–983.
- Snow, C. E., Vaneeden, R., & Muysken, P. (1981). The
805 interactional origins of foreigner talk – Municipal employees
806 and foreign-workers. *International Journal of the Sociology*
807 *of Language*, 81–91.
- Thiessen, E. D., & Saffran, J. R. (2003). When cues collide: Use of
809 stress and statistical cues to word boundaries by 7- to 9-month-
810 old infants. *Developmental Psychology*, 39, 706–716.
- 811
- Received January 18, 2009 812
- Revision received March 9, 2009 813
- Accepted March 16, 2009 814
- 815
- 816
- 817
-
- Toni Cunillera
-
- Dept. Psicologia Bàsica 818
- Facultat de Psicologia 819
- Universitat de Barcelona 820
- Passeig de la Vall d'Hebron 821
- 171 Barcelona 08035 822
- Spain 823
- Tel. +34 93 3125144 824
- Fax +34 93 4021363 825
- E-mail tcunillera@ub.edu 826
-
- 827