

Informatics 1 Cognitive Science

Lecture 8: Word Segmentation

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Speech Segmentation and Language Development

Transitional Probability

Word Segmentation Experiments

Minimum Description Length

Recap

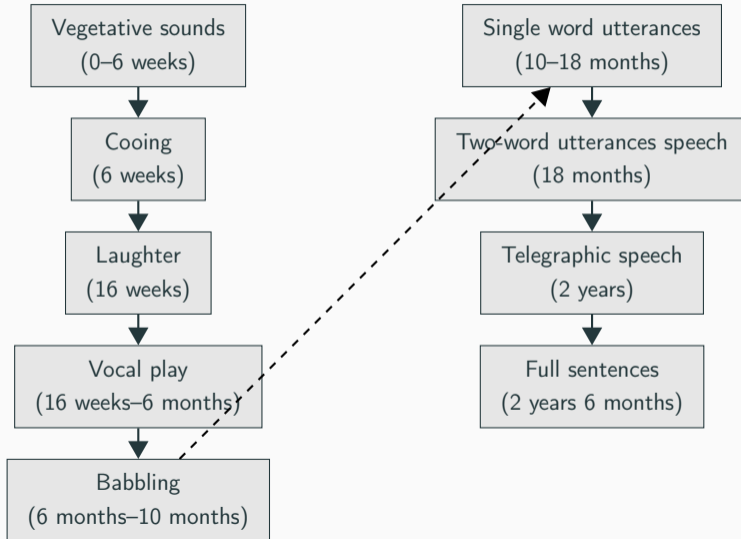
- So far, we have seen **rule-based models** and **neural network models**. These are at the extremes of the rationalist–empiricist debate.
- We've also seen how these two modeling frameworks can be applied to capture aspects of **language development**, such as past tense learning.
- Over the next few lectures, we will introduce a third modeling framework, **probabilistic modeling**.
- This approach offers a way of combining rules with numerical information (probabilities).
- The rules are pre-existing (maybe innate), while the probabilities are learned. So we combine aspects of rationalism and empiricism.
- Again, we will model aspects of language development: **word segmentation** (this lecture) and **word learning** (next week).

Speech Segmentation and Language Development

The Development of Language



The Development of Language



How Do We Learn Words?



- Knowing a language implies having a **mental lexicon**.
- Memorized set of associations among sound sequences, their meanings, and their syntax.
- Speech stream lacks any acoustic analog of the **blank spaces** between printed words.
- Basic units of linguistic input are not words but **entire utterances**.
- Child's task: to **discover the words** themselves in addition to meaning and syntax.

What do Infants Hear?

hamuchosañosquebuscoelyermo
hamuchosañosquevivotriste
hamuchosañosqueestoyenfermo
yesporellibroquetúescribiste
okempisantesdeleerteamaba
laluzlasvegaselmarocéano
mastúdiijistequetodoacaba
quetodomuerequetodoesvano

What do Infants Hear?

A Kempis by Amado Nervo

hamuchosañosquebuscoelyermo
hamuchosañosquevivotriste
hamuchosañosqueestoyenfermo
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quetodomuerequetodoesvano

<https://www.poemas-del-alma.com/a-kempis.htm>

What do Infants Hear?

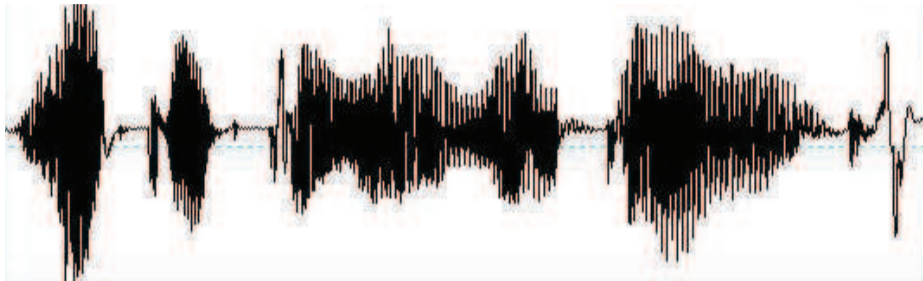
A Kempis by Amado Nervo

hamuchosañosquebuscoelyermo
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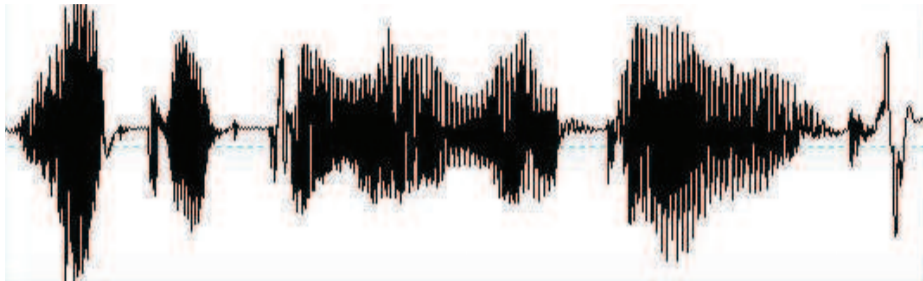
ASL demo: https://youtube.com/playlist?list=PLx1wHz1f-8J_xKVdU7DGa5RWIwWzRWNVt

Where Are the Words?



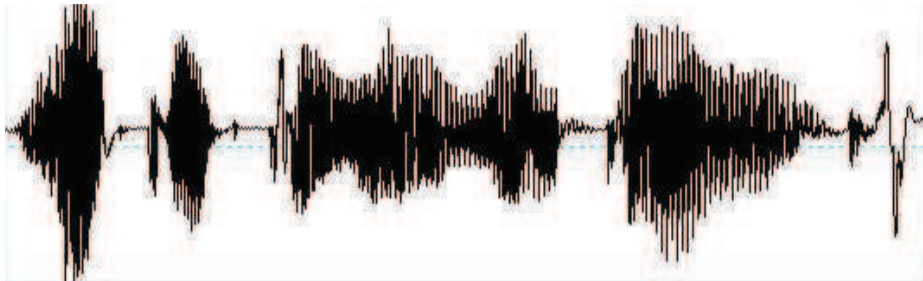
THEREDONATEAKETTLEOFTENCHIPS

Where Are the Words?



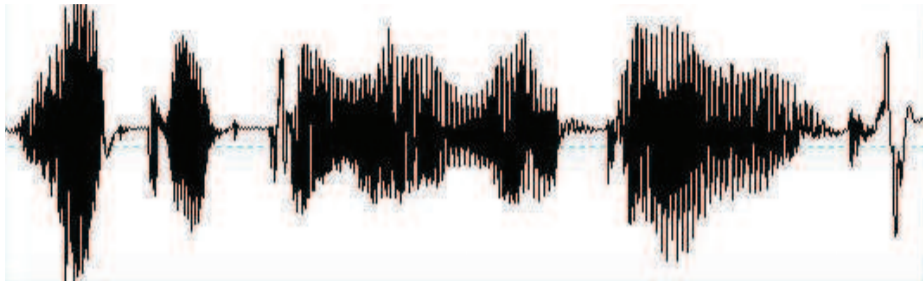
THEREDONATEAKETTLEOFTENCHIPS
THE RED ON A TEA KETTLE OFTEN CHIPS

Where Are the Words?



THEREDONATEAKETTLEOFTENCHIPS
THE RED ON A TEA KETTLE OFTEN CHIPS
THERE, DON ATE A KETTLE OF TEN CHIPS

Where Are the Words?



THEREDONATEAKETTLEOFTENCHIPS
THE RED ON A TEA KETTLE OFTEN CHIPS
THERE, DON ATE A KETTLE OF TEN CHIPS
THERE, DONATE A KETTLE OF TEN CHIPS

Important Questions

Things we need to understand before we can even start to study language acquisition:

- How does an infant divide the input into reusable units?
- How does she represent those units?
- What does she know about them and when?

This is not an end in itself: speech segmentation provides **useful units** (Peters, 1983) for learning a grammar: lexicon, morphology, syntax, phonology.

How do Infants Segment Speech?

Infants make use of **multiple cues** in the input, most popularly:

- **Stress patterns:** English usually stresses first syllable, French always the last; final syllables of words are longer (*hamster* vs. *ham stir*).
- **Phonotactic constraints:** every word must contain a vowel, finite set of consonant clusters at the beginning of a word, etc. (*gdog* not a possible English word).
- **Statistical regularities:** within words, there is a consistent sequence of elements.
- **Bootstrapping** from known words.

How do Infants Segment Speech?

Time for a short quiz on Wooclap!



<https://app.wooclap.com/FQGMXM>

Transitional Probability

Transitional Probability

Words create **regularities** in the sound sequences of a language.

- There is a **consistent sequence** of elements within words
- Sequences that don't occur within words can only occur at word boundaries.
- Sequences that don't occur within a word will tend to occur infrequently.
- Thus, we can find word boundaries by looking for **unlikely transitions**.

Transitional Probability

$$P(y|x) = \frac{p(x,y)}{p(x)} \approx \frac{\text{freq}(x,y)}{\text{freq}(x)}$$

Transitional Probability

Suppose the phoneme [ð] occurs 200,000 times in a text:

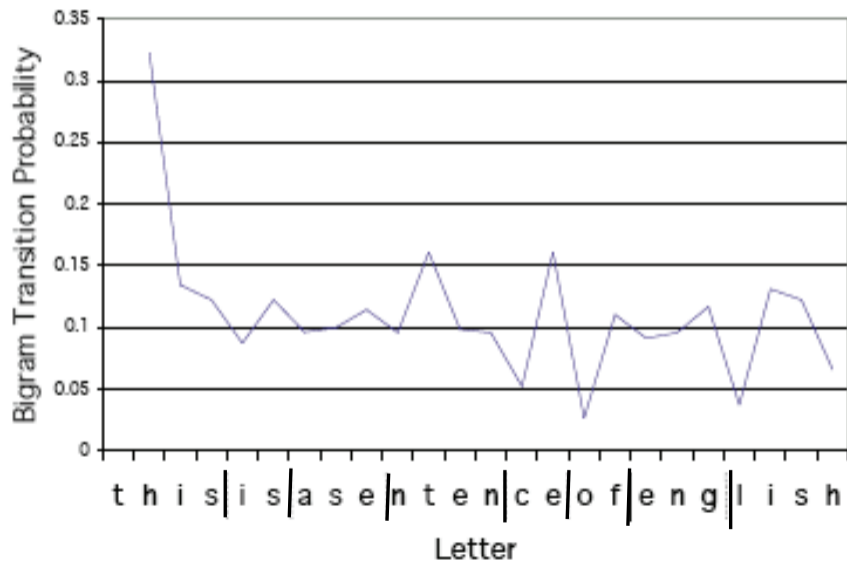
- 190,000 times are before a vowel (as in *the*, *this*);
- 200 times are before [m].

Transitional Probability

$$p(\text{vowel}|\text{ð}) = \frac{190,000}{200,000} = .95$$

$$P(m|\text{ð}) = \frac{200}{200,000} = .001$$

Transitional Probability



Word Segmentation Experiments

Do Children Make Use of Such Statistical Information?

Saffran et al. (1996) asked whether 8-month-old infants can extract information about word boundaries solely on the basis of statistics:

1. Create a “language” from nonsense words.
2. Infants listen to synthesized language (*tokibu, gikoba*).
3. Then, test: can infants distinguish words (*tokibu*) from part-words (*bugiko*)?

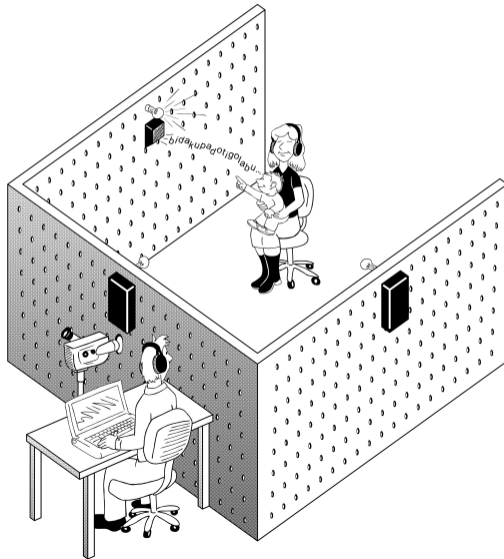
**pa bi ku ti bu do go la tu ti bu
do da ro pi pa bi ku go la tu ti
bu do pa bi ku go la tu da ro
pi pa bi ku da ro pi pa bi ku ti
bu do go la tu ti bu do**

pa bi ku ti bu do go la tu ti bu
do da ro pi pa bi ku go la tu ti
bu do pa bi ku go la tu da ro
pi pa bi ku da ro pi pa bi ku ti
bu do go la tu ti bu do

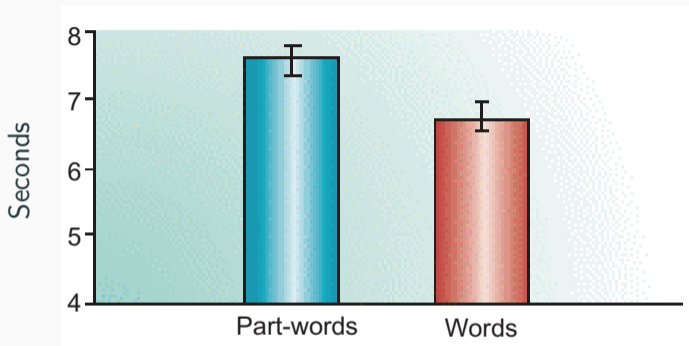
Word Segmentation Experiments

- Infants are exposed for 2 minutes to nonsense language (*tokibu, gopila, gikoba, tipolu*).
- Only **statistical cues** to word boundaries.
- Then record how long they attend to novel sets of stimuli that either do or do not share some property with the familiarization data.
- Discrimination between *words* and *part-words* (sequences spanning word boundaries)
- If **there's a difference**, there has been some **learning** during familiarization.

Headturn Preference Procedure



Results



- Infants show longer listening times for part-words
- Infants can extract information about sequential statistics of syllables (input contained no pauses or intonational patterns)

Saffran's work (and much subsequent research) shows:

- Humans can use statistical information to segment speech.
- But all words were trisyllabic.
- So, transitional probabilities were either 1 or .33
- Will this work if these are varied in a more naturalistic way?

Patricia Kuhl: The genius of babies

https://www.ted.com/talks/patricia_kuhl_the_linguistic_genius_of_babies

Time for a short quiz on Wooclap!



<https://app.wooclap.com/FQGMXM>

Minimum Description Length

Lexicons and Segmentation

- The use of transitional probabilities to do word segmentation is not sufficient.
- It ignores the fact that many **words** are being **learned at the same time**.
- There are statistical methods for speech segmentation that incorporate the learning of a lexicon as a sub-component.
- Brent and Cartwright (1996): find the **lexicon** which **minimizes the description** of the observed data

Minimum Description Length

$$\text{size}(\text{description}) = \text{size}(\text{lexicon}) + \text{size}(\text{data-encoding})$$

Minimum Description Length

$$\text{size}(\text{description}) = \text{size}(\text{lexicon}) + \text{size}(\text{data-encoding})$$

- The MDL principle minimizes the length of words:
shorter words are more plausible
- It minimizes the number of different words:
try to make use of words you already know
- It maximizes the probability of each word:
words recur as often as possible

Brent and Cartwright (1996)

Input

doyouseethekitty

seethekitty

doyoulikethekitty

Brent and Cartwright (1996)

Input

doyouseethekitty

seethekitty

doyoulikethekitty

Segmentation 1

do you see thekitty

see thekitty

do you like thekitty

Brent and Cartwright (1996)

Input

doyouseethekitty

seethekitty

doyoulikethekitty

Segmentation 1

do you see thekitty

see thekitty

do you like thekitty

Lexicon 1

1 do 2 thekitty 3 you

4 like 5 see

Brent and Cartwright (1996)

Input

doyouseethekitty

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Segmentation 1

do you see thekitty

see thekitty

do you like thekitty

Lexicon 1

1 do 2 thekitty 3 you

4 like 5 see

Derivation 1

1 3 5 2

5 2

1 3 4 2

Brent and Cartwright (1996)

Input

doyouseethekitty

seethekitty

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Segmentation 1

do you see thekitty

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Derivation 1

1 3 5 2

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1 3 4 2

Minimum Description Length

$$\text{size}(\text{description}) = \text{size}(\text{lexicon}) + \text{size}(\text{data-encoding})$$
$$\text{size}(\text{lexicon}) = \text{number of characters} \\ \text{characters} = \text{letters and digits}$$
$$\text{size}(\text{data-encoding}) = \text{number of} \\ \text{characters in derivation}$$

Brent and Cartwright (1996)

Input

doyouseethekitty

seethekitty

doyoulikethekitty

Segmentation 1

do you see thekitty

see thekitty

do you like thekitty

Lexicon 1

1 do 2 thekitty 3 you

4 like 5 see

Derivation 1

1 3 5 2

5 2

1 3 4 2

Minimum Description Length

$\text{size}(\text{description}) = \text{size}(\text{lexicon}) + \text{size}(\text{data-encoding})$

$\text{size}(\text{lexicon}) = \text{number of characters}$
characters = letters and digits

$\text{size}(\text{data-encoding}) = \text{number of characters in derivation}$

Length: 25 + 10 = 35

Brent and Cartwright (1996)

Input

doyouseethekitty

seethekitty

doyoulikethekitty

Segmentation 2

do you see the kitty

see the kitty

do you like the kitty

Lexicon 2

1 do 2 the 3 you

4 like 5 see 6 kitty

Derivation 2

1 3 5 2 6

5 2 6

1 3 4 2 6

Minimum Description Length

$$\text{size}(\text{description}) = \text{size}(\text{lexicon}) + \text{size}(\text{data-encoding})$$
$$\text{size}(\text{lexicon}) = \text{number of characters} \\ \text{characters} = \text{letters and digits}$$
$$\text{size}(\text{data-encoding}) = \text{number of} \\ \text{characters in derivation}$$

Brent and Cartwright (1996)

Input

doyouseethekitty

seethekitty

doyoulikethekitty

Segmentation 2

do you see the kitty

see the kitty

do you like the kitty

Lexicon 2

1 do 2 the 3 you

4 like 5 see 6 kitty

Derivation 2

1 3 5 2 6

5 2 6

1 3 4 2 6

Minimum Description Length

$\text{size}(\text{description}) = \text{size}(\text{lexicon}) + \text{size}(\text{data-encoding})$

$\text{size}(\text{lexicon}) = \text{number of characters}$
characters = letters and digits

$\text{size}(\text{data-encoding}) = \text{number of characters in derivation}$

Length: 26 + 13 = 39

Brent and Cartwright (1996)

- MDL model is tested on (phonetically) transcribed speech from the CHILDES corpus.
- An **idealization of** the raw **acoustic signal**.
- Model searches for segmentation of the input with least MDL.
- Search algorithm is **not incremental**; it reads in the entire input before segmenting any part of it.
- Approach does not rely on language-specific input!
- Computational simulations systematically explore hypothesis that distributional regularity is useful for word segmentation.

In order to acquire a lexicon young children segment speech into words using multiple sources of support.

In this lecture, we focused on distributional regularities:

- transitional probability provides cues
- verified by Saffran et al. (1996) experiments
- computational model of word segmentation
- based on Minimum Description Length Principle

Next lecture: Bayesian modeling.