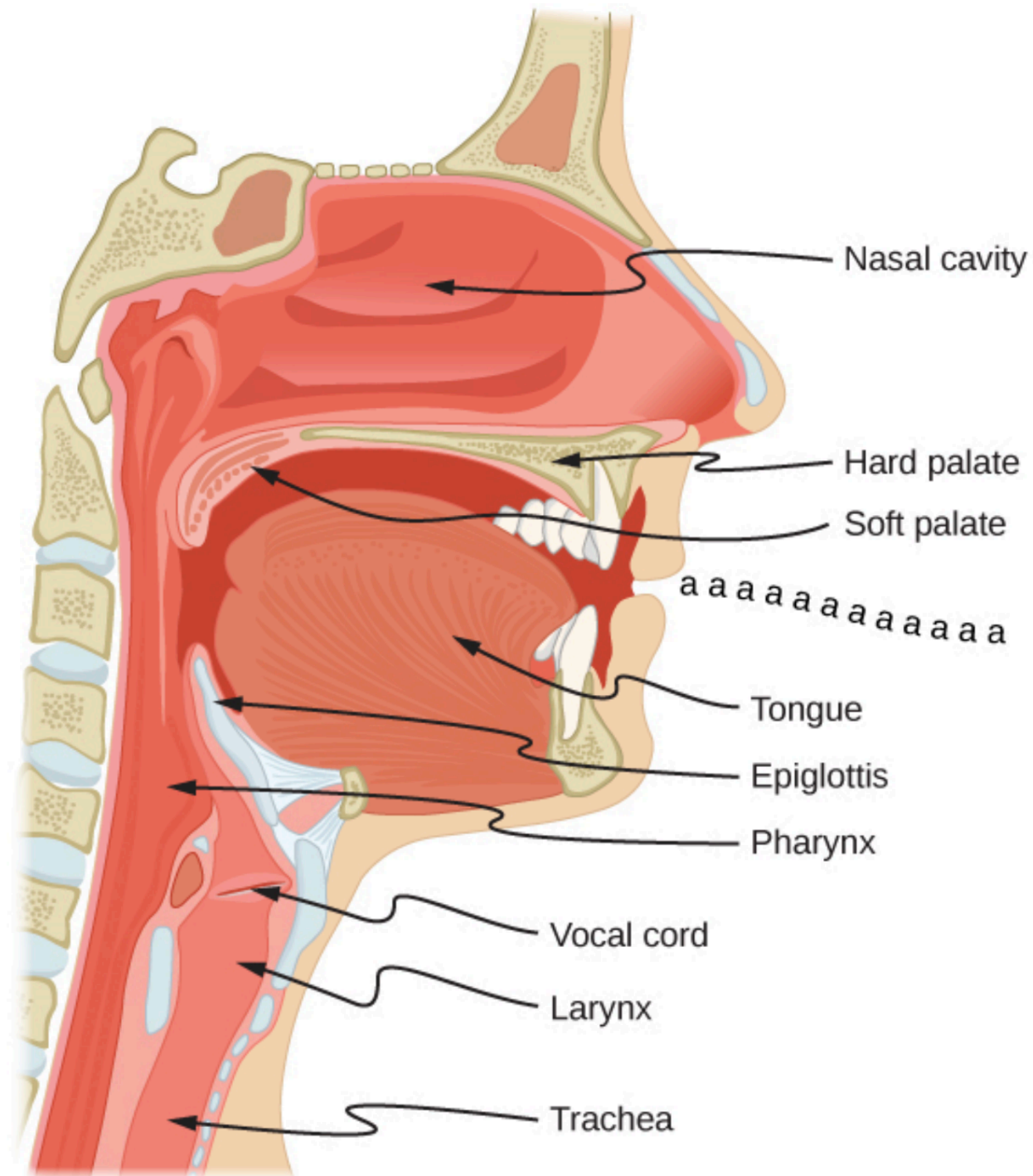


Linguistics: Speech and Lexical Semantics



EECS 183/283a: Natural Language Processing

Speech Production



Vocal articulators that produce speech

- Air passing through vocal articulators produces speech
- Vocal folds, tongue, jaw, lips, velum are both independently and jointly controlled to produce different sounds
 - E.g., vocal fold vibration causes voicing
- The output of vocal articulation is an acoustic pressure wave

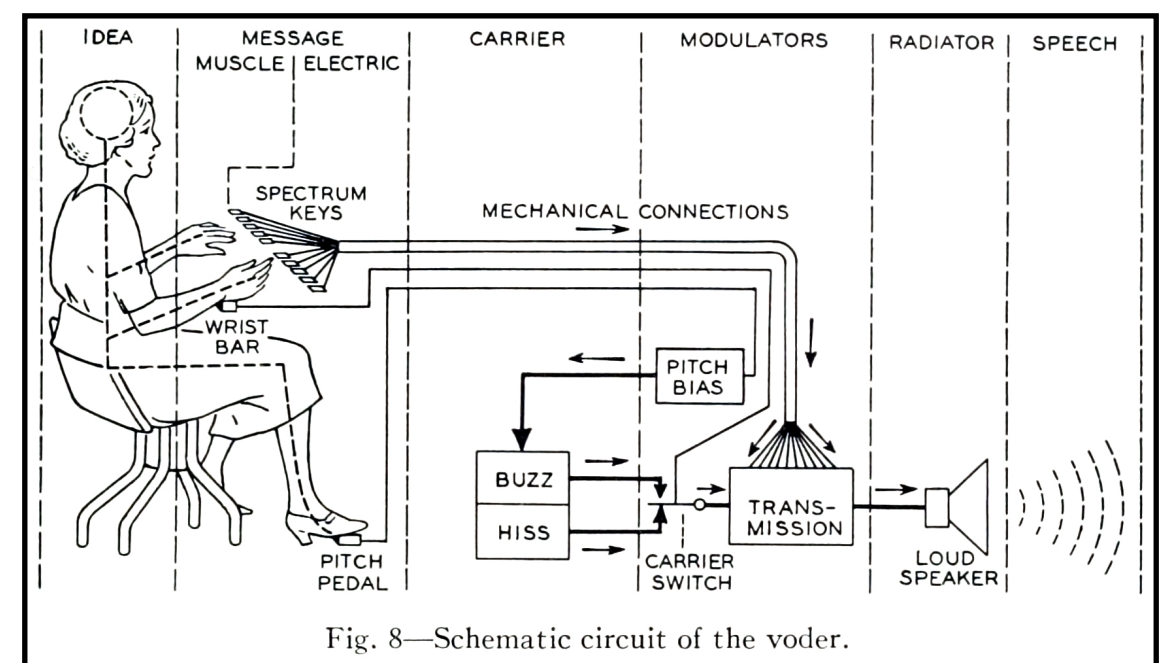
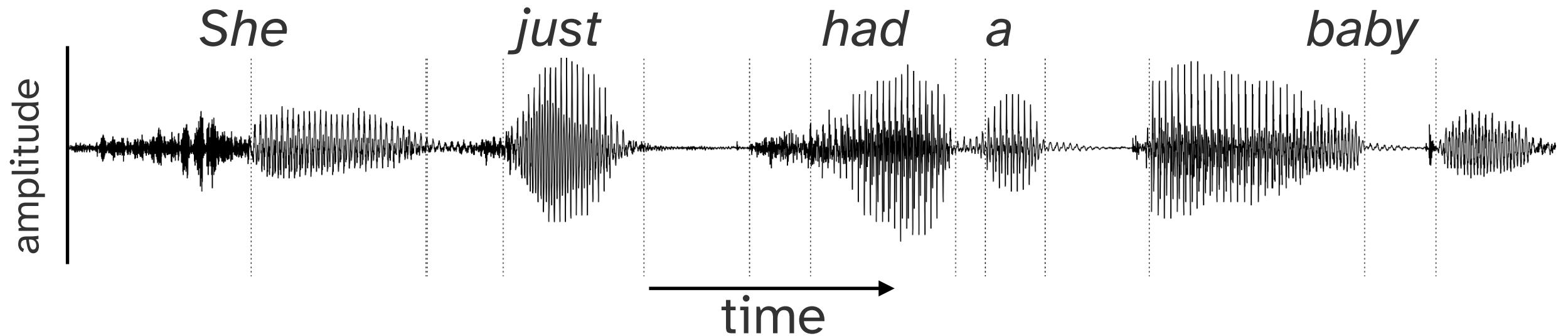


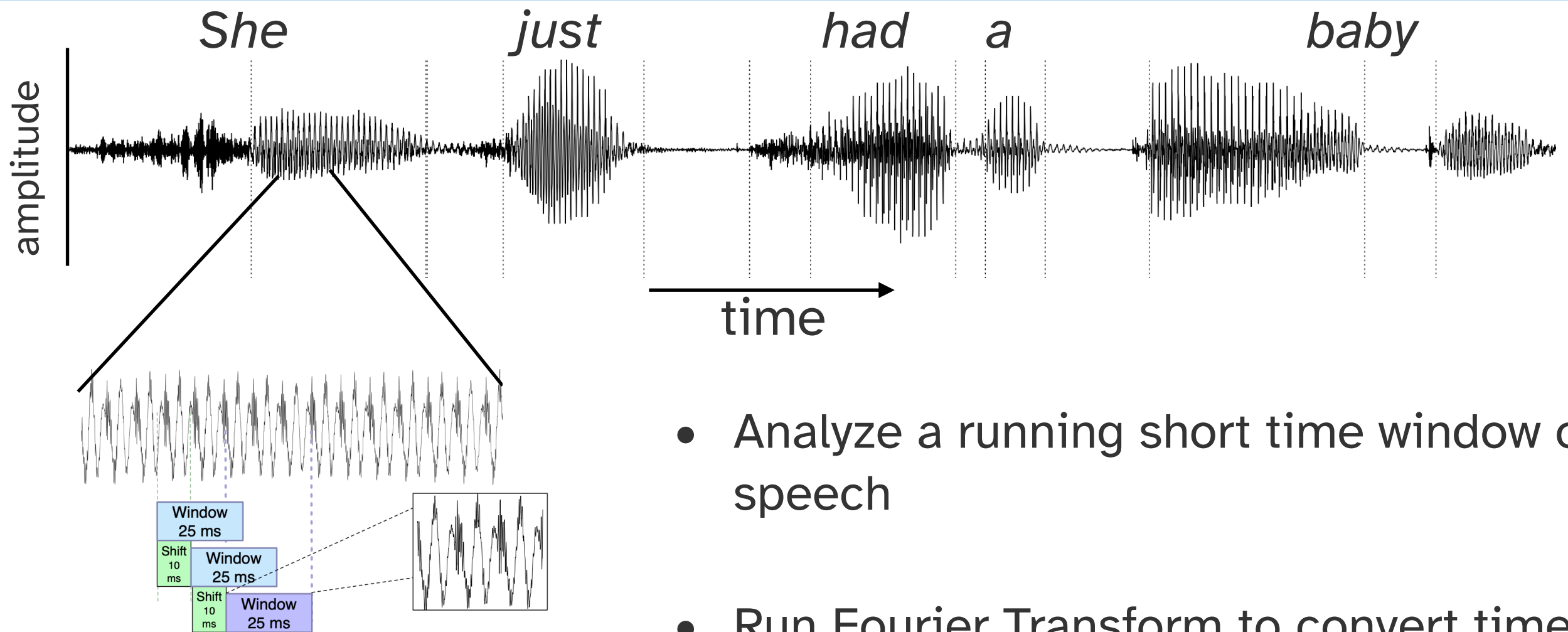
Fig. 8—Schematic circuit of the voder.

Speech Representations



- The result of the vocal articulation is an acoustic pressure wave
- Speech can thus be represented as an **acoustic waveform**
- Waveforms are continuous time series cannot be easily analyzed or interpreted, or computed with
- Signal processing can give more interpretable information

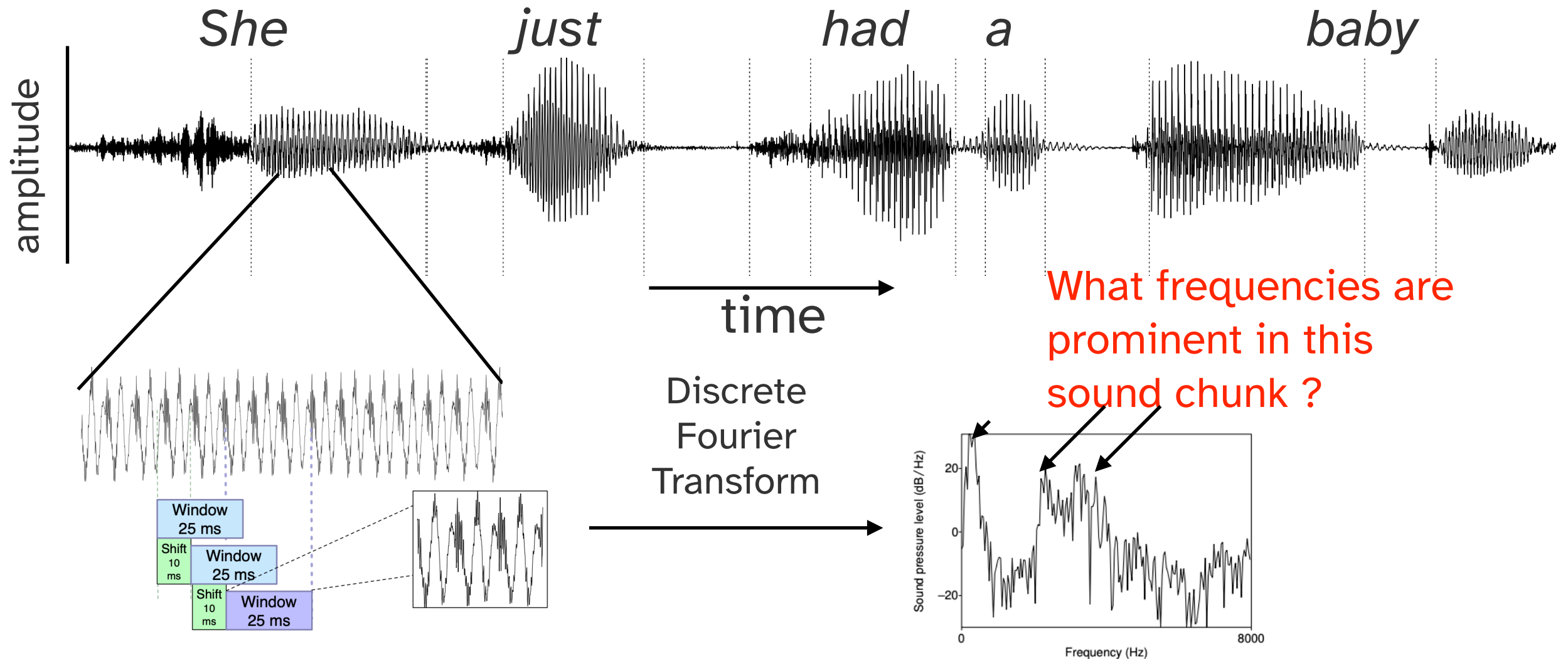
Speech Waveform



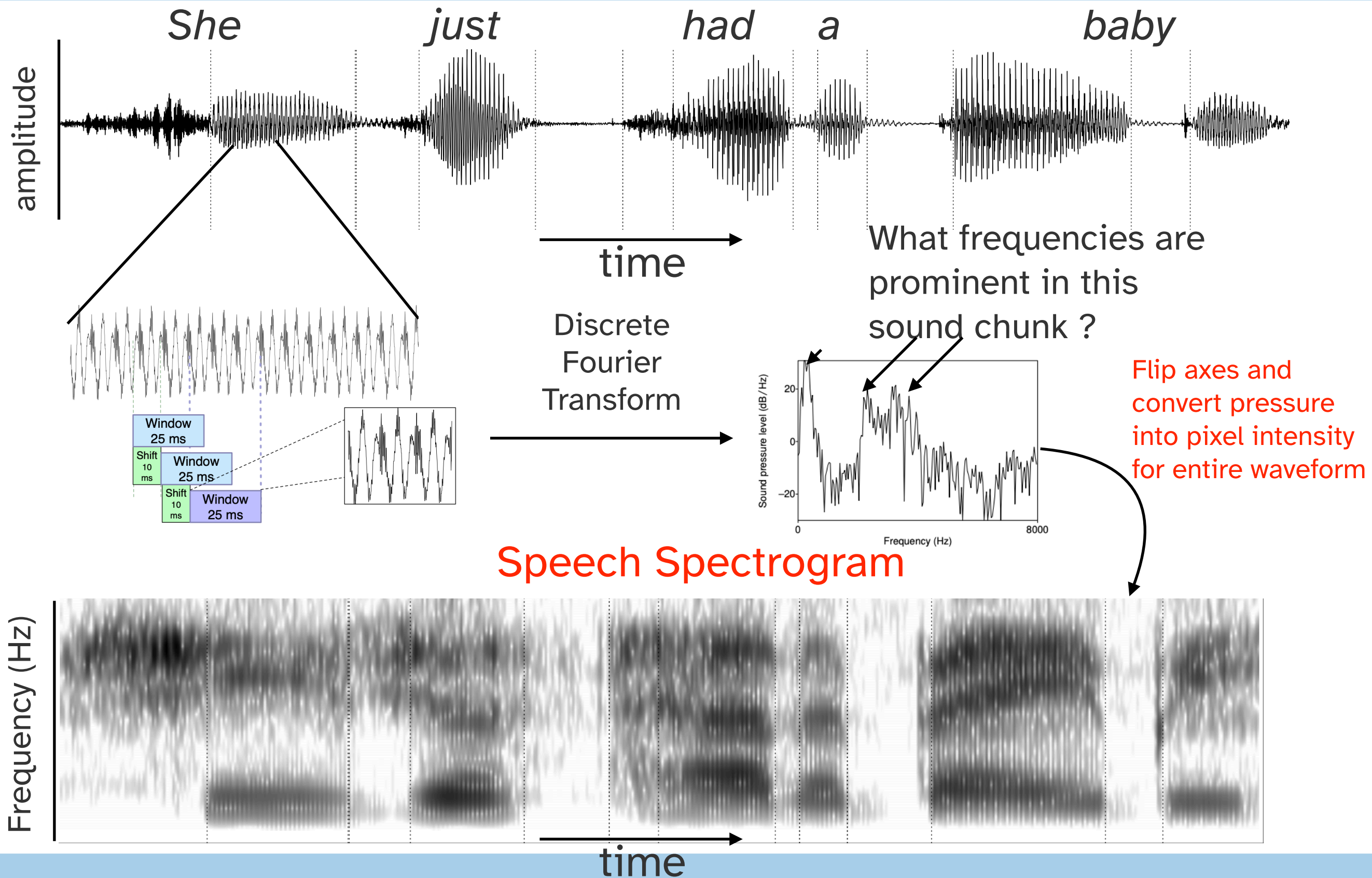
- Analyze a running short time window of speech
- Run Fourier Transform to convert time to time-frequency representation



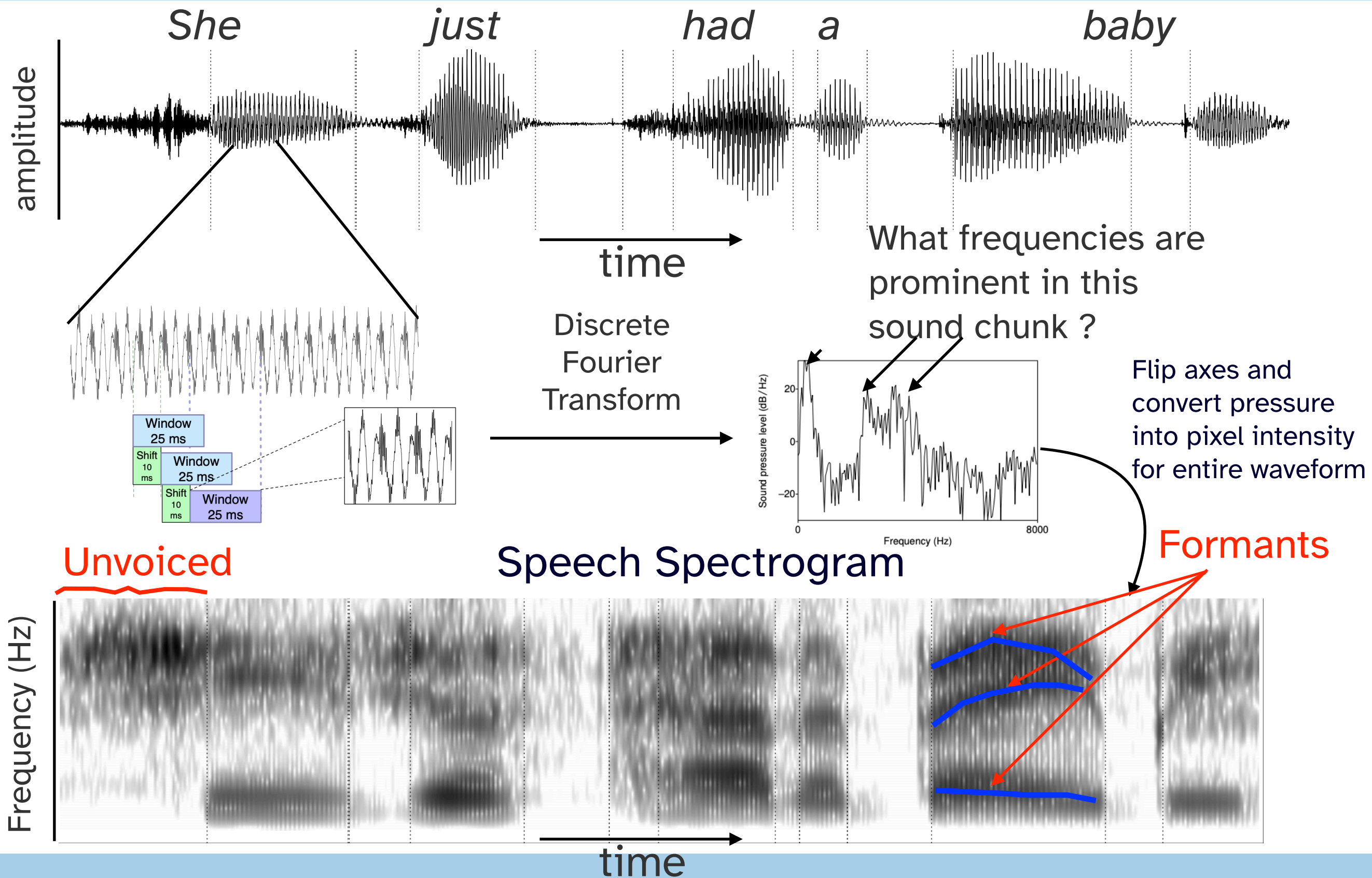
Speech Waveform



Speech Spectrogram



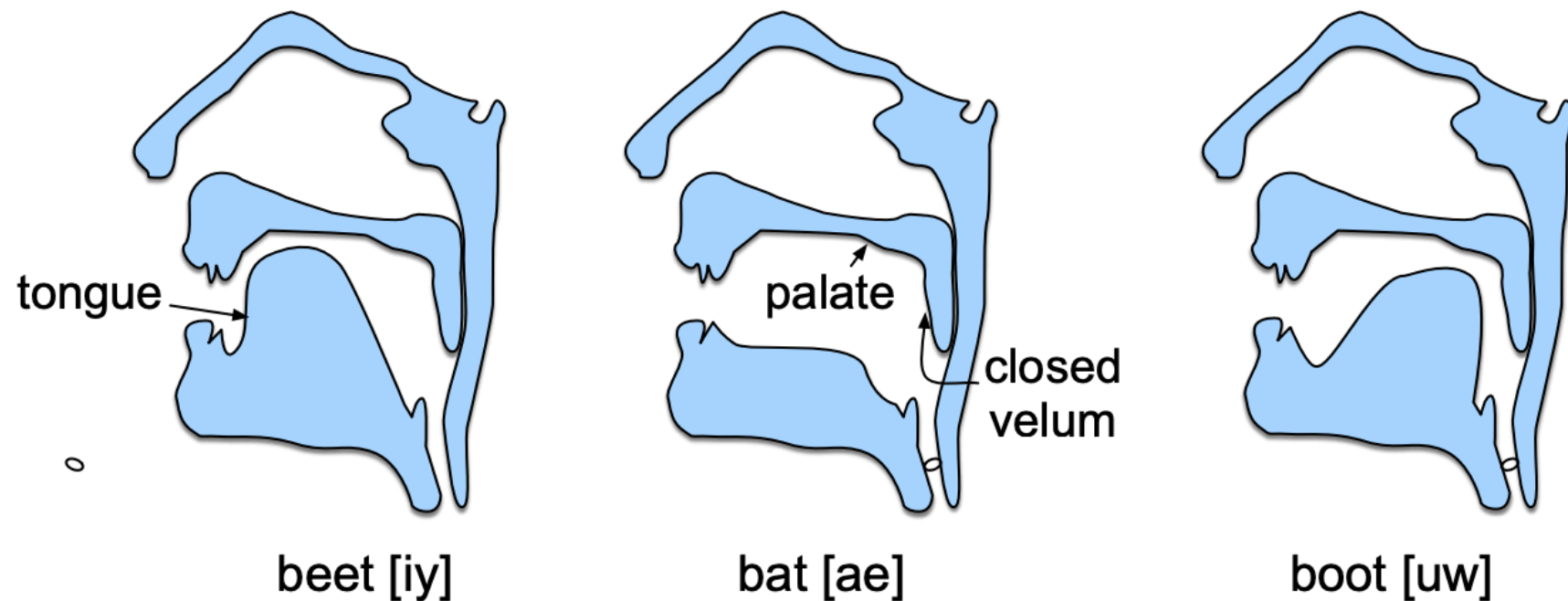
Speech Spectrogram



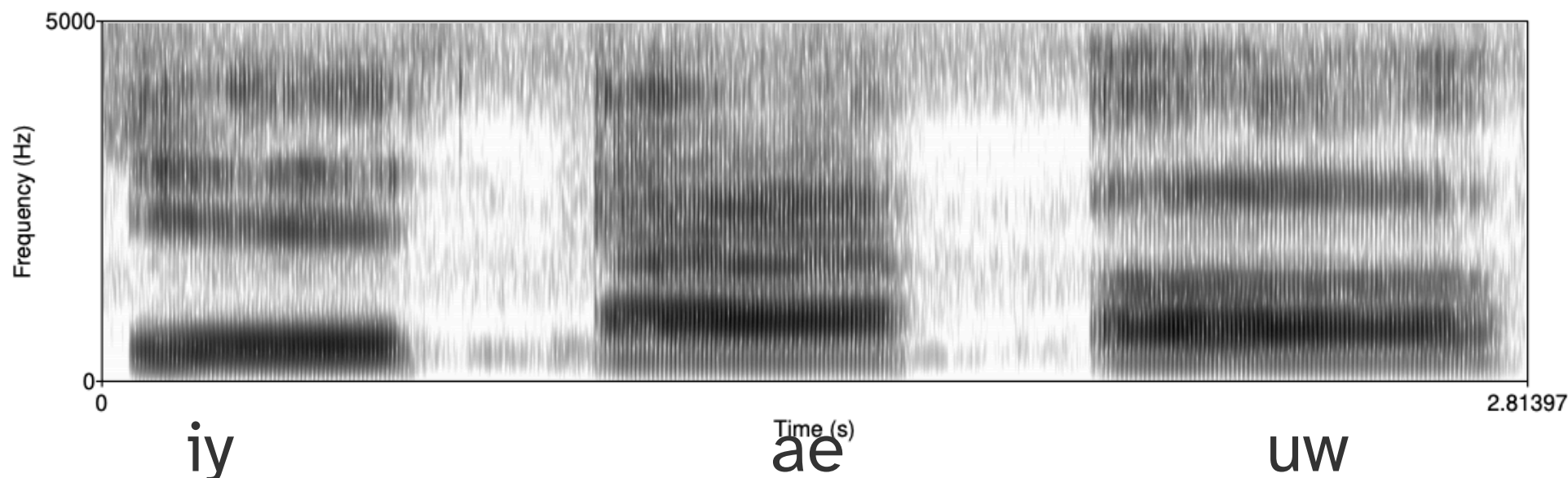
Phonetics



Study of speech sounds — their physical production, spectral and perceptual properties

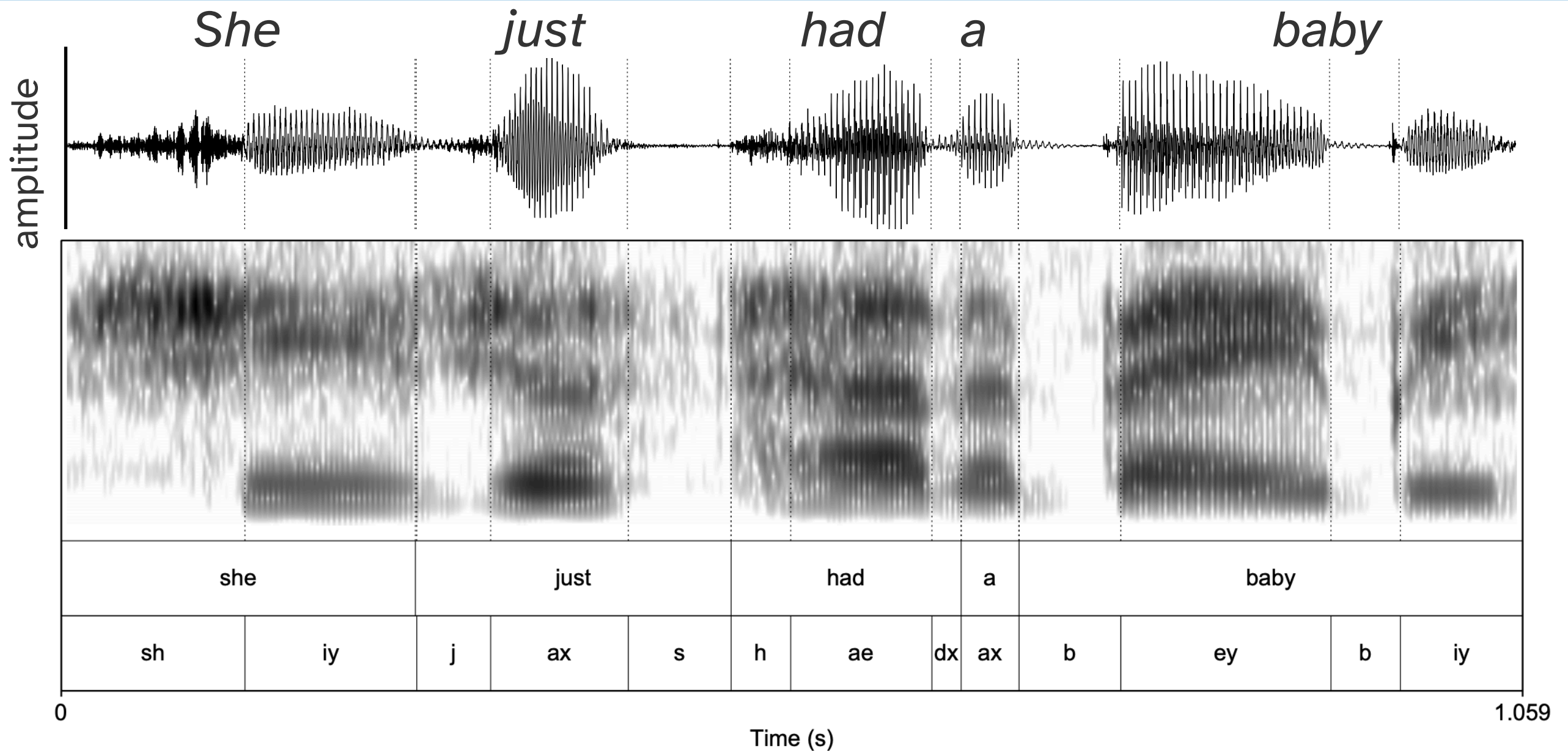


Articulatory
Phonetics



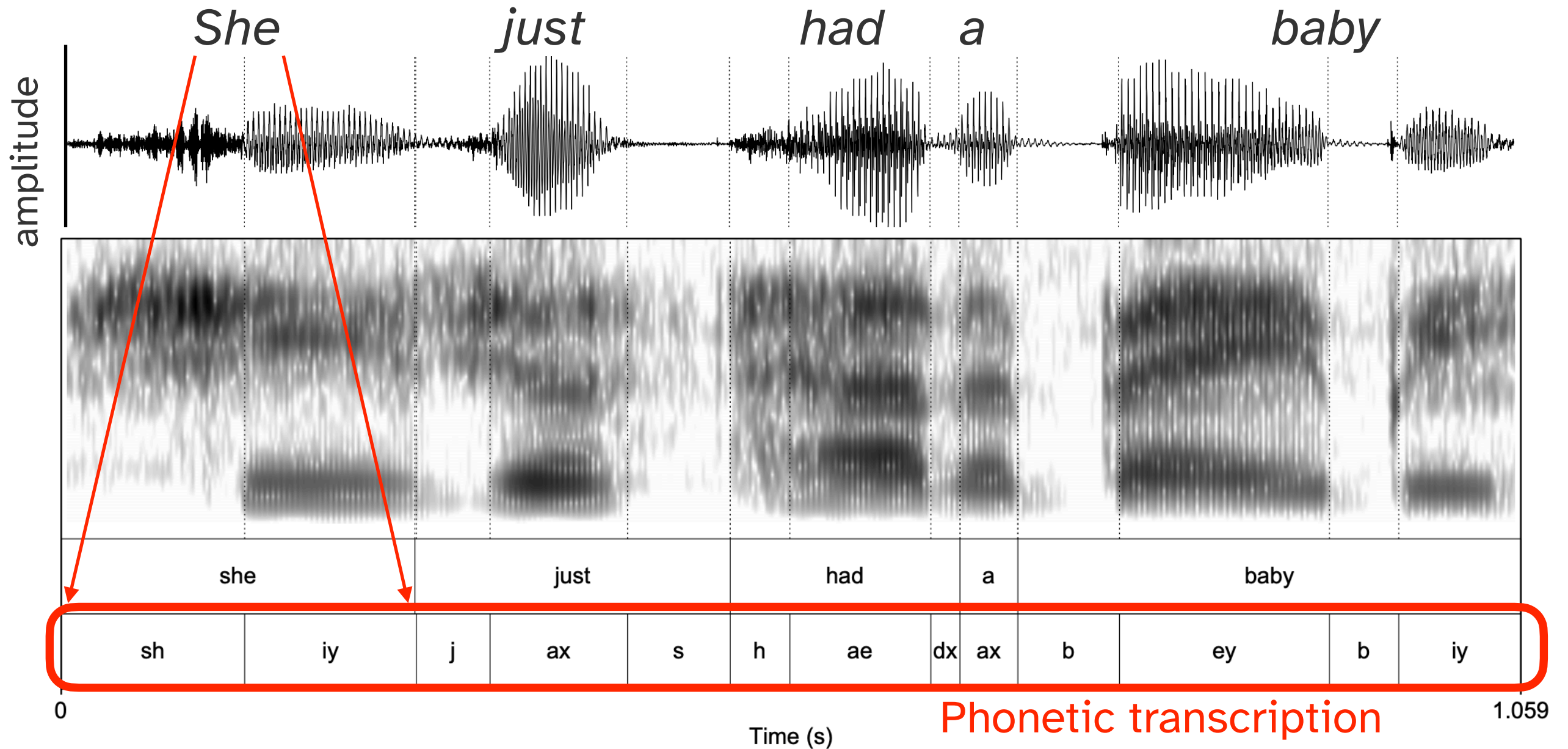
Acoustic
Phonetics

Acoustic Phonetics



- Spectrogram reveals some segmental structure with distinct properties
- These are **phonemes** — perceptually distinct speech sounds

Acoustic Phonetics



- Spectrogram reveals some segmental structure with distinct properties
- These are **phonemes** — perceptually distinct speech sounds

- women
ghoti
though nation

© 2020 IPA

Symbols to the right in a cell are voiced, to the left are voiceless. Shaded areas denote articulations judged impossible.

VOWELS

ɱ Voiceless labial-velar fricative	ɕ ʑ Alveolo-palatal fricatives
ʋ Voiced labial-velar approximant	ɭ Voiced alveolar lateral flap
ɥ Voiced labial-palatal approximant	ɥ̟ Simultaneous ɥ and ɣ
ħ Voiceless epiglottal fricative	
ʕ Voiced epiglottal fricative	Affricates and double articulations
ʡ Epiglottal plosive	can be represented by two symbols
	joined by a tie bar if necessary.

◌ [◌] Voiceless	<u>n</u> <u>d</u>	◌ [◌] Breathy voiced	<u>ṇ</u> <u>ḍ</u>	◌ [◌] Dental	<u>ṅ</u> <u>ṁ</u>
◌ [◌] Voiced	<u>s</u> <u>t</u>	◌ [◌] Creaky voiced	<u>ḥ</u> <u>ṭ</u>	◌ [◌] Apical	<u>ṇ</u> <u>ṁ</u>
^h Aspirated	^h t ^h d ^h	◌ [◌] Linguolabial	<u>ṭ</u> <u>ḍ</u>	◌ [◌] Laminal	<u>ṇ</u> <u>ṁ</u>
◌ [◌] More rounded	<u>ɔ</u>	◌ [◌] Labialized	<u>ṭ^w</u> <u>ḍ^w</u>	◌ [◌] Nasalized	<u>ṇ̃</u>
◌ [◌] Less rounded	<u>ɔ̄</u>	◌ [◌] Palatalized	<u>ṭ^j</u> <u>ḍ^j</u>	◌ [◌] Nasal release	<u>ṇ̥</u>
◌ [◌] Advanced	<u>u</u>	◌ [◌] Velarized	<u>ṭ^ʷ</u> <u>ḍ^ʷ</u>	◌ [◌] Lateral release	<u>ṇ̥^l</u>
◌ [◌] Retracted	<u>ɐ</u>	◌ [◌] Pharyngealized	<u>ṭ^ʕ</u> <u>ḍ^ʕ</u>	◌ [◌] No audible release	<u>ṇ̥⁰</u>
◌ [◌] Centralized	<u>ë</u>	◌ [◌] Velarized or pharyngealized	<u>ṭ̤</u>		
◌ [◌] Mid-centralized	<u>ẽ</u>	◌ [◌] Raised	<u>ṭ̚</u> (<u>ṭ̚</u> = voiced alveolar fricative)		
◌ [◌] Syllabic	<u>n̩</u>	◌ [◌] Lowered	<u>ṭ̤̚</u> (<u>ṭ̤̚</u> = voiced bilabial approximant)		
◌ [◌] Non-syllabic	<u>ɐ̯</u>	◌ [◌] Advanced Tongue Root	<u>ṭ̤̚̚</u>		
◌ [◌] Rhoticity	<u>ɐ̯</u> <u>ṛ</u>	◌ [◌] Retracted Tongue Root	<u>ṭ̤̚̚̚</u>		

Some diacritics may be placed above a symbol with a descender, e.g. $\overset{\circ}{n}$

	Primary stress	ˈ	fəʊnə	ˈ	tɪfən
	Secondary stress				
ː	Long		eː		
ˑ	Half-long		eˑ		
◌	Extra-short		ẽ		
	Minor (foot) group				
	Major (intonation) group				
.	Syllable break		ɪ.ækt		
◌◌◌	Linking (absence of a break)				

LEVEL	CONTOUR
ě or  Extra high	ě or  Rising
é  High	ê  Falling
ē  Mid	ẽ  High rising
è  Low	ẽ  Low rising
ě  Extra low	ẽ  Rising-falling
↓ Downstep	 Global rise
↑ Upstep	 Global fall

International Phonetic Alphabet (IPA)



- Phoneticians compiled a common set of sounds used to codify different speech sounds (across languages)

Central Rotokas
(Papua New Guinea): 12 consonants, 10 vowels

	Bilabial	Alveolar	Velar
Voiceless	p	t	k
Voiced	b	d	g

	Front	Central	Back
Close	i (i:)		u (u:)
Close-mid	e (e:)		o (o:)
Open		a (a:)	

Archi
(Dagestan, Russia)

			Labial	Dental		(Post)-alveolar		(Pre-)velar		Uvular				Epiglottal	Glottal
				plain	lab.	plain	lab.	plain	lab.	plain	lab.	phar.	phar.+lab.		
Nasal			m	n											
Plosive	voiced		b	d	d ^{w2}			g	g ^w						
	voiceless		p	t	t ^w			k	k ^w	q	q ^w	q ^ɿ	q ^{ɿw}	ʔ	ʔ ¹
	fortis		p: ¹	t: ¹				k: ¹	k ^w : ²	q: ¹		q ^ɿ : ¹			
	ejective		p' ¹	t' ¹				k' ¹	k ^w ': ²	q' ¹	q ^w ': ¹	q ^ɿ ': ¹	q ^{ɿw} ': ¹		
Affricate	voiceless	lenis		ts	ts ^{w2}	tʃ	tʃ ^w	k ^ɿ	k ^{wɿ}						
		fortis		ts: ³											
	ejective	lenis		ts'	ts ^w ': ¹	tʃ'	tʃ ^w ': ¹	k ^ɿ ': ¹	k ^{wɿ} ': ¹						
		fortis		ts': ¹		tʃ': ²									
Fricative	voiceless	lenis		s	s ^{w2}	ʃ	ʃ ^w	ɬ	ɬ ^w	χ	χ ^w	χ ^ɿ	χ ^{ɿw}		h
		fortis		s:	s ^w : ²	ʃ:	ʃ ^w :	ɬ:	ɬ ^w :	χ:	χ ^w :	χ ^ɿ :	χ ^{ɿw} :		
	voiced			z	z ^w	ʒ	ʒ ^w	ɮ ¹		ʁ	ʁ ^{w2}	ʁ ^ɿ	ʁ ^{ɿw}		
Trill				r										ʀ	
Approximant				l		j		w							

THE INTERNATIONAL PHONETIC ALPHABET (revised to 2020)

CONSONANTS (PULMONIC)

© 2020 IPA

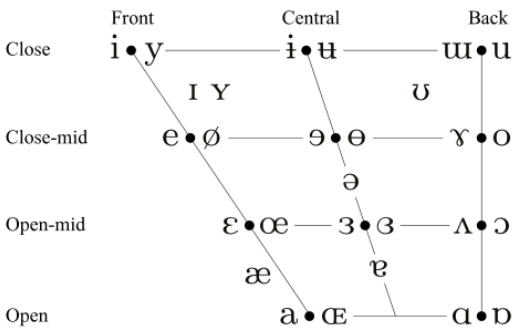
	Bilabial	Labiodental	Dental	Alveolar	Postalveolar	Retroflex	Palatal	Velar	Uvular	Pharyngeal	Glottal
Plosive	p b			t d		ʈ ɖ	c ɟ	k ɡ	q ɢ		ʔ
Nasal	m	ɱ		n		ɳ	ɲ	ŋ	ɴ		
Trill	ʙ			r					ʀ		
Tap or Flap		ⱱ		ɾ		ɽ					
Fricative	ɸ β	f v	θ ð	s z	ʃ ʒ	ʂ ʐ	ç ʝ	x ɣ	χ ʁ	ħ ʕ	h ɦ
Lateral fricative				ɬ ɮ							
Approximant		ʋ		ɹ		ɻ	j	ɰ			
Lateral approximant				l		ɭ	ʎ	ʟ			

Symbols to the right in a cell are voiced, to the left are voiceless. Shaded areas denote articulations judged impossible.

CONSONANTS (NON-PULMONIC)

Clicks	Voiced implosives	Ejectives
ʘ Bilabial	ɓ Bilabial	ʼ Examples:
ǀ Dental	ɗ Dental/alveolar	pʼ Bilabial
ǃ (Post)alveolar	ɟ Palatal	tʼ Dental/alveolar
ǂ Palatoalveolar	ɡ Velar	kʼ Velar
ǁ Alveolar lateral	ɠ Uvular	sʼ Alveolar fricative

VOWELS



Where symbols appear in pairs, the one to the right represents a rounded vowel.

OTHER SYMBOLS

ɱ Voiceless labial-velar fricative	ɕ ʑ Alveolo-palatal fricatives
ʋ Voiced labial-velar approximant	ɭ Voiced alveolar lateral flap
ɰ Voiced labial-palatal approximant	ɧ Simultaneous ʃ and x
ħ Voiceless epiglottal fricative	
ʕ Voiced epiglottal fricative	Affricates and double articulations can be represented by two symbols joined by a tie bar if necessary.
ʡ Epiglottal plosive	

ts kp

DIACRITICS

◌ [◌] Voiceless	◌ [◌] Breathy voiced	◌ [◌] Dental
◌ [◌] Voiced	◌ [◌] Creaky voiced	◌ [◌] Apical
◌ ^h Aspirated	◌ ^h Linguolabial	◌ ^h Laminal
◌ [◌] More rounded	◌ [◌] Labialized	◌ [◌] Nasalized
◌ [◌] Less rounded	◌ [◌] Palatalized	◌ [◌] Nasal release
◌ [◌] Advanced	◌ [◌] Velarized	◌ [◌] Lateral release
◌ [◌] Retracted	◌ [◌] Pharyngealized	◌ [◌] No audible release
◌ [◌] Centralized	◌ [◌] Velarized or pharyngealized	
◌ [◌] Mid-centralized	◌ [◌] Raised	
◌ [◌] Syllabic	◌ [◌] Lowered	
◌ [◌] Non-syllabic	◌ [◌] Advanced Tongue Root	
◌ [◌] Rhoticity	◌ [◌] Retracted Tongue Root	

Some diacritics may be placed above a symbol with a descender, e.g. ɳ̥̊

SUPRASEGMENTALS

ˈ Primary stress	ˈfounəˈtɪʃən
ˌ Secondary stress	
ː Long	eː
ˑ Half-long	eˑ
◌̥ Extra-short	ẽ
◌̥ Minor (foot) group	
◌̥ Major (intonation) group	
◌̥ Syllable break	ˌi.ækt
◌̥ Linking (absence of a break)	

TONES AND WORD ACCENTS

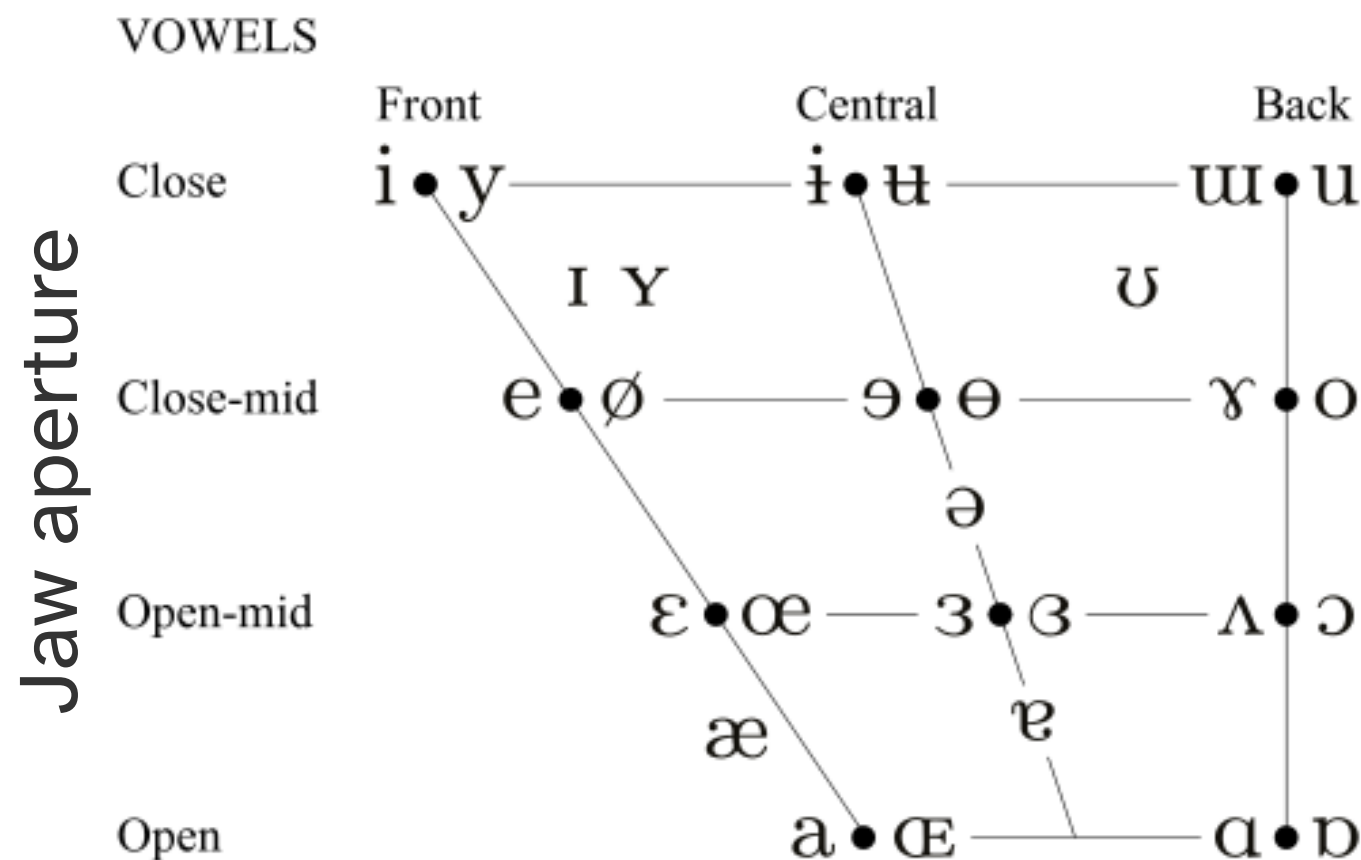
LEVEL	CONTOUR
ẽ or ̥	Extra high
é or ̥	High
ē or ̥	Mid
è or ̥	Low
ë or ̥	Extra low
↓	Downstep
↑	Upstep
ẽ or ̥	Rising
ē or ̥	Falling
ē or ̥	High rising
è or ̥	Low rising
ë or ̥	Rising-falling
↗	Global rise
↘	Global fall

International Phonetic Alphabet (IPA)

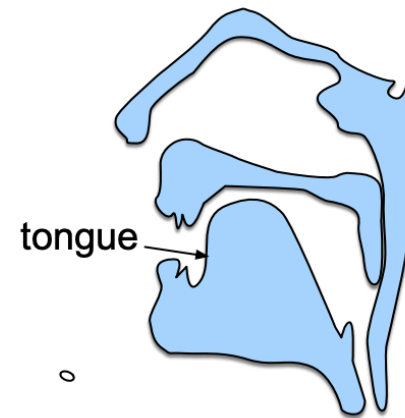


- Vowels are characterized by jaw position and tongue shape
- Some vowels also use lips (eg. sound *uw* in *cool*)

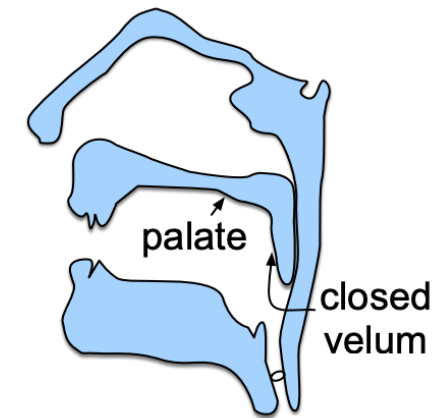
Tongue frontness



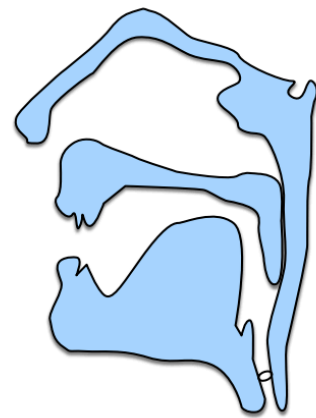
Where symbols appear in pairs, the one to the right represents a rounded vowel.



beet [iy]



bat [ae]

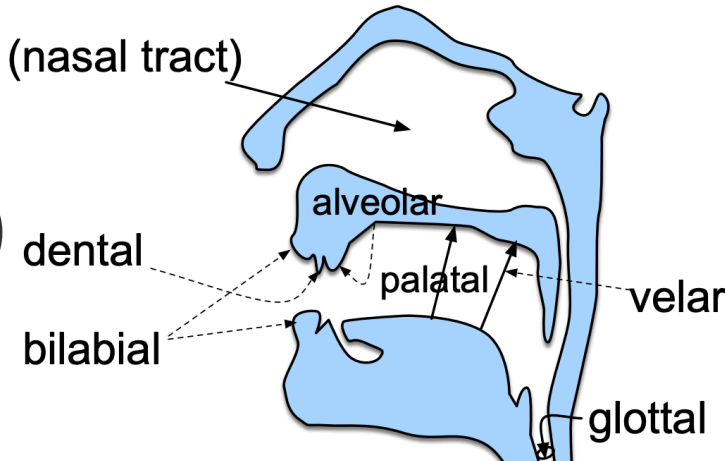


boot [uw]

International Phonetic Alphabet (IPA)



- Consonants are characterized by place and manner of articulation
 - /p/ is caused by constriction at lips (labial)
 - /p/ is caused by sudden release of air (plosive)



Place of articulation

CONSONANTS (PULMONIC)

© 2020 IPA

Manner of articulation

	Bilabial	Labiodental	Dental	Alveolar	Postalveolar	Retroflex	Palatal	Velar	Uvular	Pharyngeal	Glottal
Plosive	p b		t d			ʈ ɖ	c ɟ	k ɡ	q ɢ		ʔ
Nasal	m	ɱ		n		ɳ	ɲ	ŋ	ɴ		
Trill	ʙ			r					ʀ		
Tap or Flap		ⱱ		ɾ		ɽ					
Fricative	ɸ β	f v	θ ð	s z	ʃ ʒ	ʂ ʐ	ç ʝ	x ɣ	χ ʁ	ħ ʕ	h ɦ
Lateral fricative				ɬ ɮ							
Approximant		ʋ		ɹ		ɻ	j	ɰ			
Lateral approximant				l		ɭ	ʎ	ʟ			

Symbols to the right in a cell are voiced, to the left are voiceless. Shaded areas denote articulations judged impossible.

Lexical Phonology



- Phonology is the study of rules that govern the organization of sounds in a language (Phonemes → Syllables → Words)

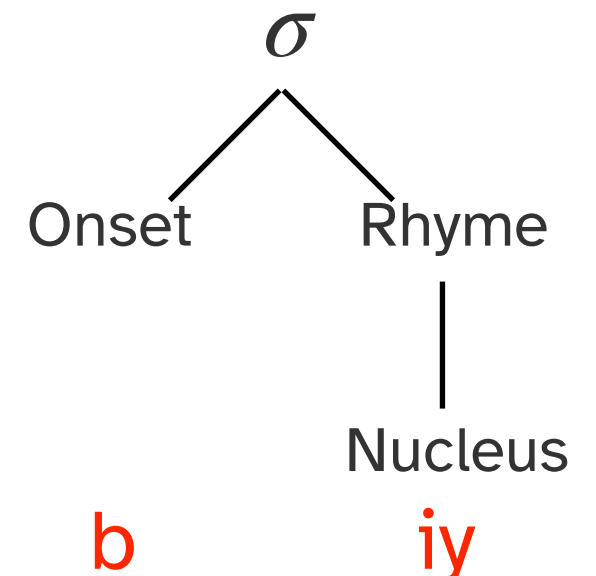
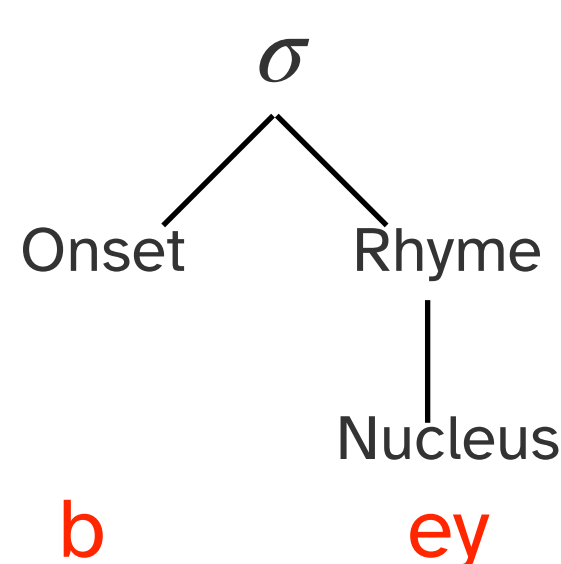
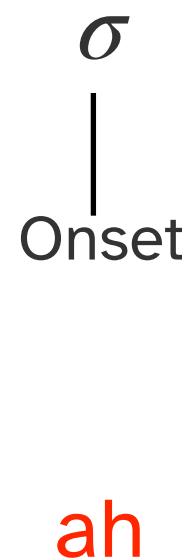
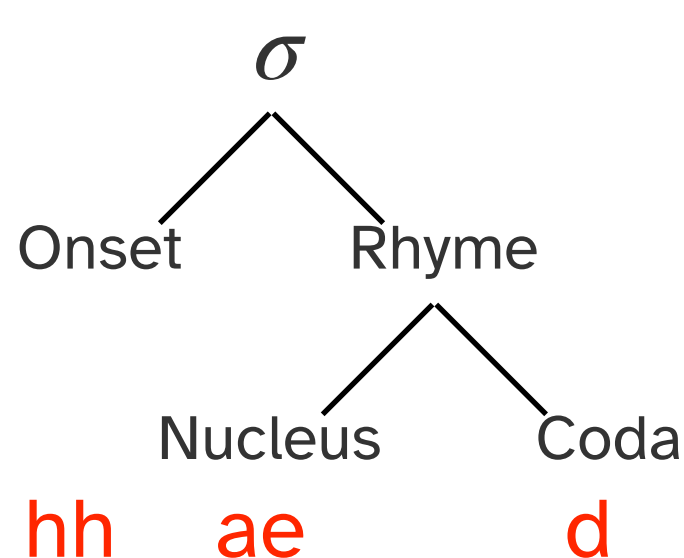
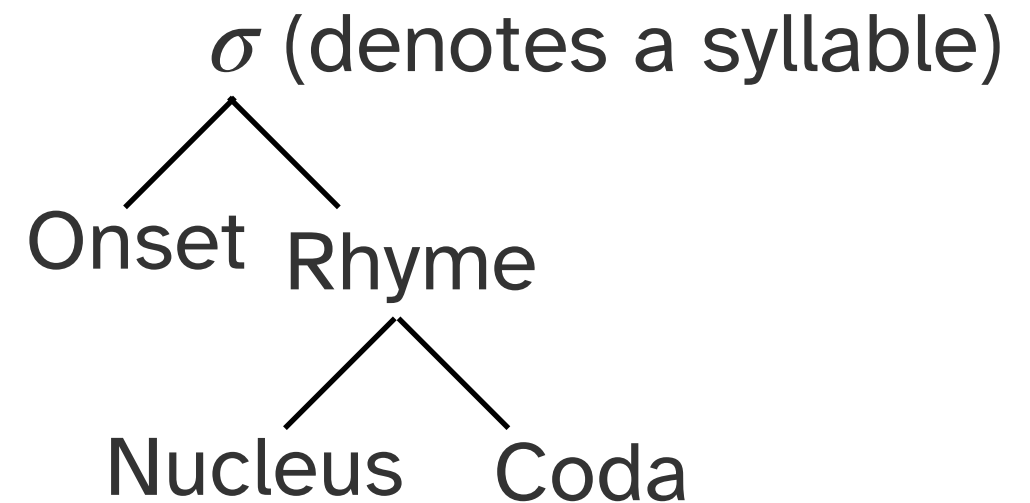
- English syllable structure: (C3)V(C4)

/strɛŋkθs/

- Hawaiian syllable structure: (C)V(C)

- Georgian syllable structure: (C8)V(C5)

გვბრდღვნიხ /'gvbrdɣvnis/



Letters to Sounds



- Pronunciation dictionaries (often made by linguists) give the syllables and phonemes within each word in vocabulary
 - CMU Phonetic Dictionary gives the syllabic and phonetic spellings for >110K words in English
 - ML based phonetizers are built on such phonetic dictionaries

Graphemes She just had a baby

IPA ʃiː dʒʌst hæd ə 'beɪbi

Arpabet sh iy jh ah s t h ae d ah b ey b iy

Arpabet is an ASCII friendly representation of IPA

Letters to Sounds



The CMU Pronouncing Dictionary

[query](#) | [phonemes](#) | [about](#) | | [Speech at](#)

- Look up the pronunciation for a word or phrase in CMUdict (version 0.7b)

Look It Up

☒ Show Lexical Stress

● BERKELEY

● B ER1 K L IY0 .

Berkeley

[Entry](#) [Discussion](#) [Citations](#)

English [\[edit\]](#)

Alternative forms [\[edit\]](#)

- (*surname*): [Barclay](#), [Barclift](#), [Barkley](#)

Etymology [\[edit\]](#)

From [Old English](#) *beorc* (“birch”) + *leah* (“meadow, leigh”). Equivalent to *birch* +

Pronunciation [\[edit\]](#)

- (*Received Pronunciation*) [IPA^{\(key\)}](#): (English town, barony) /'bɑːkli/, (American city) /'bɜːkli/
- (*General American*) [IPA^{\(key\)}](#): /'bɜːkli/
- Audio (*Southern England*):  0:01

Variety	Location	 edit
Mandarin	Beijing	/kuo ³⁵ /
	Harbin	/kuo ²¹³ /
	Tianjin	/kuo ⁴⁵ /
	Jinan	/kuə ⁵⁵ /
	Qingdao	/kuə ⁵⁵ /
	Zhengzhou	/kuo ²⁴ / /kuɛ ²⁴ /
	Xi'an	/kuei ²¹ /
	Xining	/kuɪ ⁴⁴ /
	Yinchuan	/kuə ¹³ /
	Lanzhou	/kuə ¹³ /
	Ürümqi	/kuɤ ²¹³ /
	Wuhan	/kuɤ ²¹³ /
	Chengdu	/kue ³¹ /
	Guiyang	/kuɛ ²¹ /
	Kunming	/ko ³¹ /
	Nanjing	/kueɿ ⁵ /
	Hefei	/kueɿ ⁵ /

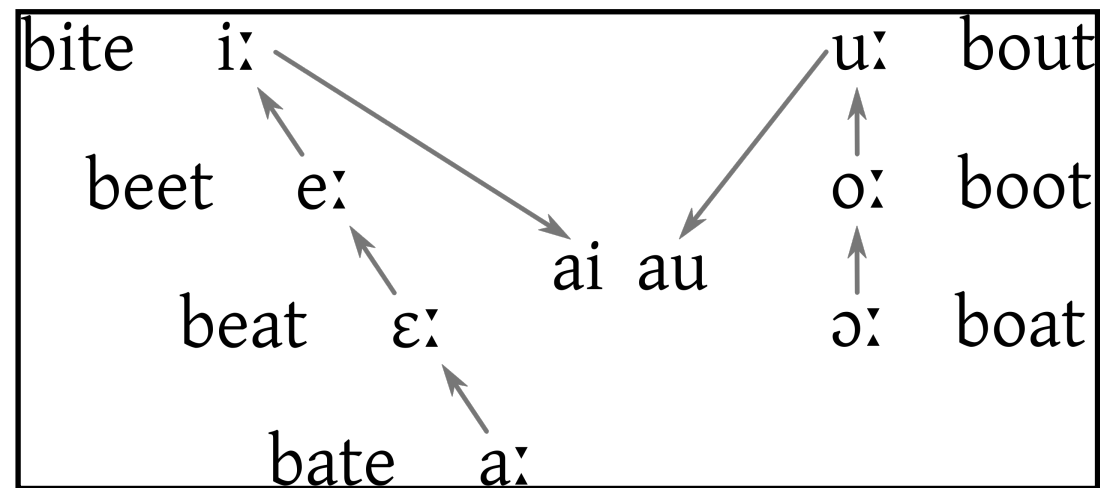


Wiktionary
The free dictionary

Sound Variation



- What might cause differences in pronunciation?
 - Dialectal differences
 - Native language when speaking a non-native language
 - Context — formality, noise levels, etc.
 - Language change over time



Goran tek-en, Wikipedia

Word	Vowel pronunciation	
	Late Middle English before the GVS	Modern English after the GVS
bite	[i:]	[aɪ]
meet	[e:]	[i:]
meat	[ɛ:]	
serene		
mate	[a:]	[eɪ]
out	[u:]	[aʊ]
boot	[o:]	[u:]
boat	[ɔ:]	[oʊ]
stone		

- We learn sounds before we are born!

Phonemes and Graphemes



- Words are composed of atomic units based on sound (for spoken languages)
- Sounds are a function of how we move our vocal tracts and mouth anatomy
- Languages have distinct sets of possible sounds (phonemic inventory)
- And “rules” governing which sound sequences are likely (syllable structure)

Linguistic Units

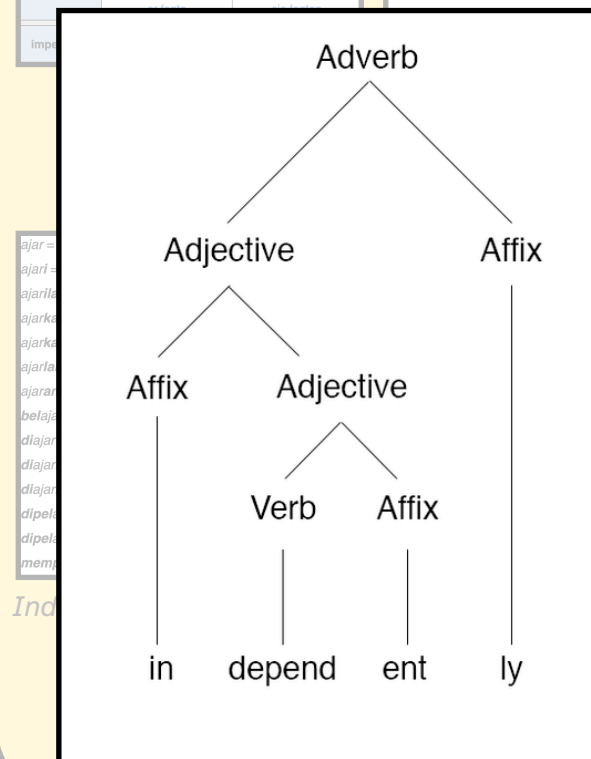


Morphemes
/Lexemes



ASL mouth morphemes
Large (CHA) and Small (OO)
learnhowtosign.com

present	ich fegge du fegst er fegt sie fegen	wir fegen ihr fegt sie fegen
preterite	ich fegte du fegtest er fegte sie fegten	wir fegten ihr fegtet sie fegten



Wikipedia (Annie Yang)

fǎn

反 — "anti-"

fǎn kǒng

- 反 恐 [反恐] — "anti-terror"

fǎn jiàojiànde

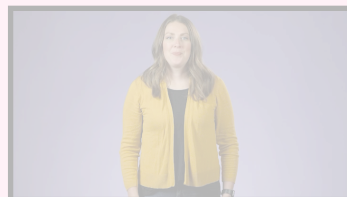
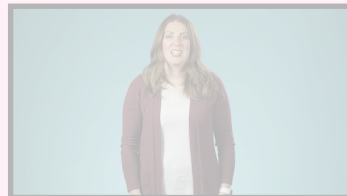
- 反 教 权的 [反教權的] — "anti-clerical"

fǎn fàxī

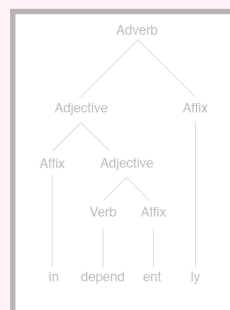
- 反 法西斯 [反法西斯] — "anti-fascist"

Wikipedia

Linguistic Units



ASL mouth morphemes
Large (CHA) and Small (OO)
learnhowtosign.com



Wikipedia (Annie Yang)

fǎn
反 — "anti-"
fǎn kǒng
• 反 恐 [反恐] — "anti-terror"
fǎn jiàoxué de
• 反 教 权 的 [反教权的] — "anti-clerical"
fǎn fǎxīst
• 反 法 西 斯 [反法西斯] — "anti-fascist"

Wikipedia

Morphemes
/Lexemes

Words



"language" in ASL
lifeprint.com

	indicative	
	singular	plural
present	ich feg	wir fegen
	du fegst	ihr fegt
	er fegt	sie fegen
preterite	ich fegte	wir fegten
	du fegtest	ihr fegtet
	er fegte	sie fegten
imperative	feg (du)	fegt (ihr)
	feg	fegt

conjugation of German
"fegen" (to sweep)
(Wiktionary)

ajar = to teach
ajari = to teach (imperative, locative)
ajarilah = to teach (jussive, locative)
ajarkan = to teach (imperative, causative/applicative)
ajarkanlah = to teach (jussive, causative/applicative)
ajarlah = to teach (jussive, active)
ajaran = teachings
belajar = to learn (intransitive, active)
diajar = to be taught (intransitive)
diajari = to be taught (transitive, locative)
diajarkan = to be taught (transitive, causative/applicative)
dipelajari = to be studied (locative)
dipelajarkan = to be studied (causative/applicative)
mempelajari = to study (locative)

Indonesian "ajar" (to teach)
with affixes
(Wikipedia)

Word Types and Tokens



The most natural and basic form of language use is dialogue: Every language user, including young children and illiterate adults, can hold a conversation, whereas reading, writing, preparing speeches and even listening to speeches are far from universal skills.

*from Pickering and Garrod 2004,
“A mechanistic psychology of dialogue”*

- Text data can be viewed as a sequence of words
- First step in building a language technology: building a function that maps from arbitrary text data to that sequence

```
['The', 'most', 'natural', 'and', 'basic', 'form', 'of', 'language',  
'use', 'is', 'dialogue', ':', 'Every', 'language', 'user', ',',  
'including', 'young', 'children', 'and', 'illiterate', 'adults', ',',  
'can', 'hold', 'a', 'conversation', ',', 'whereas', 'reading', ',',  
'writing', ',', 'preparing', 'speeches', 'and', 'even', 'listening',  
'to', 'speeches', 'are', 'far', 'from', 'universal', 'skills', '.']
```

tokenized text

Word Types and Tokens



The most natural and basic form of language use is dialogue: Every language user, including young children and illiterate adults, can hold a conversation, whereas reading, writing, preparing speeches and even listening to speeches are far from universal skills.

*from Pickering and Garrod 2004,
“A mechanistic psychology of dialogue”*

- Type-token distinction:
 - Type: a unique word in a text corpus
 - Token: an *instance* of a word type, appearing in a particular context

```
[ 'The', 'most', 'natural', 'and', 'basic', 'form', 'of', 'language',  
'use', 'is', 'dialogue', ':', 'Every', 'language', 'user', ',',  
'including', 'young', 'children', 'and', 'illiterate', 'adults', ',',  
'can', 'hold', 'a', 'conversation', ',', 'whereas', 'reading', ',',  
'writing', ',', 'preparing', 'speeches', 'and', 'even', 'listening',  
'to', 'speeches', 'are', 'far', 'from', 'universal', 'skills', '.']
```

tokenized text

Word Types and Tokens



The most natural and basic form of language use is dialogue: Every language user, including young children and illiterate adults, can hold a conversation, whereas reading, writing, preparing speeches and even listening to speeches are far from universal skills.

*from Pickering and Garrod 2004,
“A mechanistic psychology of dialogue”*

```
{ '.', ',', ':', 'Every', 'The', 'a', 'adults', 'and', 'are', 'basic',  
'can', 'children', 'conversation', 'dialogue', 'even', 'far', 'form',  
'from', 'hold', 'illiterate', 'including', 'is', 'language', 'listening',  
'most', 'natural', 'of', 'preparing', 'reading', 'skills', 'speeches',  
'to', 'universal', 'use', 'user', 'whereas', 'writing', 'young' }
```

wordtypes (vocabulary)



```
[ 'The', 'most', 'natural', 'and', 'basic', 'form', 'of', 'language',  
'use', 'is', 'dialogue', ':', 'Every', 'language', 'user', ',',  
'including', 'young', 'children', 'and', 'illiterate', 'adults', ',',  
'can', 'hold', 'a', 'conversation', ',', 'whereas', 'reading', ',',  
'writing', ',', 'preparing', 'speeches', 'and', 'even', 'listening',  
'to', 'speeches', 'are', 'far', 'from', 'universal', 'skills', '.' ]
```

tokenized text

Word Types and Tokens



The most natural and basic form of language use is dialogue: Every language user, including young children and illiterate adults, can hold a conversation, whereas reading, writing, preparing speeches and even listening to speeches are far from universal skills.

*from Pickering and Garrod 2004,
"A mechanistic psychology of dialogue"*

```
{ '\.', ',', ':', 'Every', 'The', 'a', 'adults', 'and', 'are', 'basic',  
'can', 'children', 'conversation', 'dialogue', 'even', 'far', 'form',  
'from', 'hold', 'illiterate', 'including', 'is', 'language', 'listening',  
'most', 'natural', 'of', 'preparing', 'reading', 'skills', 'speeches',  
'to', 'universal', 'use', 'user', 'whereas', 'writing', 'young' }
```

wordtypes (vocabulary)

instances (tokens) of wordtype `\.'`

```
[ 'The', 'most', 'natural', 'and', 'basic', 'form', 'of', 'language',  
'use', 'is', 'dialogue', ':', 'Every', 'language', 'user', ',',  
'including', 'young', 'children', 'and', 'illiterate', 'adults', ',',  
'can', 'hold', 'a', 'conversation', ',', 'whereas', 'reading', ',',  
'writing', ',', 'preparing', 'speeches', 'and', 'even', 'listening',  
'to', 'speeches', 'are', 'far', 'from', 'universal', 'skills', '\.' ]
```

tokenized text

Word Types and Tokens



The most natural and basic form of language use is dialogue: Every language user, including young children and illiterate adults, can hold a conversation, whereas reading, writing, preparing speeches and even listening to speeches are far from universal skills.

*from Pickering and Garrod 2004,
“A mechanistic psychology of dialogue”*

```
{ '\.', ',', ':', 'Every', 'The', 'a', 'adults', 'and', 'are', 'basic',  
'can', 'children', 'conversation', 'dialogue', 'even', 'far', 'form',  
'from', 'hold', 'illiterate', 'including', 'is', 'language', 'listening',  
'most', 'natural', 'of', 'preparing', 'reading', 'skills', 'speeches',  
'to', 'universal', 'use', 'user', 'whereas', 'writing', 'young' }
```

wordtypes (vocabulary)

instances (tokens) of wordtype `','`

```
[ 'The', 'most', 'natural', 'and', 'basic', 'form', 'of', 'language',  
'use', 'is', 'dialogue', ':', 'Every', 'language', 'user', ',',  
'including', 'young', 'children', 'and', 'illiterate', 'adults', ',',  
'can', 'hold', 'a', 'conversation', ',', 'whereas', 'reading', ',',  
'writing', ',', 'preparing', 'speeches', 'and', 'even', 'listening',  
'to', 'speeches', 'are', 'far', 'from', 'universal', 'skills', '.']
```

tokenized text

Word Types and Tokens



The most natural and basic form of language use is dialogue: Every language user, including young children and illiterate adults, can hold a conversation, whereas reading, writing, preparing speeches and even listening to speeches are far from universal skills.

*from Pickering and Garrod 2004,
“A mechanistic psychology of dialogue”*

```
{ '.', ',', ':', 'Every', 'The', 'a', 'adults', 'and', 'are', 'basic',  
'can', 'children', 'conversation', 'dialogue', 'even', 'far', 'form',  
'from', 'hold', 'illiterate', 'including', 'is', 'language', 'listening',  
'most', 'natural', 'of', 'preparing', 'reading', 'skills', 'speeches',  
'to', 'universal', 'use', 'user', 'whereas', 'writing', 'young' }
```

wordtypes (vocabulary)

instances (tokens) of wordtype 'Every'

```
[ 'The', 'most', 'natural', 'and', 'basic', 'form', 'of', 'language',  
'use', 'is', 'dialogue', ':', 'Every', 'language', 'user', ',',  
'including', 'young', 'children', 'and', 'illiterate', 'adults', ',',  
'can', 'hold', 'a', 'conversation', ',', 'whereas', 'reading', ',',  
'writing', ',', 'preparing', 'speeches', 'and', 'even', 'listening',  
'to', 'speeches', 'are', 'far', 'from', 'universal', 'skills', '.' ]
```

tokenized text

Word Types and Tokens



The most natural and basic form of language use is dialogue: Every language user, including young children and illiterate adults, can hold a conversation, whereas reading, writing, preparing speeches and even listening to speeches are far from universal skills.

*from Pickering and Garrod 2004,
“A mechanistic psychology of dialogue”*

```
{ '.', ',', ':', 'Every', 'The', 'a', 'adults', 'and', 'are', 'basic',  
'can', 'children', 'conversation', 'dialogue', 'even', 'far', 'form',  
'from', 'hold', 'illiterate', 'including', 'is', 'language', 'listening',  
'most', 'natural', 'of', 'preparing', 'reading', 'skills', 'speeches',  
'to', 'universal', 'use', 'user', 'whereas', 'writing', 'young' }
```

wordtypes (vocabulary)

instances (tokens) of wordtype 'and'

```
[ 'The', 'most', 'natural', 'and', 'basic', 'form', 'of', 'language',  
'use', 'is', 'dialogue', ':', 'Every', 'language', 'user', ',',  
'including', 'young', 'children', 'and', 'illiterate', 'adults', ',',  
'can', 'hold', 'a', 'conversation', ',', 'whereas', 'reading', ',',  
'writing', ',', 'preparing', 'speeches', 'and', 'even', 'listening',  
'to', 'speeches', 'are', 'far', 'from', 'universal', 'skills', '.' ]
```

tokenized text

Word Types and Tokens



The most natural and basic form of language use is dialogue: Every language user, including young children and illiterate adults, can hold a conversation, whereas reading, writing, preparing speeches and even listening to speeches are far from universal skills.

*from Pickering and Garrod 2004,
“A mechanistic psychology of dialogue”*

```
{ '.', ',', ':', 'Every', 'The', 'a', 'adults', 'and', 'are', 'basic',  
'can', 'children', 'conversation', 'dialogue', 'even', 'far', 'form',  
'from', 'hold', 'illiterate', 'including', 'is', 'language', 'listening',  
'most', 'natural', 'of', 'preparing', 'reading', 'skills', 'speeches',  
'to', 'universal', 'use', 'user', 'whereas', 'writing', 'young' }
```

wordtypes (vocabulary)

instances (tokens) of wordtype 'language'

```
[ 'The', 'most', 'natural', 'and', 'basic', 'form', 'of', 'language',  
'use', 'is', 'dialogue', ':', 'Every', 'language', 'user', ',',  
'including', 'young', 'children', 'and', 'illiterate', 'adults', ',',  
'can', 'hold', 'a', 'conversation', ',', 'whereas', 'reading', ',',  
'writing', ',', 'preparing', 'speeches', 'and', 'even', 'listening',  
'to', 'speeches', 'are', 'far', 'from', 'universal', 'skills', '.' ]
```

tokenized text

Word Types and Tokens



The most natural and basic form of language use is dialogue: Every language user, including young children and illiterate adults, can hold a conversation, whereas reading, writing, preparing speeches and even listening to speeches are far from universal skills.

*from Pickering and Garrod 2004,
“A mechanistic psychology of dialogue”*

```
vocabulary =  
['.', ',', ':', 'Every', 'The', 'a', 'adults', 'and', 'are', 'basic',  
'can', 'children', 'conversation', 'dialogue', 'even', 'far', 'form',  
'from', 'hold', 'illiterate', 'including', 'is', 'language', 'listening',  
'most', 'natural', 'of', 'preparing', 'reading', 'skills', 'speeches',  
'to', 'universal', 'use', 'user', 'whereas', 'writing', 'young']
```

vocabulary as a look-up table

```
tokenized_text = ['The', 'most', ... , 'skills', '.']  
tokenized_indices = [vocabulary.index(token) for token in tokenized_text]  
  
tokenized_indices: [4, 24, 25, 7, 9, 16, 26, 22, 33, 21, 13, 2, 3, 22,  
34, 1, 20, 37, 11, 7, 19, 6, 1, 10, 18, 5, 12, 1, 35, 28, 1, 36, 1, 27,  
30, 7, 14, 23, 31, 30, 8, 15, 17, 32, 29, 0]
```

text as a sequence of wordtype indices

Tokenize on Spaces



The most natural and basic form of language use is dialogue: Every language user, including young children and illiterate adults, can hold a conversation, whereas reading, writing, preparing speeches and even listening to speeches are far from universal skills.

*from Pickering and Garrod 2004,
“A mechanistic psychology of dialogue”*

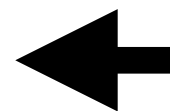
- Simplest tokenizer (for English): splitting on spaces

```
tokenized = s.split(' ')
```

```
['The', 'most', 'natural', 'and', 'basic', 'form', 'of', 'language',  
'use', 'is', 'dialogue:', 'Every', 'language', 'user,', 'including',  
'young', 'children', 'and', 'illiterate', 'adults,', 'can', 'hold', 'a',  
'conversation,', 'whereas', 'reading,', 'writing,', 'preparing',  
'speeches', 'and', 'even', 'listening', 'to', 'speeches', 'are', 'far',  
'from', 'universal', 'skills.']
```

- But this gets us some weird wordtypes:

```
'dialogue:'  
'user,'  
'skills.'
```



Not really words different from
dialogue, user, skills

Rule-Based Tokenization



The most natural and basic form of language use is dialogue: Every language user, including young children and illiterate adults, can hold a conversation, whereas reading, writing, preparing speeches and even listening to speeches are far from universal skills.

*from Pickering and Garrod 2004,
“A mechanistic psychology of dialogue”*

- nltk tokenizers, with special rules for punctuation

```
import nltk  
tokenized = nltk.word_tokenize(s)
```

```
['The', 'most', 'natural', 'and', 'basic', 'form', 'of', 'language',  
'use', 'is', 'dialogue', ':', 'Every', 'language', 'user', ',',  
'including', 'young', 'children', 'and', 'illiterate', 'adults', ',',  
'can', 'hold', 'a', 'conversation', ',', 'whereas', 'reading', ',',  
'writing', ',', 'preparing', 'speeches', 'and', 'even', 'listening',  
'to', 'speeches', 'are', 'far', 'from', 'universal', 'skills', '.']
```

- But this still loses similarity between wordtypes

```
'skills'  
'reading'  
'speeches'
```



Lexically similar to, but
morphologically distinct from
skill, read, speech

Rule-Based Tokenization



The most natural and basic form of language use is dialogue: Every language user, including young children and illiterate adults, can hold a conversation, whereas reading, writing, preparing speeches and even listening to speeches are far from universal skills.

*from Pickering and Garrod 2004,
“A mechanistic psychology of dialogue”*

- nltk tokenizers, with special rules for punctuation

```
import nltk  
tokenized = nltk.word_tokenize(s)
```

```
['The', 'most', 'natural', 'and', 'basic', 'form', 'of', 'language',  
'use', 'is', 'dialogue', ':', 'Every', 'language', 'user', ',',  
'including', 'young', 'children', 'and', 'illiterate', 'adults', ',',  
'can', 'hold', 'a', 'conversation', ',', 'whereas', 'reading', ',',  
'writing', ',', 'preparing', 'speeches', 'and', 'even', 'listening',  
'to', 'speeches', 'are', 'far', 'from', 'universal', 'skills', '.']
```

- And doesn't work for all languages

โดยที่การยอมรับนับถือเกียรติศักดิ์ประจำตัว และสิทธิเท่าเทียมกันและโอนมิได้ของ
บรรดา สมาชิก ทั้ง หลายแห่งครอบครัว มนุษย์เป็นหลักรวมเหตุแห่งอิสรภาพ ความ
ยุติธรรม และสันติภาพในโลก

Rule-Based Tokenization



The most natural and basic form of language use is dialogue: Every language user, including young children and illiterate adults, can hold a conversation, whereas reading, writing, preparing speeches and even listening to speeches are far from universal skills.

*from Pickering and Garrod 2004,
“A mechanistic psychology of dialogue”*

- nltk tokenizers, with special rules for punctuation

```
import nltk
tokenized = nltk.word_tokenize(s)
```

```
['The', 'most', 'natural', 'and', 'basic', 'form', 'of', 'language',  
'use', 'is', 'dialogue', ':', 'Every', 'language', 'user', ',',  
'including', 'young', 'children', 'and', 'illiterate', 'adults', ',',  
'can', 'hold', 'a', 'conversation', ',', 'whereas', 'reading', ',',  
'writing', ',', 'preparing', 'speeches', 'and', 'even', 'listening',  
'to', 'speeches', 'are', 'far', 'from', 'universal', 'skills', '.']
```

- Once you’ve “trained” your tokenizer, you’re stuck with it

```
vocab = ['.', ',', ':', 'Every', ... , 'writing', 'young']  
vocab.index('ChatGPT') → not found!
```

Rule-Based Tokenization



The most natural and basic form of language use is dialogue: Every language user, including young children and illiterate adults, can hold a conversation, whereas reading, writing, preparing speeches and even listening to speeches are far from universal skills.

*from Pickering and Garrod 2004,
“A mechanistic psychology of dialogue”*

- nltk tokenizers, with special rules for punctuation

```
import nltk
tokenized = nltk.word_tokenize(s)
```

```
['The', 'most', 'natural', 'and', 'basic', 'form', 'of', 'language',  
'use', 'is', 'dialogue', ':', 'Every', 'language', 'user', ',',  
'including', 'young', 'children', 'and', 'illiterate', 'adults', ',',  
'can', 'hold', 'a', 'conversation', ',', 'whereas', 'reading', ',',  
'writing', ',', 'preparing', 'speeches', 'and', 'even', 'listening',  
'to', 'speeches', 'are', 'far', 'from', 'universal', 'skills', '.']
```

- Once you’ve “trained” your tokenizer, you’re stuck with it

```
vocab = ['.', ',', ':', 'Every', ... , 'writing', 'young', '<UNK>']  
tokenized_indices =  
    [vocabulary.index(token) for token in tokenized_text  
     if token in vocabulary  
     else vocabulary.index('<UNK>')]
```

↑
**add a special token
for unknown words**

A stylized illustration of a building with a triangular roof and arched windows. The building is composed of a light blue base with several arched windows, and a white triangular roof with a dark blue outline. The background is a solid light blue.

from Pickering and Garrod 2004,
"A mechanistic psychology of dialogue"

- [illegible]

Bytes as Wordtypes



The most natural and basic form of language use is dialogue: Every language user, including young children and illiterate adults, can hold a conversation, whereas reading, writing, preparing speeches and even listening to speeches are far from universal skills.

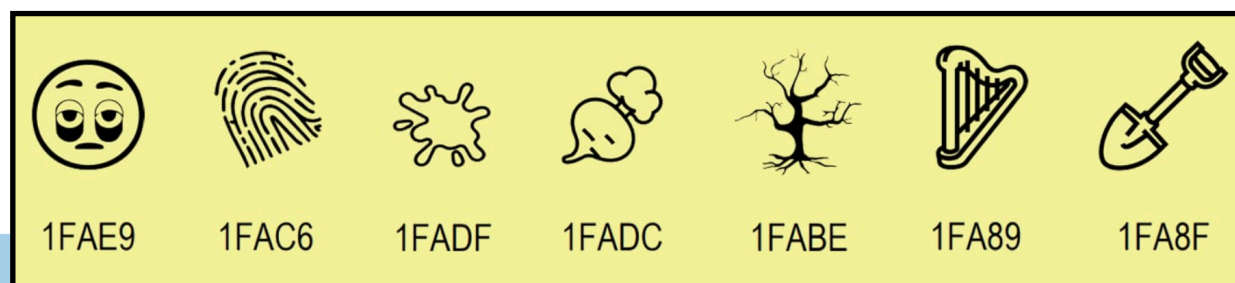
*from Pickering and Garrod 2004,
“A mechanistic psychology of dialogue”*

- Strings are sequences of characters (bytes)!

```
tokenized = s.encode()
```

```
54 68 65 20 6d 6f 73 74 20 6e 61 74 75 72 61 6c 20 61 6e 64 20 62 61 73 69 63 20 66 6f 72 6d 20 6f 66 20
6c 61 6e 67 75 61 67 65 20 75 73 65 20 69 73 20 64 69 61 6c 6f 67 75 65 3a 20 45 76 65 72 79 20 6c 61 6e
67 75 61 67 65 20 75 73 65 72 2c 20 69 6e 63 6c 75 64 69 6e 67 20 79 6f 75 6e 67 20 63 68 69 6c 64 72 65
6e 20 61 6e 64 20 69 6c 6c 69 74 65 72 61 74 65 20 61 64 75 6c 74 73 2c 20 63 61 6e 20 68 6f 6c 64 20 61
20 63 6f 6e 76 65 72 73 61 74 69 6f 6e 2c 20 77 68 65 72 65 61 73 20 72 65 61 64 69 6e 67 2c 20 77 72 69
74 69 6e 67 2c 20 70 72 65 70 61 72 69 6e 67 20 73 70 65 65 63 68 65 73 20 61 6e 64 20 65 76 65 6e 20 6c
69 73 74 65 6e 69 6e 67 20 74 6f 20 73 70 65 65 63 68 65 73 20 61 72 65 20 66 61 72 20 66 72 6f 6d 20 75
6e 69 76 65 72 73 61 6c 20 73 6b 69 6c 6c 73 2e
```

- And we don't have any unknown wordtypes
(until the Unicode Consortium adds new emoji)



Bytes as Wordtypes



The most natural and basic form of language use is dialogue: Every language user, including young children and illiterate adults, can hold a conversation, whereas reading, writing, preparing speeches and even listening to speeches are far from universal skills.

*from Pickering and Garrod 2004,
“A mechanistic psychology of dialogue”*

- Strings are sequences of characters (bytes)!

```
tokenized = s.encode()
```

```
54 68 65 20 6d 6f 73 74 20 6e 61 74 75 72 61 6c 20 61 6e 64 20 62 61 73 69 63 20 66 6f 72 6d 20 6f 66 20
6c 61 6e 67 75 61 67 65 20 75 73 65 20 69 73 20 64 69 61 6c 6f 67 75 65 3a 20 45 76 65 72 79 20 6c 61 6e
67 75 61 67 65 20 75 73 65 72 2c 20 69 6e 63 6c 75 64 69 6e 67 20 79 6f 75 6e 67 20 63 68 69 6c 64 72 65
6e 20 61 6e 64 20 69 6c 6c 69 74 65 72 61 74 65 20 61 64 75 6c 74 73 2c 20 63 61 6e 20 68 6f 6c 64 20 61
20 63 6f 6e 76 65 72 73 61 74 69 6f 6e 2c 20 77 68 65 72 65 61 73 20 72 65 61 64 69 6e 67 2c 20 77 72 69
74 69 6e 67 2c 20 70 72 65 70 61 72 69 6e 67 20 73 70 65 65 63 68 65 73 20 61 6e 64 20 65 76 65 6e 20 6c
69 73 74 65 6e 69 6e 67 20 74 6f 20 73 70 65 65 63 68 65 73 20 61 72 65 20 66 61 72 20 66 72 6f 6d 20 75
6e 69 76 65 72 73 61 6c 20 73 6b 69 6c 6c 73 2e
```

- But: individual characters are not meaningful

```
[ '\.', ',,', ':', 'Every', ... , 'user', 'whereas', 'writing', 'young' ]
```

VS.

```
[ '\.', ',,', ':', 'A', 'B', 'C', ... , 'a', 'b', 'c', ... , '7', '8', '9' ]
```

Bytes as Wordtypes



The most natural and basic form of language use is dialogue: Every language user, including young children and illiterate adults, can hold a conversation, whereas reading, writing, preparing speeches and even listening to speeches are far from universal skills.

*from Pickering and Garrod 2004,
“A mechanistic psychology of dialogue”*

- Strings are sequences of characters (bytes)!

```
tokenized = s.encode()
```

```
54 68 65 20 6d 6f 73 74 20 6e 61 74 75 72 61 6c 20 61 6e 64 20 62 61 73 69 63 20 66 6f 72 6d 20 6f 66 20
6c 61 6e 67 75 61 67 65 20 75 73 65 20 69 73 20 64 69 61 6c 6f 67 75 65 3a 20 45 76 65 72 79 20 6c 61 6e
67 75 61 67 65 20 75 73 65 72 2c 20 69 6e 63 6c 75 64 69 6e 67 20 79 6f 75 6e 67 20 63 68 69 6c 64 72 65
6e 20 61 6e 64 20 69 6c 6c 69 74 65 72 61 74 65 20 61 64 75 6c 74 73 2c 20 63 61 6e 20 68 6f 6c 64 20 61
20 63 6f 6e 76 65 72 73 61 74 69 6f 6e 2c 20 77 68 65 72 65 61 73 20 72 65 61 64 69 6e 67 2c 20 77 72 69
74 69 6e 67 2c 20 70 72 65 70 61 72 69 6e 67 20 73 70 65 65 63 68 65 73 20 61 6e 64 20 65 76 65 6e 20 6c
69 73 74 65 6e 69 6e 67 20 74 6f 20 73 70 65 65 63 68 65 73 20 61 72 65 20 66 61 72 20 66 72 6f 6d 20 75
6e 69 76 65 72 73 61 6c 20 73 6b 69 6c 6c 73 2e
```

- It's now the ML model's job to learn to compose words from scratch

Bytes as Wordtypes



The most natural and basic form of language use is dialogue: Every language user, including young children and illiterate adults, can hold a conversation, whereas reading, writing, preparing speeches and even listening to speeches are far from universal skills.

*from Pickering and Garrod 2004,
“A mechanistic psychology of dialogue”*

- Strings are sequences of characters (bytes)!

```
tokenized = s.encode()
```

```
54 68 65 20 6d 6f 73 74 20 6e 61 74 75 72 61 6c 20 61 6e 64 20 62 61 73 69 63 20 66 6f 72 6d 20 6f 66 20
6c 61 6e 67 75 61 67 65 20 75 73 65 20 69 73 20 64 69 61 6c 6f 67 75 65 3a 20 45 76 65 72 79 20 6c 61 6e
67 75 61 67 65 20 75 73 65 72 2c 20 69 6e 63 6c 75 64 69 6e 67 20 79 6f 75 6e 67 20 63 68 69 6c 64 72 65
6e 20 61 6e 64 20 69 6c 6c 69 74 65 72 61 74 65 20 61 64 75 6c 74 73 2c 20 63 61 6e 20 68 6f 6c 64 20 61
20 63 6f 6e 76 65 72 73 61 74 69 6f 6e 2c 20 77 68 65 72 65 61 73 20 72 65 61 64 69 6e 67 2c 20 77 72 69
74 69 6e 67 2c 20 70 72 65 70 61 72 69 6e 67 20 73 70 65 65 63 68 65 73 20 61 6e 64 20 65 76 65 6e 20 6c
69 73 74 65 6e 69 6e 67 20 74 6f 20 73 70 65 65 63 68 65 73 20 61 72 65 20 66 61 72 20 66 72 6f 6d 20 75
6e 69 76 65 72 73 61 6c 20 73 6b 69 6c 6c 73 2e
```

- But: input sequences are much longer

```
`language`
```

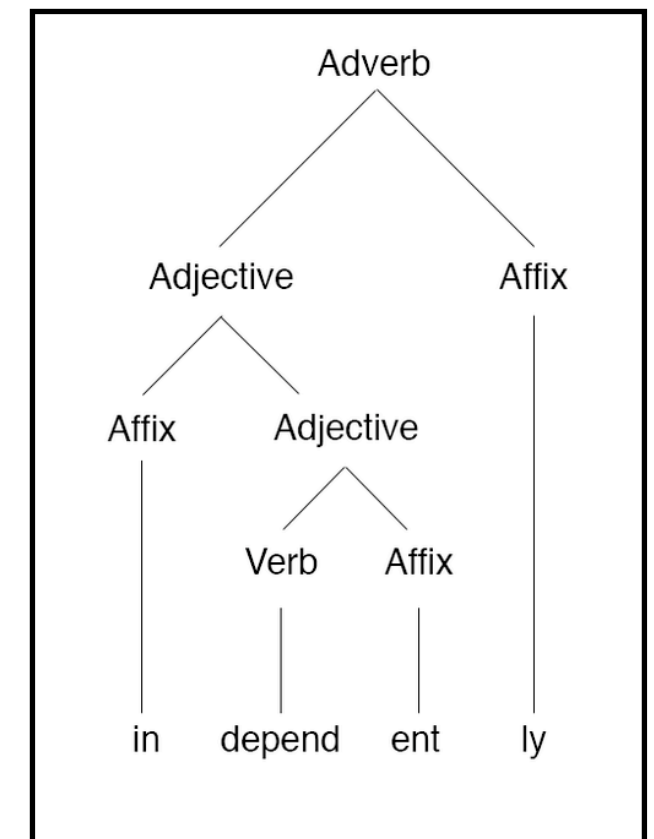
VS.

```
`l`, `a`, `n`, `g`, `u`, `a`, `g`, `e`
```

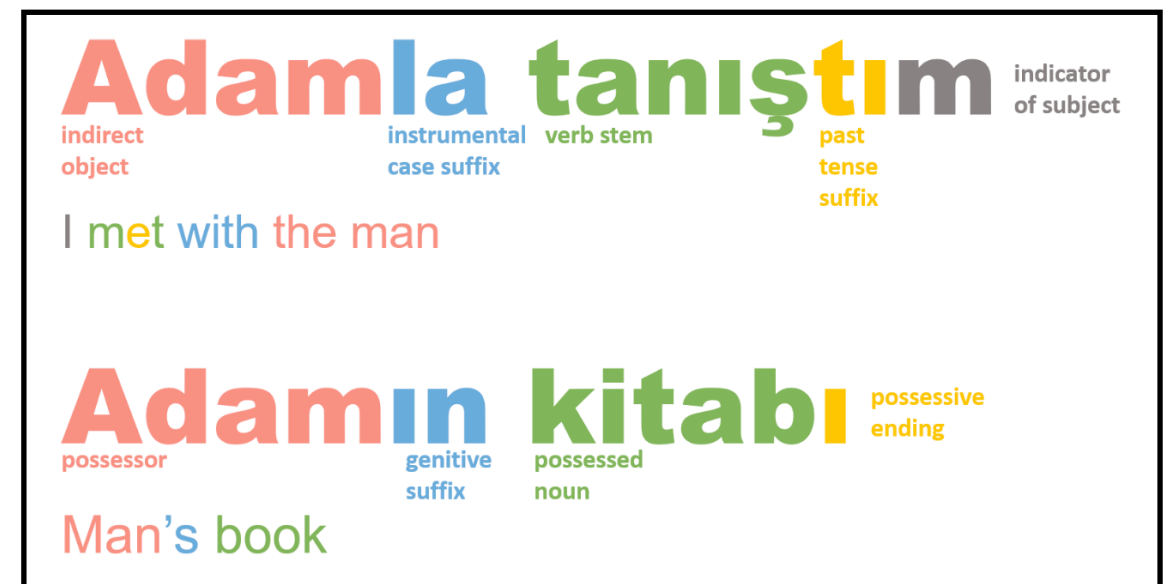

A Compromise: Subword Tokenization



- Main principle: words are (often) composed of subparts (morphemes)
- Our vocabulary should have entries for frequent words kept whole, because we have a lot of data about those words
- But it should also have entries for parts of less-frequent words, so our ML models can learn how to compose parts of words into whole words (especially unfamiliar words!)



Wikipedia (Annie Yang)



Byte Pair Encoding



- Modern standard for building a tokenizer
- Inputs: collection of texts and target vocabulary size
- Initial vocabulary is the set of all bytes (characters) across the texts
- Until the target vocabulary size is reached, repeat the following:
 - Tokenize all of the texts using the current vocabulary
 - Find the most common bigram in the tokenized texts, then add it to the vocabulary as a new wordtype

Byte Pair Encoding



Documents + frequencies: ('hug', 10), ('pug', 5), ('pun', 12), ('bun', 4), ('hugs', 5)

vocabulary

('h', 'u', 'g', 'p', 'n', 'b', 's')

Byte Pair Encoding



Documents + frequencies: ('hug', 10), ('pug', 5), ('pun', 12), ('bun', 4), ('hugs', 5)

vocabulary

tokenized texts

('h', 'u', 'g', 'p', 'n', 'b', 's') → ('h' 'u' 'g', 10), ('p' 'u' 'g', 5), ('p' 'u' 'n', 12), ('b' 'u' 'n', 4), ('h' 'u' 'g' 's', 5)

bigrams + frequencies

'h' 'u' 15

Byte Pair Encoding



Documents + frequencies: ('hug', 10), ('pug', 5), ('pun', 12), ('bun', 4), ('hugs', 5)

vocabulary

tokenized texts

('h', 'u', 'g', 'p', 'n', 'b', 's') → ('h' 'u' 'g', 10), ('p' 'u' 'g', 5), ('p' 'u' 'n', 12), ('b' 'u' 'n', 4), ('h' 'u' 'g' 's', 5)

bigrams + frequencies

'h' 'u' 15

'u' 'g' 20

Byte Pair Encoding



Documents + frequencies: ('hug', 10), ('pug', 5), ('pun', 12), ('bun', 4), ('hugs', 5)

vocabulary

tokenized texts

('h', 'u', 'g', 'p', 'n', 'b', 's') → ('h' 'u' 'g', 10), ('p' 'u' 'g', 5), ('p' 'u' 'n', 12), ('b' 'u' 'n', 4), ('h' 'u' 'g' 's', 5)

bigrams + frequencies

'h' 'u' 15

'u' 'g' 20

'p' 'u' 17

Byte Pair Encoding



Documents + frequencies: ('hug', 10), ('pug', 5), ('pun', 12), ('bun', 4), ('hugs', 5)

vocabulary

tokenized texts

('h', 'u', 'g', 'p', 'n', 'b', 's') → ('h' 'u' 'g', 10), ('p' 'u' 'g', 5), ('p' 'u' 'n', 12), ('b' 'u' 'n', 4), ('h' 'u' 'g' 's', 5)

bigrams + frequencies

'h' 'u' 15

'u' 'g' 20

'p' 'u' 17

'u' 'n' 16

Byte Pair Encoding



Documents + frequencies: ('hug', 10), ('pug', 5), ('pun', 12), ('bun', 4), ('hugs', 5)

vocabulary

tokenized texts

('h', 'u', 'g', 'p', 'n', 'b', 's') → ('h' 'u' 'g', 10), ('p' 'u' 'g', 5), ('p' 'u' 'n', 12), ('b' 'u' 'n', 4), ('h' 'u' 'g' 's', 5)

bigrams + frequencies

'h'	'u'	15
'u'	'g'	20
'p'	'u'	17
'u'	'n'	16
'b'	'u'	4

Byte Pair Encoding



Documents + frequencies: ('hug', 10), ('pug', 5), ('pun', 12), ('bun', 4), ('hugs', 5)

vocabulary

tokenized texts

('h', 'u', 'g', 'p', 'n', 'b', 's') → ('h' 'u' 'g', 10), ('p' 'u' 'g', 5), ('p' 'u' 'n', 12), ('b' 'u' 'n', 4), ('h' 'u' 'g' 's', 5)

bigrams + frequencies

'h'	'u'	15
'u'	'g'	20
'p'	'u'	17
'u'	'n'	16
'b'	'u'	4
'g'	's'	5

Byte Pair Encoding



Documents + frequencies: ('hug', 10), ('pug', 5), ('pun', 12), ('bun', 4), ('hugs', 5)

vocabulary

tokenized texts

('h', 'u', 'g', 'p', 'n', 'b', 's') → ('h' 'u' 'g', 10), ('p' 'u' 'g', 5), ('p' 'u' 'n', 12), ('b' 'u' 'n', 4), ('h' 'u' 'g' 's', 5)

bigrams + frequencies

'h'	'u'	15
'u'	'g'	20
'p'	'u'	17
'u'	'n'	16
'b'	'u'	4
'g'	's'	5

most frequent bigram;
add to vocabulary

Byte Pair Encoding



Documents + frequencies: ('hug', 10), ('pug', 5), ('pun', 12), ('bun', 4), ('hugs', 5)

vocabulary

tokenized texts

('h', 'u', 'g', 'p', 'n', 'b', 's') → ('h' 'u' 'g', 10), ('p' 'u' 'g', 5), ('p' 'u' 'n', 12), ('b' 'u' 'n', 4), ('h' 'u' 'g' 's', 5)
(('h', 'u', 'g', 'p', 'n', 'b', 's', 'ug')) → ('h' 'ug', 10), ('p' 'ug', 5), ('p' 'u' 'n', 12), ('b' 'u' 'n', 4), ('h' 'ug' 's', 5)

Byte Pair Encoding



Documents + frequencies: ('hug', 10), ('pug', 5), ('pun', 12), ('bun', 4), ('hugs', 5)

vocabulary

tokenized texts

('h', 'u', 'g', 'p', 'n', 'b', 's') → ('h' 'u' 'g', 10), ('p' 'u' 'g', 5), ('p' 'u' 'n', 12), ('b' 'u' 'n', 4), ('h' 'u' 'g' 's', 5)
(('h', 'u', 'g', 'p', 'n', 'b', 's', 'ug')) → ('h' 'ug', 10), ('p' 'ug', 5), ('p' 'u' 'n', 12), ('b' 'u' 'n', 4), ('h' 'ug' 's', 5)

Byte Pair Encoding



Documents + frequencies: ('hug', 10), ('pug', 5), ('pun', 12), ('bun', 4), ('hugs', 5)

vocabulary

tokenized texts

('h', 'u', 'g', 'p', 'n', 'b', 's') → ('h' 'u' 'g', 10), ('p' 'u' 'g', 5), ('p' 'u' 'n', 12), ('b' 'u' 'n', 4), ('h' 'u' 'g' 's', 5)
('h', 'u', 'g', 'p', 'n', 'b', 's', 'ug') → ('h' 'ug', 10), ('p' 'ug', 5), ('p' 'u' 'n', 12), ('b' 'u' 'n', 4), ('h' 'ug' 's', 5)
('h', 'u', 'g', 'p', 'n', 'b', 's', 'ug', 'un') → ('h' 'ug', 10), ('p' 'ug', 5), ('p' 'un', 12), ('b' 'un', 4), ('h' 'ug' 's', 5)

Byte Pair Encoding



Documents + frequencies: ('hug', 10), ('pug', 5), ('pun', 12), ('bun', 4), ('hugs', 5)

vocabulary

tokenized texts

('h', 'u', 'g', 'p', 'n', 'b', 's') → ('h' 'u' 'g', 10), ('p' 'u' 'g', 5), ('p' 'u' 'n', 12), ('b' 'u' 'n', 4), ('h' 'u' 'g' 's', 5)
('h', 'u', 'g', 'p', 'n', 'b', 's', 'ug') → ('h' 'ug', 10), ('p' 'ug', 5), ('p' 'u' 'n', 12), ('b' 'u' 'n', 4), ('h' 'ug' 's', 5)
('h', 'u', 'g', 'p', 'n', 'b', 's', 'ug', 'un') → ('h' 'ug', 10), ('p' 'ug', 5), ('p' 'un', 12), ('b' 'un', 4), ('h' 'ug' 's', 5)
('h', 'u', 'g', 'p', 'n', 'b', 's', 'ug', 'un', 'hug') → ('hug', 10), ('p' 'ug', 5), ('p' 'un', 12), ('b' 'un', 4), ('hug' 's', 5)

New word: “puns”

'p' 'u' 'n' 's' → 'p' 'un' 's'



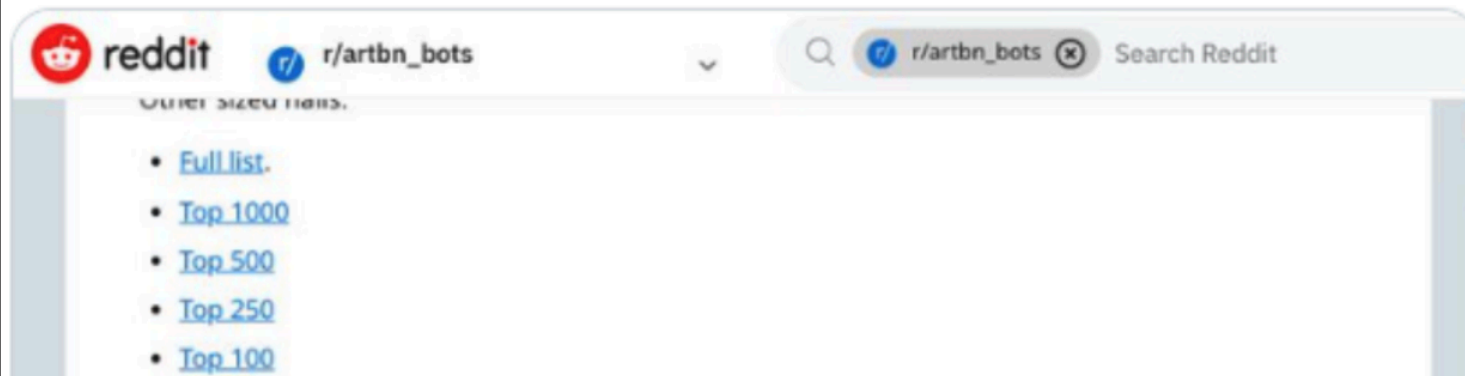
Vocabularies



Matthew Watkins
@SoC_trilogy

...

I've just found out that several of the anomalous GPT tokens ("TheNitromeFan", " SolidGoldMagikarp", " davidjl", " Smartstocks", " RandomRedditorWithNo",) are handles of people who are (competitively? collaboratively?) counting to infinity on a Reddit forum. I kid you not.

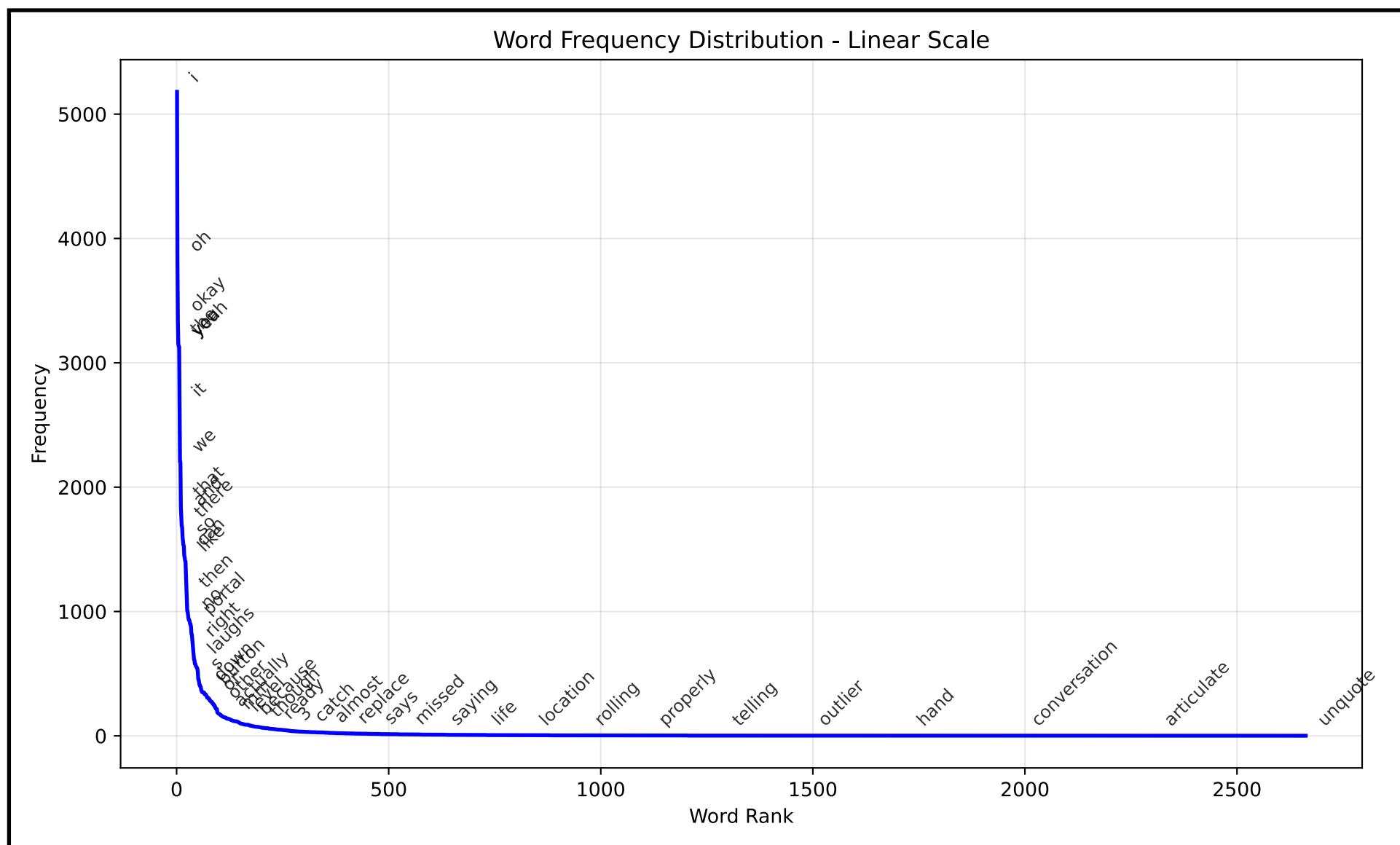


Rank	User	Counts
1	/u/davidjl123	163477
2	/u/Smartstocks	113829
3	/u/atomicimploder	103178
4	/u/TheNitromeFan	84581
5	/u/SolidGoldMagikarp	65753
6	/u/RandomRedditorWithNo	63434
7	/u/rideride	59024
8	/u/Mooraell	57785
9	/u/Removedpixel	55080
10	/u/Adinida	48415
11	/u/rschaosid	47132

Vocabularies



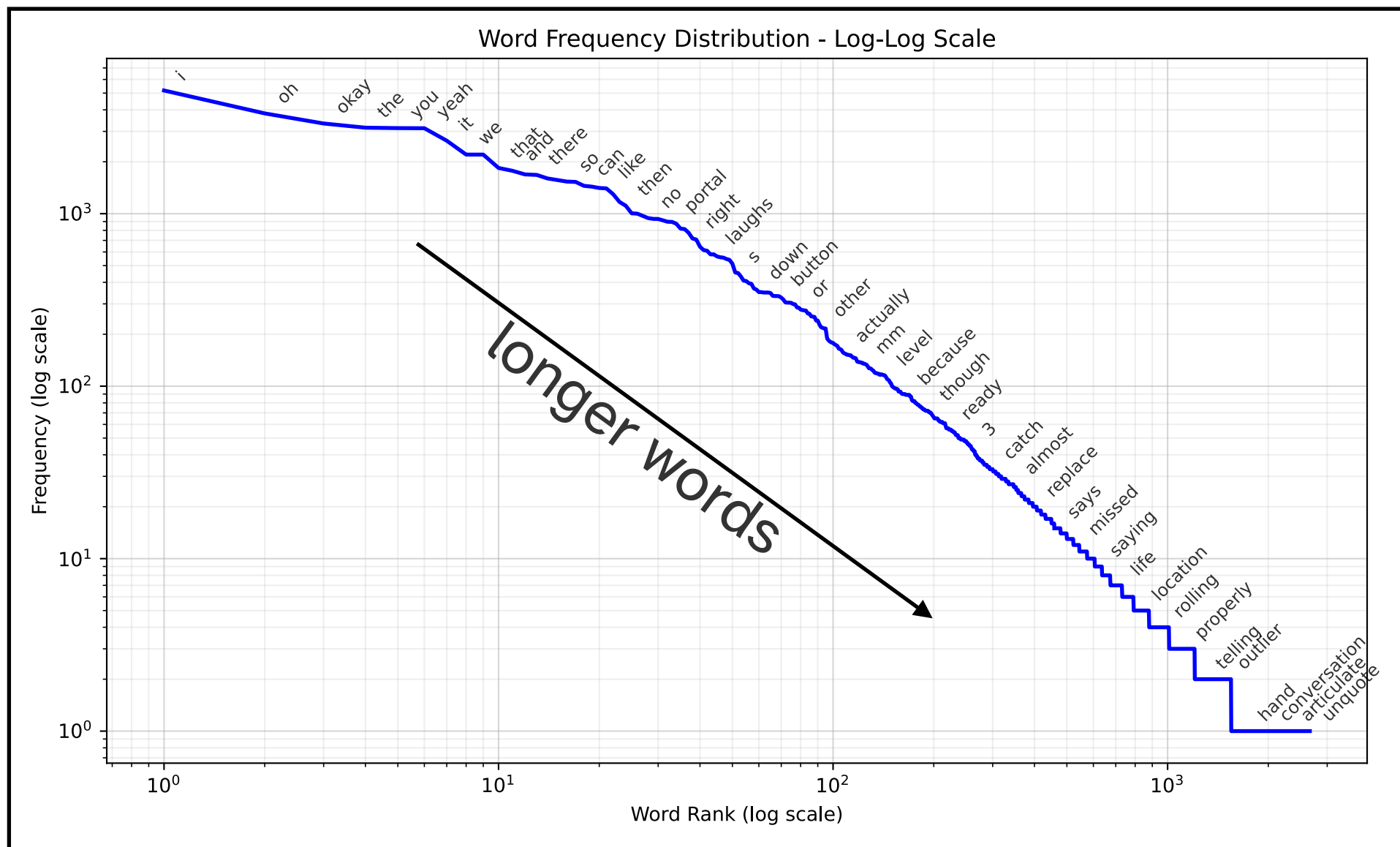
Long-tail (“Zipfian”) distribution of wordtypes (from Portal 2 dialogues)



Vocabularies



Long-tail (“Zipfian”) distribution of wordtypes (from Portal 2 dialogues)



Meaning and Representation



- Words are themselves not actual things, they only refer to things
 - Kind of like a pointer
 - Even if they aren't actual things, we can still do things to and know things about them
 - E.g., we know cats are a type of pet
- Core question: if a machine learning system is processing text, how should words be represented as input to and within the ML system?

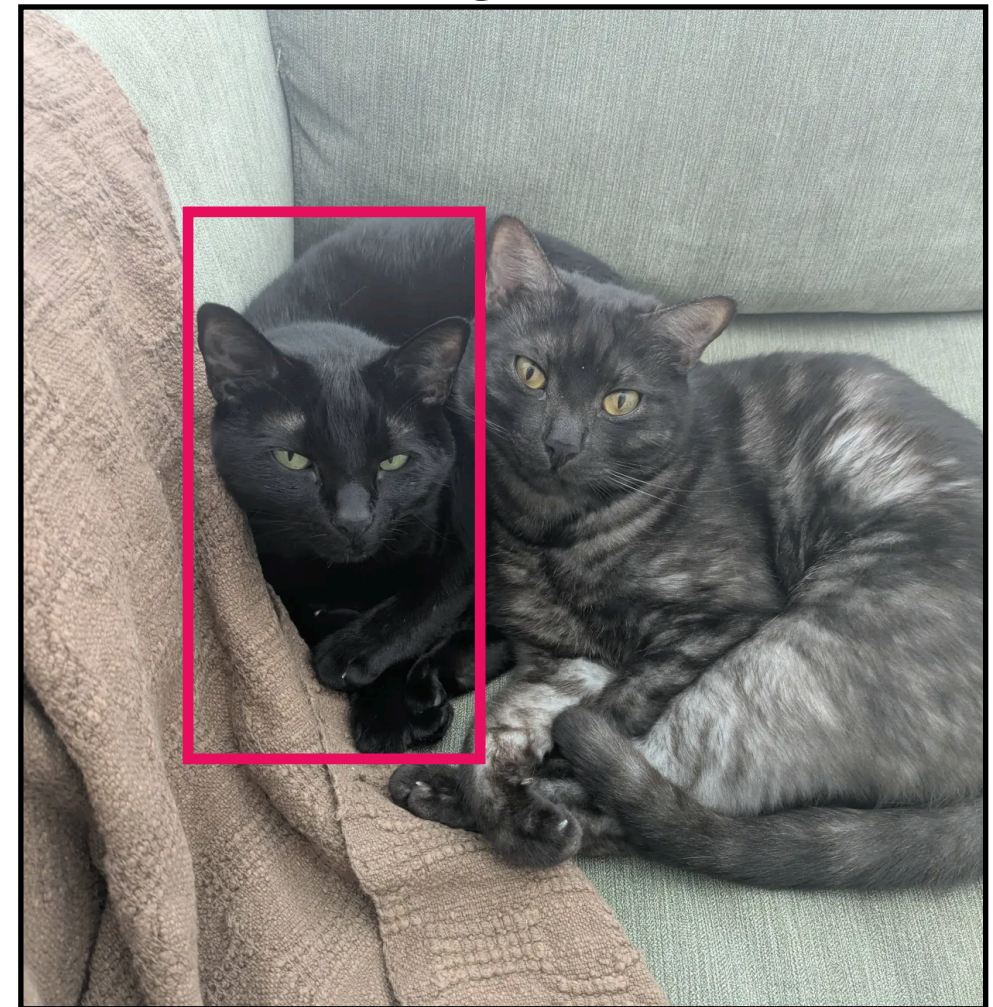
Word Meaning: Denotational Semantics



What color is Pepper?

- Let's start from the bottom up: what does each word mean?
 - What does “Pepper” mean?
- **Denotational semantics:** the symbol *refers to* something in the context in which the language is used

Image i



$$[[\text{Pepper}]]^i = \left\{ \text{Image of the black cat} \right\}$$

Word Meaning: Denotational Semantics

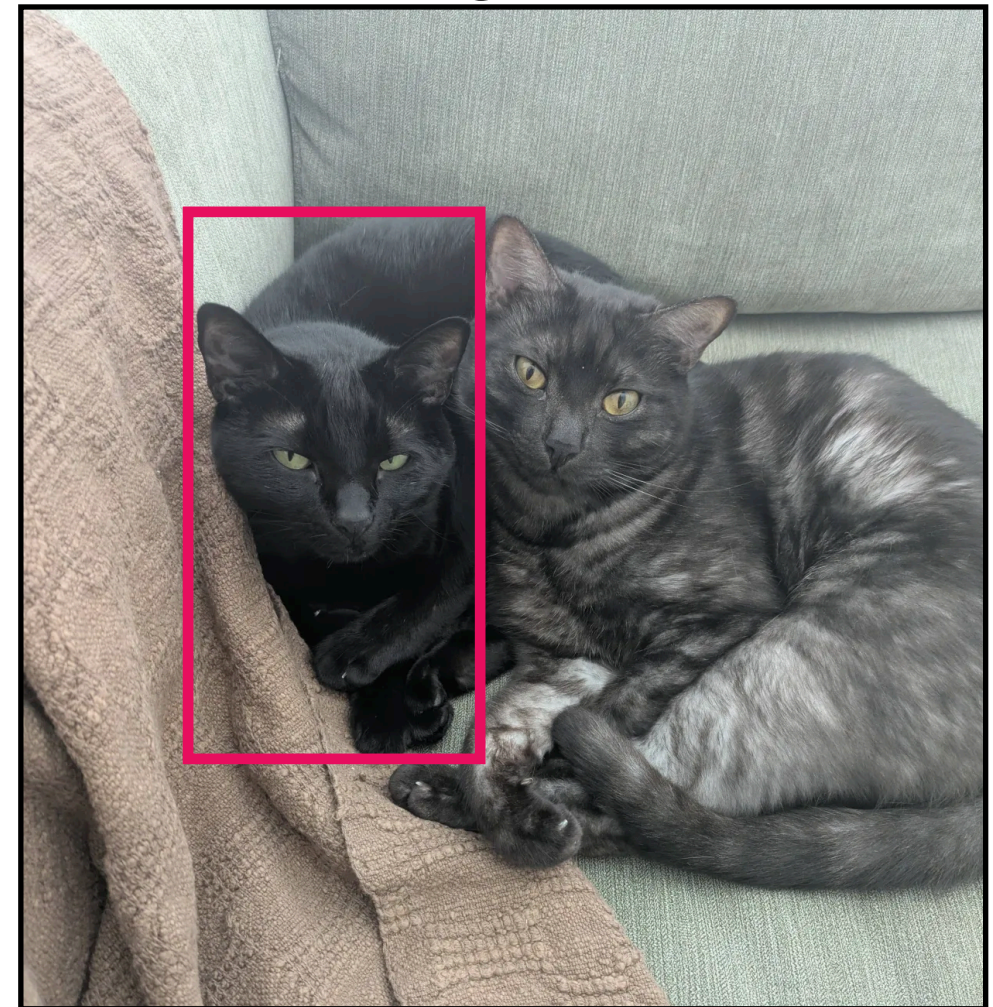


What color is  ?

Black with green eyes

- Denotational semantics allows us to *ground* language to a world outside of language use
- But what happens when we don't have an available outside world?

Image i



$$[[\text{Pepper}]]^i = \{ \img alt="A small image of a black cat with green eyes." data-bbox="844 745 934 982" \}$$

Word Meaning: Ontologies



Cats

Lucy's
Cats



Toast

Nick's
Cats



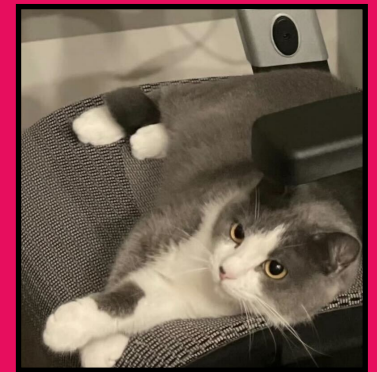
Coco

Téa's
Cats



Marceline

Zineng's
Cats



Shaomai

Alane's
Cats

Patrick



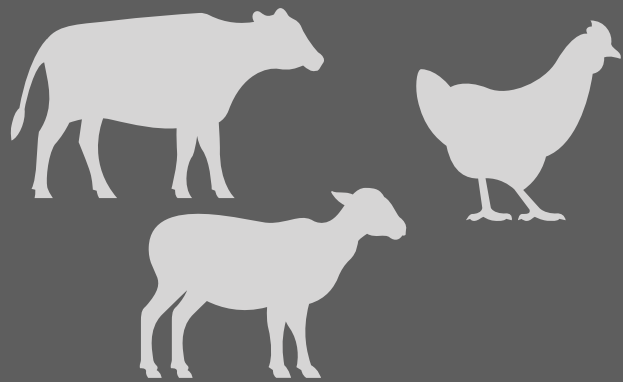
Pepper

Word Meaning: Ontologies

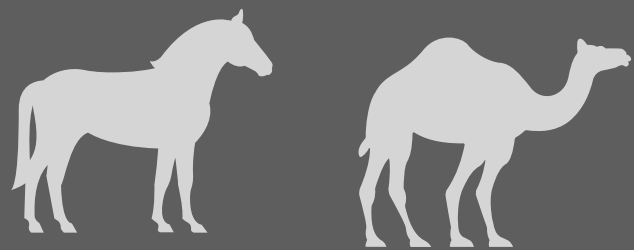


Domesticated Animals

Livestock



Draught Animals

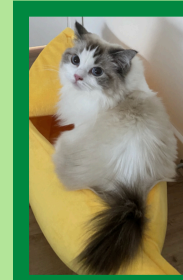


Cats

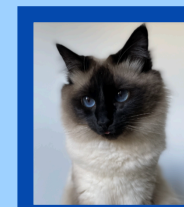


Alane's Cats

Lucy's
Cats



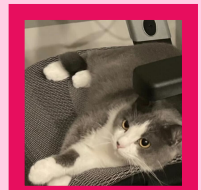
Nick's
Cats



Téa's
Cats



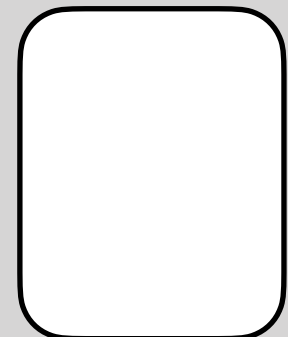
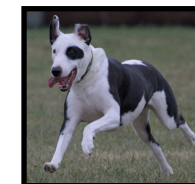
Zineng's
Cats



Pets

Dogs

Mer's Dogs



Word Meaning: Ontologies

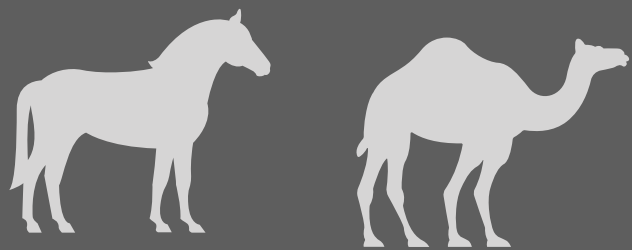


Domesticated Animals

Hyponym / hypernym relation:

“pet” is a hypernym of “dog”;
“draught animal” is a hyponym of
“domesticated animals”

Draught Animals



Pets

Dogs

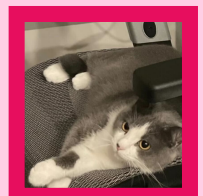
Mer's Dogs



Téa's
Cats



Zineng's
Cats



Synsets:

{“draught animal”, “beast of
burden”, “draft animal”}

Word Meaning: Ontologies



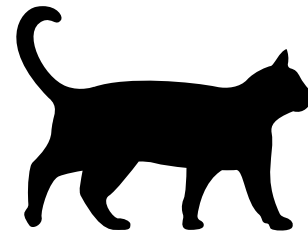
What color is Pepper?

- From our ontology, we know
 - `isa (Pepper, Alane's cat)`
 - `isa (Alane's cat, cat)`
- ← (hypernym relation)
- We might know something about cats, too:
 - Cats are typically not blue, pink, yellow, etc.
 - So the answer is probably not one of these

Word Meaning: Ontologies



What color is pepper?



- Missing context-dependence (polysemy)

Word Meaning: Ontologies



- Missing context-dependence (polysemy)
- Missing meaning of new words

In [cryptography](#), a **pepper** is a secret added to an input such as a [password](#) during [hashing](#) with a [cryptographic hash function](#). This value differs from a [salt](#) in that it is not stored alongside a password hash, but rather the pepper is kept separate in some other medium, such as a Hardware Security Module.^[1] Note that the [National Institute of Standards and Technology](#) refers to this value as a **secret key** rather than a **pepper**. A pepper is similar in concept to a [salt](#) or an [encryption key](#). It is like a salt in that it is a randomized value that is added to a password hash, and it is similar to an encryption key in that it should be kept secret.

Webster, Craig (2009-08-03). "[Securing Passwords with Salt, Pepper and Rainbows](#)" [↗](#).
Barking Iguana. Retrieved 2020-11-11.

Word Meaning: Ontologies



- Missing context-dependence (polysemy)
- Missing meaning of new words
- Requires human labor

There are perhaps fifty thousand *Capsicum* cultivars grown worldwide.^[1]

- *Capsicum annuum*, which includes bell peppers, cayennes, friggitello, jalapeños, paprika, and serrano.
 - *Capsicum annuum* 'New Mexico Group', common name Hatch or Anaheim, which includes Big Jim, Chimayó, and Sandia peppers.
- *Capsicum baccatum*, which includes the South American varieties, such as ají amarillo, ají limón, and criolla sella.
- *Capsicum chinense*, which includes all of the habaneros,^[5] Scotch bonnets, Trinidad Scorpions, the Bhut Jolokia, and the Carolina Reaper.
- *Capsicum frutescens*, which includes the Tabasco pepper and many of the peppers grown in India;^[6] sometimes not distinguished as a species separate from *C. annuum*.^{[7][8]}
- *Capsicum pubescens*, which includes the rocoto and manzano pepper, are distinctive plants, having violet flowers, black seeds, and hairy dark green leaves, and grow as a large, multi-stemmed vine up to 5 meters long.



Paullassiter, Wikipedia



Falcodigiada, Wikipedia



Hankwang, Wikipedia



Daniel Risacher, Wikipedia



Endwints, Wikipedia

Word Meaning: Ontologies



- Missing context-dependence (polysemy)
- Missing meaning of new words
- Requires human labor
- Subjective and culturally dependent



Word Meaning: Discrete Symbols



- Machine learning models expect numerical inputs
- Represent each word as a unique **one-hot vector** (vector with values 0, except for one value of 1)

$$\begin{aligned}\phi(\text{cat}) &= [0 \ 0 \ 0 \ 0 \ 1 \ 0 \ 0 \ 0 \ 0 \ 0] \\ \phi(\text{kitten}) &= [0 \ 0 \ 1 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0 \ 0]\end{aligned}$$

- Vector dimension: the same size as the vocabulary
- Problem: no notion of similarity

How can I make my house
safe for a new kitten?

*How can I make my house
safe for a new cat?*

- (Is there some way we could use WordNet to get more informative numerical representations of words?)

Word Meaning: Continuous Vectors



- Instead: represent words as continuous vectors

$$\begin{aligned}\phi(\text{cat}) &= \begin{bmatrix} -0.6 & 1.3 & -0.1 & 0.3 \end{bmatrix} \\ \phi(\text{kitten}) &= \begin{bmatrix} -0.7 & -1.5 & 0.0 & 0.3 \end{bmatrix}\end{aligned}$$

- Implicitly provides notions of similarity:

$$||\phi(\text{cat}), \phi(\text{kitten})|| < ||\phi(\text{cat}), \phi(\text{dog})||$$

- Similarity is learnable from text at scale:

*... feeding time in multiple **cat** households can often be ...*

*... is to relieve the **cat** from the stress of ...*

*... effect on the feral **cat** population in rural areas ...*

*... So I wrapped the **kitten** up in a towel ...*

*... they noticed that the **kitten** population was getting high ...*

*... were looking for a **kitten** , hoping that someone ...*

Word Embeddings



- How do we get these vectors?
- Core principle: **distributional hypothesis**
 - Words that are used in similar contexts have similar meanings
 - Context: typically, other words in a text, but really anything can be context!

Distributional Hypothesis



A bottle of tesgüino is
on the table.

Everybody likes
tesgüino.

Tesgüino makes you
drunk.

We make tesgüino out
of corn.

- Found in bottles
- People like it
- It makes you drunk
- It's made of corn
- **Maybe:** similar to beer, wine, whiskey, etc.

Word Co-Occurrence



- Co-occurrence matrix, counting number of texts in which both words occur
- Word similarity $\approx$ cosine similarity of vector representations ϕ
- But embedding is huge: the size of the vocabulary!

	tesgüino	beer	wine	whiskey
drunk	3	400	450	600
bottle	10	600	1000	600
corn	15	0	0	400
hops	0	450	0	0
grapes	0	0	1000	0
dairy	0	0	0	0

Word2Vec: Skip-Gram



- Input: corpus D including pairs (w, c) where w is a word and c is a context, e.g.:

- “Everybody likes tesgüino”

$(w = \text{“tesgüino”}, c = \text{“likes”})$

$(w = \text{“tesgüino”}, c = \text{“everybody”})$

e.g., context is every
word in the sentence

- We want to model: given that we observe word w , what’s likely in the context c ?

$$p(c \mid w; \theta)$$

- Our objective: find parameters that maximize the corpus probability

$$\arg \max_{\theta} \prod_{(w, c) \in \mathcal{D}} p(c \mid w; \theta)$$

- (There are other methods for word2vec (e.g., GloVe, CBOW), but we will only look at Skip-Gram in this class)

Word2Vec: Skip-Gram



Objective: find

$$\arg \max_{\theta} \prod_{(w,c) \in \mathcal{D}} p(c \mid w; \theta)$$

where

$$p(c \mid w; \theta) = \frac{\exp(\text{score}(c, w))}{\sum_{c' \in \mathcal{C}} \exp(\text{score}(c', w))}$$

score: similarity of

we use cosine similarity because, in contrast to dot product, it is unbiased towards vector

representations of word and context

unbiased towards vector

magnitude: $\text{score}(c, w) = \cos(\phi(c), \phi(w)) = \frac{\phi(c) \cdot \phi(w)}{||\phi(c)|| ||\phi(w)||}$

Word2Vec: Skip-Gram



Objective: find

$$\arg \max_{\theta} \prod_{(w,c) \in \mathcal{D}} p(c \mid w; \theta)$$

where

$$p(c \mid w; \theta) = \frac{\exp(\text{score}(c, w))}{\sum_{c' \in \mathcal{C}} \exp(\text{score}(c', w))}$$

softmax function:
normalizes categorical
scores to a smooth
probability distribution

Word2Vec: Skip-Gram



Objective: find

$$\arg \max_{\theta} \prod_{(w,c) \in \mathcal{D}} \frac{\exp(\text{score}(c, w))}{\sum_{c' \in \mathcal{C}} \exp(\text{score}(c', w))}$$

**Problem: intractable to
sum over all contexts**

Word2Vec: Skip-Gram



- * Solution: negative sampling
 - * Instead of summing over all possibly contexts in which a word could appear
 - * Approximate via learning from contexts in which a word *doesn't* appear
- * Model whether or not a word-context pair is likely to exist in the dataset

$$p((w, c) \in \mathcal{D}) = \frac{1}{1 + \exp(-\text{score}(c, w))}$$

similarity of representations of
word and context

Word2Vec: Skip-Gram



- * Solution: negative sampling
 - * Instead of summing over all possible contexts in which a word could appear
 - * Approximate via learning from contexts in which a word *doesn't* appear
- * Model whether or not a word-context pair is likely to exist in the dataset

$$p((w, c) \in \mathcal{D}) = \frac{1}{1 + \exp(-\text{score}(c, w))}$$

sigmoid function: normalizes a scalar into a probability

Word2Vec: Skip-Gram



NEW objective: find

$$\arg \max_{\theta} \prod_{(w,c) \in \mathcal{D}} p((w,c) \in \mathcal{D}) \prod_{(w,c) \in \mathcal{D}'} p((w,c) \notin \mathcal{D})$$

What is $\mathcal{D}'$?

where

$$p((w,c) \in \mathcal{D}) = \frac{1}{1 + \exp(-\text{score}(c, w))}$$

$$p((w,c) \notin \mathcal{D}) = 1 - p((w,c) \in \mathcal{D})$$

Word2Vec: Skip-Gram



NEW objective: find

$$\arg \max_{\theta} \prod_{(w,c) \in \mathcal{D}} p((w,c) \in \mathcal{D}) \prod_{(w,c) \in \mathcal{D}'} p((w,c) \notin \mathcal{D})$$

- * Negative samples: sample and train on $l * |D|$ pairs (w', c') where

$$w' \sim p(W) \quad c' \sim p(C)$$

unigram prior unigram prior
over words over contexts

- * How to train?
 - * Gradient descent
 - * In practice, we work with log probabilities, not direct probabilities, to avoid float underflow

Word2Vec: Skip-Gram



What counts as context? Basically anything you want

A bottle of tesgüino is
on the table.

Everybody likes
tesgüino.

Tesgüino makes you
drunk.

We make tesgüino out
of corn.

	$w = \text{tesgüino}$
of w	1
bottle of w	1
bottle __ w	1
cup __ w	0
w is	1
w is on	1
w is on the table	1
w __ __ table	1
{corn, w }	1
make w out	1

Word Vectors



- * In Skip-Gram, the parameters of our models are the vectors for words and the vectors for contexts

$$p((w, c) \in \mathcal{D}) = \frac{1}{1 + \exp(-\cos(\phi(c), \phi(w)))}$$

context vectors word vectors

- * Once we've trained a model, we can use these vectors for whatever we'd like!

Vector Arithmetic



$$a : b :: c : ? \equiv d = \arg \max_{i \in \mathcal{W}} \frac{(\phi(b) - \phi(a) + \phi(c)) \cdot \phi(i)}{\|(\phi(b) - \phi(a) + \phi(c)) \cdot \phi(i)\|}$$

vector representing the
relationship between a and b

Vector Arithmetic



$$a : b :: c : ? \equiv d = \arg \max_{i \in \mathcal{W}} \frac{(\phi(b) - \phi(a) + \phi(c)) \cdot \phi(i)}{\|(\phi(b) - \phi(a) + \phi(c)) \cdot \phi(i)\|}$$

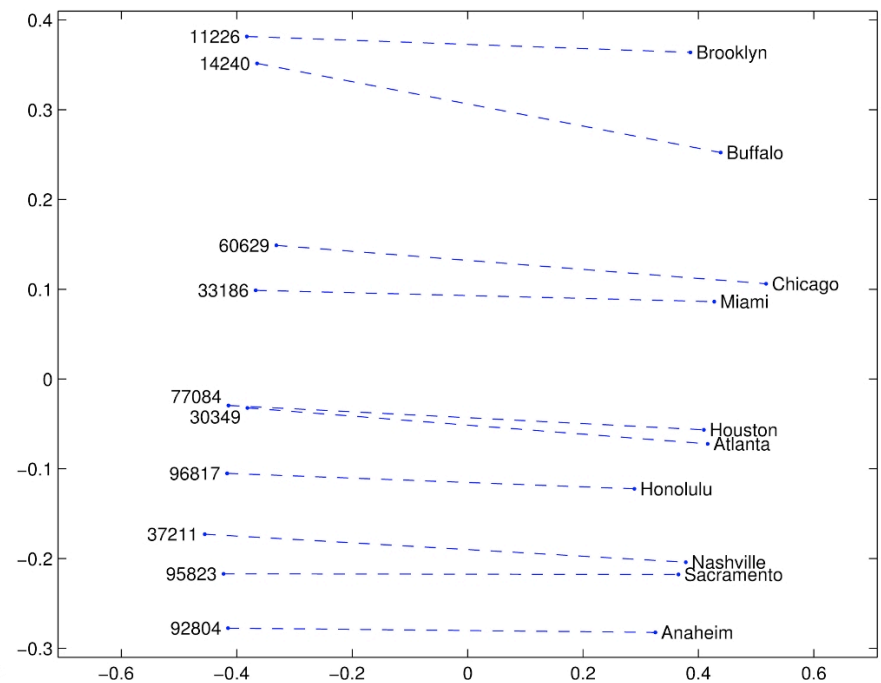
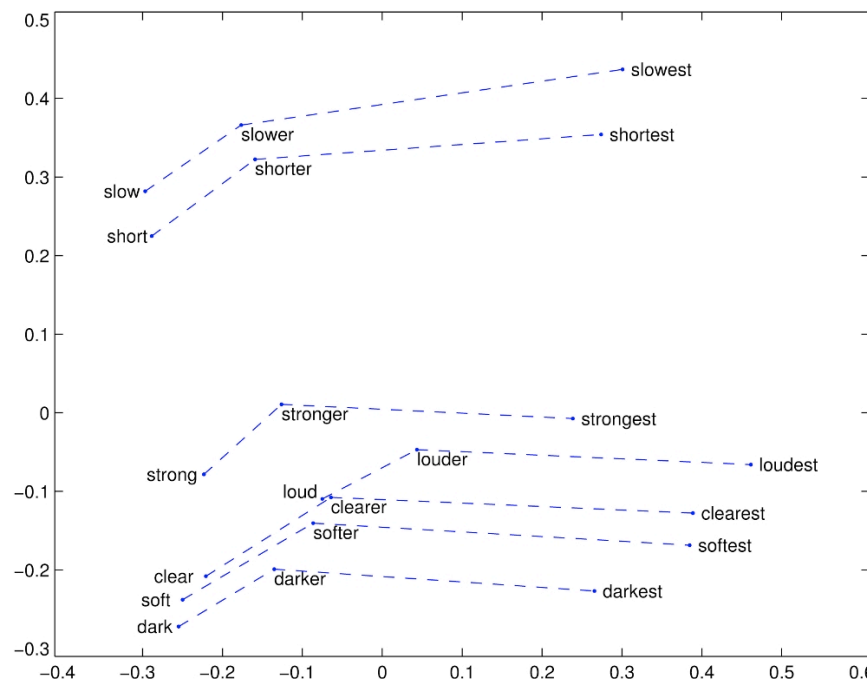
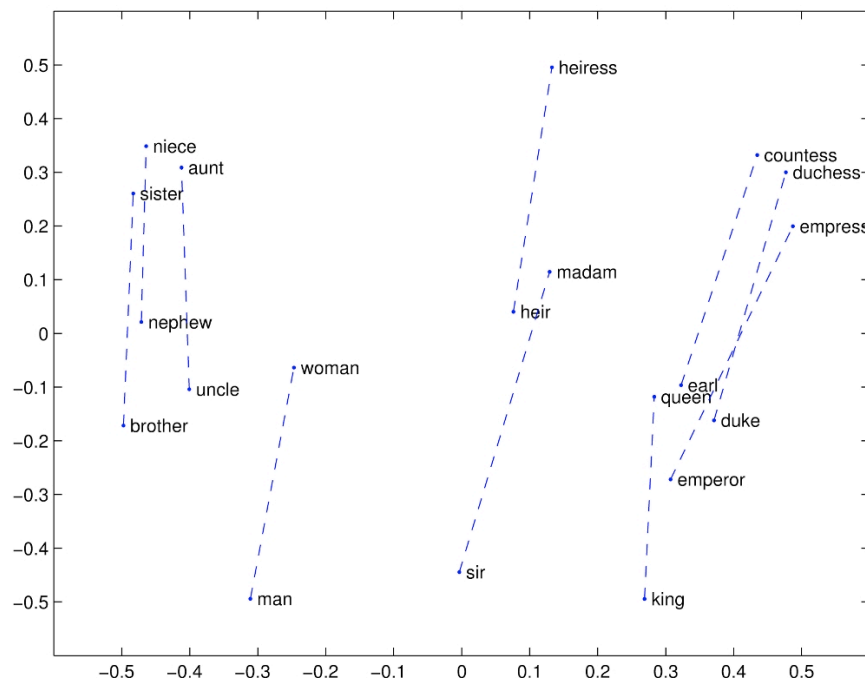
apply this relationship
to a new word c

Vector Arithmetic



$$a : b :: c : ? \equiv d = \arg \max_{i \in \mathcal{W}} \frac{(\phi(b) - \phi(a) + \phi(c)) \cdot \phi(i)}{\|(\phi(b) - \phi(a) + \phi(c)) \cdot \phi(i)\|}$$

cosine similarity of the expected representation of the analogy word and some other word i



Vector Arithmetic



Pairs (x, y) that maximize $\cos(\phi(\text{he}) - \phi(\text{she}), \phi(x) - \phi(y))$

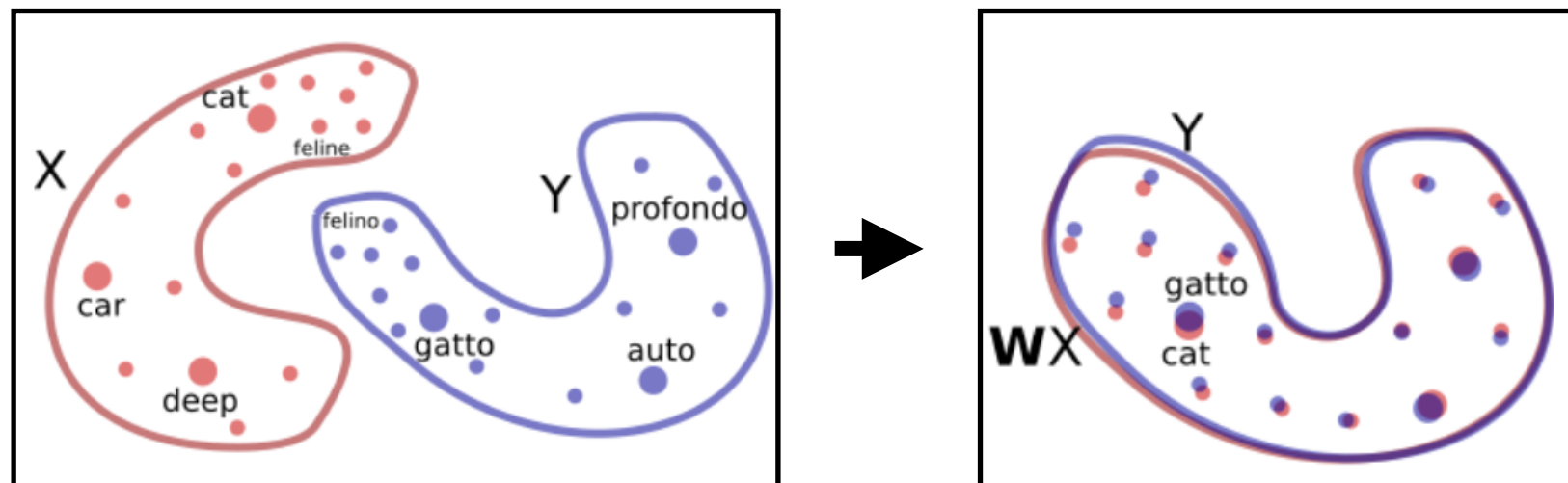
Gender stereotype <i>she-he</i> analogies.		
sewing-carpentry	register-nurse-physician	housewife-shopkeeper
nurse-surgeon	interior designer-architect	softball-baseball
blond-burly	feminism-conservatism	cosmetics-pharmaceuticals
giggle-chuckle	vocalist-guitarist	petite-lanky
sassy-snappy	diva-superstar	charming-affable
volleyball-football	cupcakes-pizzas	hairstylist-barber
Gender appropriate <i>she-he</i> analogies.		
queen-king	sister-brother	mother-father
waitress-waiter	ovarian cancer-prostate cancer	convent-monastery

- Stereotypes are embedded in text data
- One way to “debias”: identify category subspace (e.g., “gender” subspace), project words onto that subspace, then subtract those projections from the original word

Vector Arithmetic



- What if we have embeddings in different languages and want to align them? Given:
 - ϕ^A : language A's embedding function
 - ϕ^B : language B's embedding function
 - $\mathcal{D} = \{w_i^A, w_i^B\}_{i=1}^M$: dataset pairing M word translations
- Find a matrix W such that
$$\min_W \sum_{i=1}^M ||W\phi(w_i^A) - \phi(w_i^B)||^2$$





Challenge: Polysemy

Verb [\[edit \]](#)

run (*third-person singular simple present* **runs**, *present participle* **running**, *simple past* **ran**, *past participle* **run** or (*nonstandard, colloquial*) **ran**)

1. To move swiftly.

1. (*intransitive*) To **move** forward **quickly** upon **two feet** by alternately making a short **jump** off either foot. [\[coordinate term ▲\]](#) [\[quotations ▼\]](#)

Coordinate term: **walk**

***Run**, and you might still catch the train!*

2. (*intransitive*) To **go** at a **fast** pace; to move **quickly**. [\[quotations ▼\]](#)

*I have been **running** all over the building looking for him.*

*Sorry, I've got to **run**; my house is on fire!*

3. (*transitive*) To cover (a course or a distance) by running.

*I can **run** a mile, but I can't **run** the cross-country course.*

4. (*transitive*) To complete a running course or event in (a given time).

*I was hoping to make the team, but I didn't **run** the qualifying time.*

5. (*intransitive*) To move briskly or smoothly with a motion of sliding, rolling, sweeping etc.

*The shuttle **runs** back and forth on these rollers.*

*As its name suggests, the monorail **runs** on a single rail.*

*I felt her fingers **running** over my cheek.*

• • •

31. To tend, as to an effect or consequence; to incline. [\[quotations ▼\]](#)

32. To have a legal course; to be attached; to continue in force, effect, or operation; to follow; to go in company. [\[quotations ▼\]](#)

*Certain covenants **run** with the land.*

33. To **encounter** or **suffer** (a particular, usually bad, **fate** or **misfortune**). [\[quotations ▼\]](#)

34. (*golf*) To strike (the ball) in such a way as to cause it to run along the ground, as when approaching a hole.

35. (*video games, rare*) To **speedrun**.

36. (*sports, especially baseball*) To **eject** from a game or match.

*Jackson got himself **run** in the top of the sixth for arguing a borderline strike three call.*

37. To press (a bank, etc.) with immediate demands for payment.

Noun [\[edit \]](#)

run (*plural* **runs**)

1. Act or instance of running, of moving rapidly using the feet. [\[quotations ▼\]](#)

*I just got back from my morning **run**.*

2. Act or instance of **hurrying** (to or from a place) (*not necessarily on foot*); **dash** or **errand, trip**. [\[quotations ▼\]](#)

*I need to make a **run** to the store.*

3. A pleasure trip. [\[quotations ▼\]](#)

*Let's go for a **run** in the car.*

4. **Flight**, instance or **period** of **fleeing**. [\[quotations ▼\]](#)

5. **Migration** of fish.

6. A group of fish that migrate, or ascend a river for the purpose of spawning.

7. A literal or figurative path or course for movement relating to:

1. A (regular) **trip** or **route**. [\[quotations ▼\]](#)

*The bus on the Cherry Street **run** is always crowded.*

2. The **route** taken while running or skiing.

*Which **run** did you do today?*

• • •

21. A line of **knit stitches** that have **unravelled**, particularly in a **nylon stocking**. [\[quotations ▼\]](#)

*I have a **run** in my stocking.*

22. (*nautical*) The **stern** of the underwater body of a **ship** from where it begins to curve upward and inward.

23. (*mining*) The horizontal distance to which a **drift** may be carried, either by licence of the **proprietor** of a **mine** or by the nature of the formation; also, the direction which a **vein** of **ore** or other substance takes.

24. A pair or set of **millstones**.

25. One's **gait** while running; the way one runs.

*I think they only have a weird **run** because their leg hurts.*

Adjective [\[edit \]](#)

run (*not* **comparable**)

1. In a **liquid** state; **melted** or **molten**. [\[quotations ▼\]](#)

*Put some **run** butter on the vegetables.*

2. **Cast** in a **mould**. [\[quotations ▼\]](#)

3. **Exhausted**; **depleted** (*especially with "down" or "out"*).

4. (*of a zoology*) **Travelled**, **migrated**; having made a **migration** or a spawning **run**. [\[quotations ▼\]](#)

5. **Smuggled**.

***run** brandy*

Challenge: Polysemy



- Word embedding now has to somehow encode the meanings of all 67 senses of “run”...
 - Why is this hard?
- What could we do instead?
- Contextualized word embeddings: compute representation of individual words dependent on the context in which it appears
 - The resulting architectures were some of the main catalysts for the design of modern LLMs!

Opportunity: Sentence Representations



- If we have representations for parts of sentences (wordtypes), then can we get a representation of the whole sentence?
- One option: bag-of-words representation

$$\phi(\bar{x}) = \frac{1}{|\bar{x}|} \sum_{i=0}^{|\bar{x}|} \phi(x_i)$$

- What's missing?

Word Embeddings in the Age of LLMs

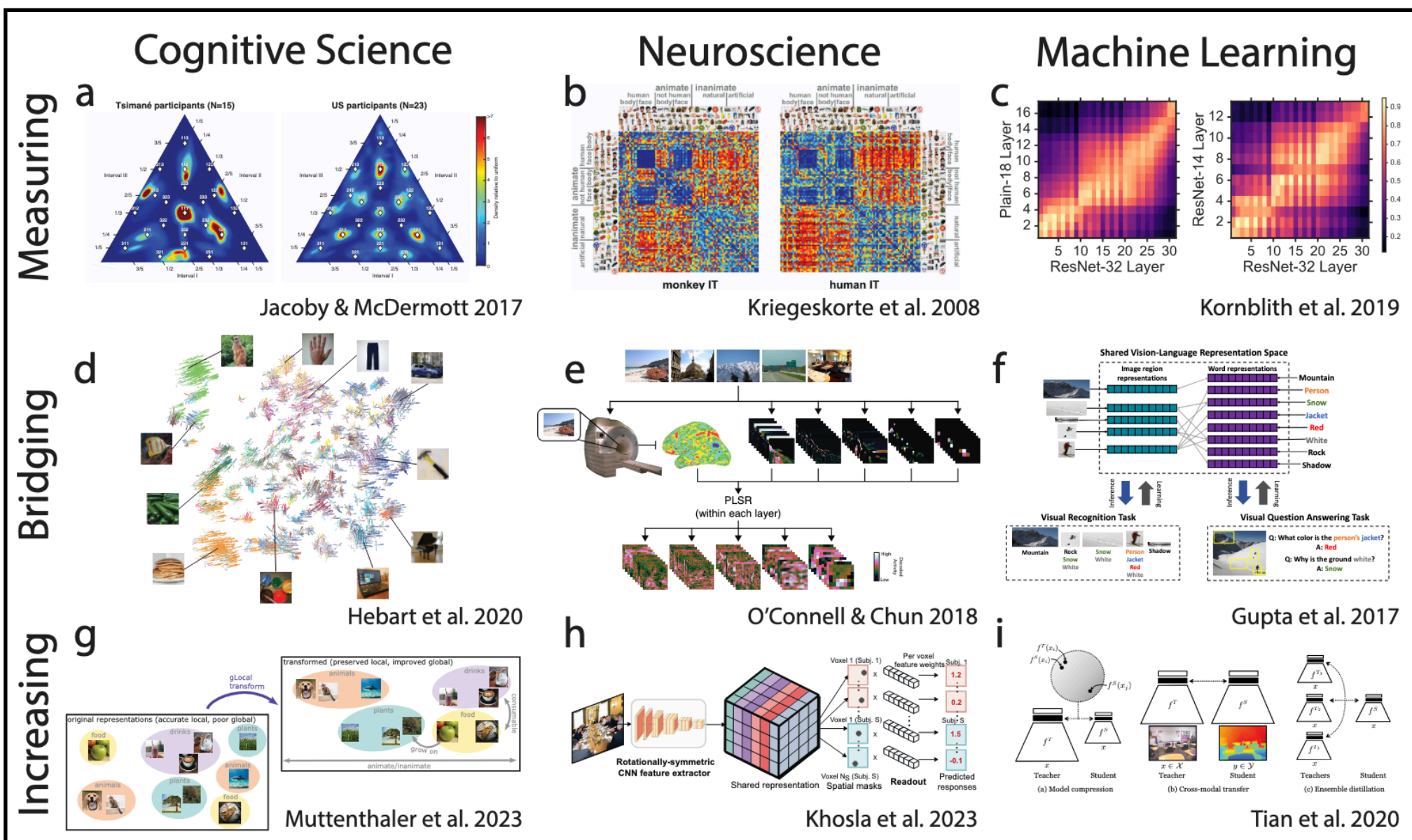


- Building a language technology pre-LLMs, for a target task:
 - Download some pre-trained word embeddings from the Internet
 - Initialize your language model with these word embeddings (all other parameters randomized)
 - Use your task-specific data to fine-tune the other parameters (possibly also fine-tuning the word embeddings)
 - Having a good starting point via word embeddings is **really** important, especially with little task-specific data
- But we don't do this anymore
 - Language models still learn embeddings specific to each wordtype
 - But we don't download them from the Internet — we download the whole language model

Word Embeddings in the Age of LLMs



So why might we still care about word embeddings?

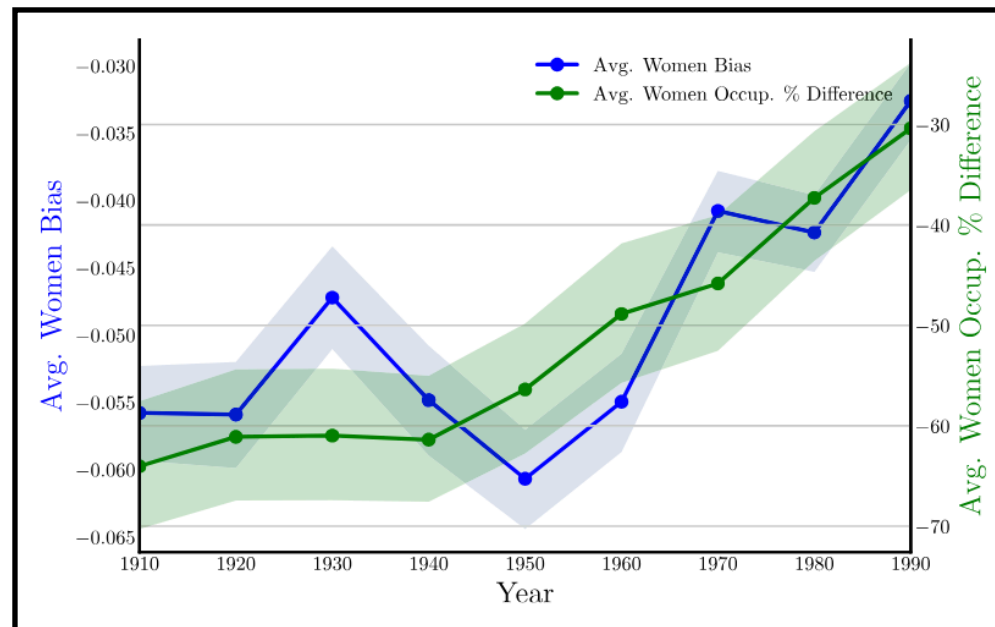
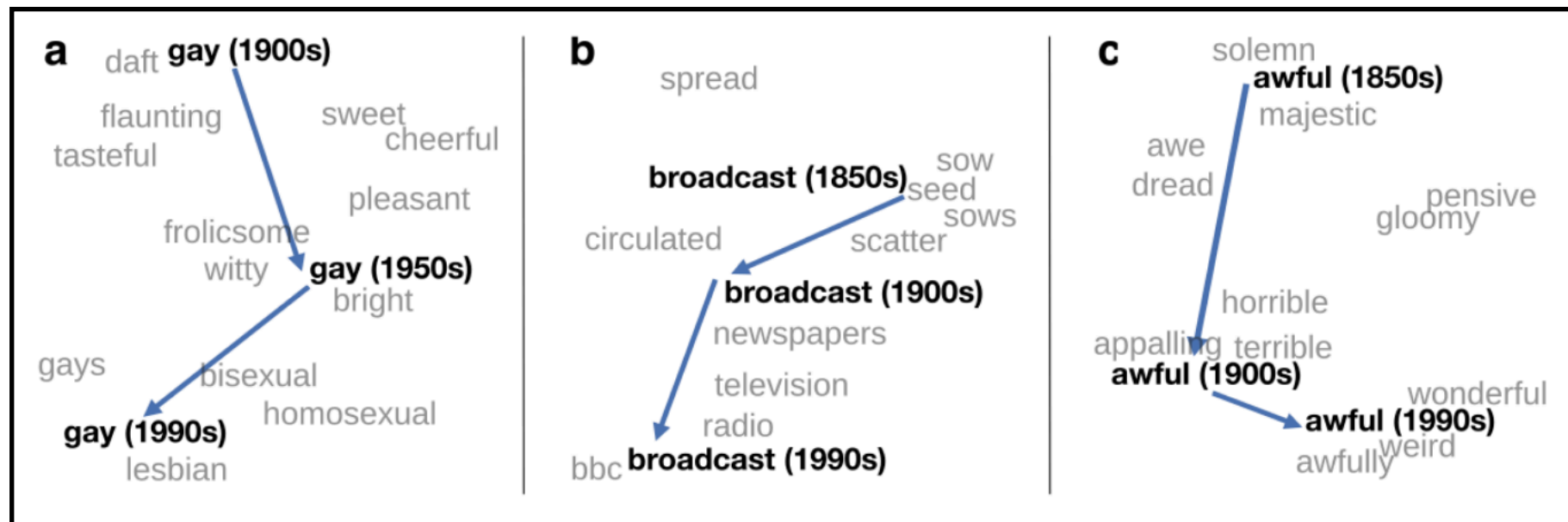


Evaluating how well an LLM's learned concept space correlates with human concept spaces

Word Embeddings in the Age of LLMs



So why might we still care about word embeddings?



1910	1950	1990
Charming	Delicate	Maternal
Placid	Sweet	Morbid
Delicate	Charming	Artificial
Passionate	Transparent	Physical
Sweet	Placid	Caring
Dreamy	Childish	Emotional
Indulgent	Soft	Protective
Playful	Colorless	Attractive
Mellow	Tasteless	Soft
Sentimental	Agreeable	Tidy

Analyze how language changes over time