

Stancetaking in Spontaneous Speech

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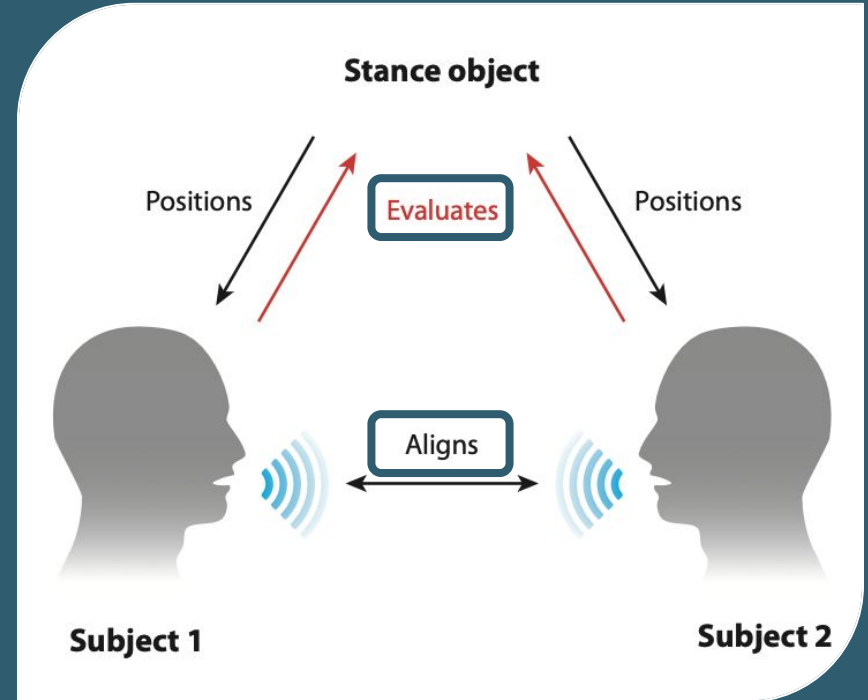
01

Background and Motivation

Stance and stancetaking

- Stance and stancetaking refer to the ways people position themselves in conversations
 - ◆ In terms of politeness, certainty, or emotion
 - ◆ In regard to interactants and content/objects of conversation
- Helps explain the patterning of language and the motivations for the use of lexical items, constructions, and discourse markers
- Method of analysis used in discourse analysis (and sociolinguistics)
- Many different theories and models...

The Stance Triangle



(Du Bois, 2007, as cited in Kiesling, 2022, p. 418)

Updated Stancetaking Model

→ 3 “speaker” participant roles

- ◆ Author
- ◆ Animator
- ◆ Principal

→ 3 dimensions

- ◆ Evaluation
- ◆ Alignment
- ◆ **Investment**

(2a) Scott **likes** coffee.

(2b) Scott **loves** coffee.

(2c) **I think** Scott loves coffee.

(2d) **I know** that Scott loves coffee.

(2e) **Kim told me** that Scott likes coffee.

(2f) **Dude,** coffee is great.

(Kiesling, 2022, p. 420)

Investment

Investment can be modulated through epistemicity and evidentiality...BUT it can also be embedded in the meanings of...

- Lexical items and constructions
(*like* vs. *love*)
- Discourse markers
(*I mean, just sayin'*)
- Hedges and boosters

Classification	Example
Hedges	
modal verbs	<i>could, might, would</i>
epistemic adjectives, adverbs and nouns	<i>perhaps, likely, interpretation</i>
lexical verbs	<i>seem, assume, suggest</i>
Boosters	
modal verbs	<i>must, will</i>
epistemic adjectives, adverbs and nouns	<i>obvious, always, argument</i>
lexical verbs	<i>demonstrate, show, find</i>

(Wang & Jiang, 2018)

The Present Study

Investigating investment through lexical items and discourse markers in a corpus of spontaneous speech

- Lexical items: *like* vs. *love*
- Discourse markers: *well, fine, good, great, right, I mean*

Research Questions:

- Which lexical item/discourse marker is used most frequently in this corpus?
- Which is most informative of the animator's investment?
- What generalizations can be made for the investment level of each item/marker?

Santa Barbara Corpus of Spoken American English (SBCSAE)

- 60 recordings/transcriptions and ~249,000 words
 - ◆ Only using 43 of the 60 (face-to-face spontaneous speech)
- All over the U.S. – wide variety of people of different regional origins, ages, occupations, genders, and ethnic and social backgrounds

SBC001 *Actual Blacksmithing*

This is a conversation recorded in rural Hardin, Montana. Mae Lynne is a student of equine science, and is the main speaker. She is telling Lenore (a visitor and near stranger) about her studies. Doris, Mae Lynne's mother, is doing housework, but joins the conversation near the end to discuss friends of their family.

Audio: [WAV](#) [MP3](#) Text: [TRN](#) [CHAT](#)

SBC002 *Lambada*

After-dinner conversation among four friends in San Francisco, California. Participants are in their late twenties or early thirties. Harold and Jamie are a married couple, Miles is a doctor, and Pete is a graduate student from Southern California.

Audio: [WAV](#) [MP3](#) Text: [TRN](#) [CHAT](#)

SBC003 *Conceptual Pesticides*

A conversation among three friends who are preparing dinner together, recorded in Southern California. Roy and Marilyn are a married couple, and Pete is a friend visiting from out of town. All participants are in their early thirties.

Audio: [WAV](#) [MP3](#) Text: [TRN](#) [CHAT](#)

SBC004 *Raging Bureaucracy*

Family conversation recorded in Santa Fe, New Mexico. The primary participants are three sisters all in their twenties.

Audio: [WAV](#) [MP3](#) Text: [TRN](#) [CHAT](#)

SBCSAE – SBC001.trn

0.00 9.21
9.21 9.52
9.52 14.10
14.10 15.78
15.01 16.78
16.78 18.32
18.33 18.85
18.85 20.69
20.69 21.19
21.19 21.74
21.26 22.24
22.24 23.23
22.28 24.25
24.25 25.07
25.07 26.14
26.14 26.62
26.62 26.87
26.87 28.37
28.37 30.10
30.10 32.36
32.36 32.59
32.59 33.90
33.90 34.22
34.22 35.23
35.23 36.88
36.88 39.30
39.30 39.80
39.80 40.27

LENORE:

LYNNE:

DORIS:

LYNNE:

DORIS:

LYNNE:

... So you don't need to go ... borrow equipment from anybody,
to --
... to do the feet?
... [do the hooves]?
[(H)=] <YWN Well,
we're gonna have to find somewhere,
to get,
(Hx) ... something (Hx) YWN>.
.. So,
[~Mae-]
[I'm gonna] (Hx) --
[2~Mae ~Lynne XX2]
[2(H) We're not2] gonna do the feet today,
I'm gonna wait till like,
early in the morning=,
.. to do those,
cause y- --
I mean you get s=s ti=red.
(H) ... n- you just,
... it takes % --
well,
it takes me longer than most people,
cause you know,
I'm not as stro=ng and,
(H) ... and I'm not as good,
as like somebody that would do it .. all the ti=me.
.. You know.
.. I mean,

SBCSAE – Annotation key

Santa Barbara Corpus of Spoken American English Part-II

Units	
Intonation Unit	RETURN
Truncated intonation unit	---
word	SPACE
truncated word	-

Speakers	
Speaker identity/turn start	:
Speech overlap	[]

Transitional Continuity	
Final	.
Continuing	,
Appeal	?

Terminal Pitch Direction	
Fall	\
Rise	/
Level	-

Accent and Lengthening

Primary accent	^
Secondary accent	'
Booster	!
Lengthening	=

Tone

Fall	\
Rise	/
Fall-rise	\ /
Rise-fall	/ \
Level	-

Pause

Long	...(N)
Medium	...
Short	..
Latching	(0)

Vocal Noises

Vocal noises	()
Inhalation	(H)
Exhalation	(Hx)
Glottal stop	%
Laughter	@

Quality

Quality	<Y Y>
Laugh quality	<@ @>
Quotation quality	<Q Q>
Multiple quality features	<Y <Z Z> Y>

Phonetics

Phonetic transcription	(/ /)
------------------------	-------

Transcriber's Perspective

Researcher's comments	(())
Uncertain hearing	<X X>
Indecipherable syllable	X

Specialized notation

Duration	(N)
Intonation unit continued	&
Intonation subunit boundary	
Embedded intonation unit	< >
Reset	
False start	< >
Codeswitching	<L2 L2>

Non-transcription Lines

Comment	\$
Interlinear gloss	\$G

Reserved Symbols

Phonemic/orthographic	'
Morphosyntactic coding	= * # { }
User-definable	" ~ ;

Why spontaneous speech?

Spontaneous speech vs. read speech

- Boundaries between tone units at different points
- Different positions of stress
- Fewer pauses in read speech
- Locations of pauses differ

Discourse analysis

- Studies natural interactions between interlocutors
- Read speech could help gather acoustic measurements, but spontaneous speech is needed to conduct a discourse analysis

Motivation

- Discourse Analysis with Dr. Scott Kiesling
- Working with speech data with Dr. Dan Villarreal



- 21 S: I mean this- ↑these are expensive and look how gro:ss it is ((pan clinks)) and then this thing=
- 22 A: =but ↑why would they make them so expensive if you can't use it?
- 23 S: you can use it but you have to-they're s-they're so ↑go:od that they need so: much care=
- 24 A: =they're so good that they're ↓bad

A Damaged Pan

The conversation below takes place on a Wednesday morning between Anjali and her mother, Shilpa. Both of them are in the kitchen getting breakfast between college/work meetings when Shilpa begins passionately informing Anjali about a damaged pan.

Transcription Key

↑↓	rising and falling intonation
:	extension of a syllable
	overlapping utterances
=	no interval between utterances
-	stutter
-	emphasis
((sound))	non-word sounds
(0.0)	interval of silence over ½ second

Transcript

01 Shilpa: um this th-thing is like
02 it's so expensive and a:ll this stuff is so messed up and
the-they're like
03 Anjali: ↑what?
04 Shilpa: they're like
05 I mean ↑look at it like it's
06 I-I did I did bak:ing soda so now it looks [better]
07 Anjali: [↑↑0.0]
08 Shilpa: but I was they were like you're only supposed to u- they
said this is so (1.1) sensitive you're only supposed to
use ↓low to medium ↑heat you can only use ↑butter like
there's all [these rules.]
09 Anjali: [↑why:??] it's a ↑0a:0
10 Shilpa: no but it's like (0.6) t- and ↑then I'm like I ↑said um
(0.9)
11 I've bought these I've given 'em as gifts (0.7) so I was
like
12 so: are you 0000a honor your warranty
13 Anjali: [↑↑0.0]
14 Shilpa: =and they said well you're ↑using it ↓wrong so no:
15 Anjali: ((stuttered laugh))
16 Shilpa: but then they said okay we 0000l
17 so: they're 0000a send me a new one and take 000- take
this back (1.1)
18 Anjali: [but what's the]
19 Shilpa: [but now I have to] use it at low to medium-
look I mean lo:ok at it it's like 0000l- like
20 I mean this- ↑these are expensive and look how 0000ss it
21 is ((pan clinks)) and then this thing=

02

Methods and Data

Extracting tokens

```
1 wells = {}
2 fines = {}
3 goods = {}
4 greats = {}
5 rights = {}
6 imeans = {}
7
8 # iterate through each df in the dict
9 for df_name, df in dfs.items():
10
11     # extract rows where the target word is found in the text column
12     well_rows = df[df['text'].str.contains(r'\bwell\b', case=False, regex=True)]
13     fine_rows = df[df['text'].str.contains(r'\bfine\b', case=False, regex=True)]
14     good_rows = df[df['text'].str.contains(r'\bgood\b', case=False, regex=True)]
15     great_rows = df[df['text'].str.contains(r'\bgreat\b', case=False, regex=True)]
16     right_rows = df[df['text'].str.contains(r'\bright\b', case=False, regex=True)]
17     imean_rows = df[df['text'].str.contains(r'\bI\smean\b', case=False, regex=True)]
18
19     # list of tuples containing (row #, text) for each occurrence
20     well_occurrences = [(index, row['text']) for index, row in well_rows.iterrows()]
21     fine_occurrences = [(index, row['text']) for index, row in fine_rows.iterrows()]
22     good_occurrences = [(index, row['text']) for index, row in good_rows.iterrows()]
23     great_occurrences = [(index, row['text']) for index, row in great_rows.iterrows()]
24     right_occurrences = [(index, row['text']) for index, row in right_rows.iterrows()]
25     imean_occurrences = [(index, row['text']) for index, row in imean_rows.iterrows()]
26
27     # save the occurrences in the dict
28     wells[df_name] = well_occurrences
29     fines[df_name] = fine_occurrences
30     goods[df_name] = good_occurrences
31     greats[df_name] = great_occurrences
32     rights[df_name] = right_occurrences
33     imeans[df_name] = imean_occurrences
```

1 goods

'SBC001.trn': [(24, "(H) ... and I'm not as good,"),
use you don't want to .. cripple up a .. (H) really go
good stereo.'], (490, "[(H) That's pretty] good,"), (8
good .. lambada dancer?'), (1487, '... Well that was g
Good good.'], (111, '... that sounds good.'], (410, "...
it's all fine [and good ge[2neration2]
[2That's2] good.'], (937, "... How that carrot's good.'
at looks good.'], (1003, "... that's good.'], (1098, '...
are so good leftover '... that would be good.'])
eally good.'], (1476, '... That looks good.']) 'SBC00
t's good is] .. hibiscus cooler.'], (28, '...
morni=ng,'], (331, 'good bye=,'], (610, "... he's real
s,'], (852, '(H) And then give those p=asses out for g
et really good,']), 'SBC005.trn': [(65, 'k=- so good g
good time Saturday,'], (542, "that's a good [spot for
ood friend herself.'], (1167, 'I2] had a good time wit
ew friends,'], (1226, '(H) was his good buddy.'], (141
[That's good P>].']), 'SBC007.trn': [(93, 'a real good
good idea for you to go up there in the winter.']), 'S

Keys: file name
Values: list of tuples
(line #, text)

Isolating indices and exploring context

```
1 indices = {key: [tpl[0] for tpl in value] for key, value in goods.items()}
2 print(indices)
```

```
{'SBC001.trn': [24, 40, 496], 'SBC002.trn': [236, 490, 874, 940, 1407], 'SBC003.trn': [768, 932, 935, 947, 1003, 1098, 1263, 1430, 1433, 1453, 1476], 'SBC004.trn': [1, 864], 'SBC005.trn': [65], 'SBC006.trn': [179, 542, 721, 744, 1167, 1206, 1226, 128, 8], 'SBC008.trn': [1029, 1488], 'SBC009.trn': [688], 'SBC010.trn': [766, 993], [02, 652, 755, 849], 'SBC013.trn': [247, 339, 382, 786, 787, 854, 865, 968, 1232, 7, 1789, 1813, 1821, 1994, 1996, 2207, 2234], 'SBC014.trn': [493, 617, 641, 657, 764], 'SBC016.trn': [3, 13, 81, 386, 388, 523, 617, 694, 1070, 1135, 1148, 1417, C017.trn': [], 'SBC018.trn': [4, 106, 163, 176], 'SBC019.trn': [96, 247, 251, 29190], 'SBC022.trn': [5, 23, 316, 336, 366, 367, 404, 405, 418, 453], 'SBC023.trn': [1353, 1391, 1394], 'SBC024.trn': [80, 742, 869, 871], 'SBC029.trn': [58, 67, 251, 8, 799, 981, 982, 1011, 1082, 1097], 'SBC031.trn': [45, 47, 162, 191, 291, 346, 8, 1489, 1510, 1511], 'SBC032.trn': [84, 345, 346, 582, 583, 849, 1306, 1368, 1384, 34, 338, 393, 508, 522], 'SBC034.trn': [26, 360, 504, 553], 'SBC035.trn': [157, 619, 814, 1054, 1118, 1143, 1201, 1226, 1293], 'SBC036.trn': [10, 53, 108, 378, 797], 'SBC037.trn': [86, 428, 528, 579, 629, 644, 663], 'SBC042.trn': [457, 679, 47, 310, 311, 413, 442, 444, 488, 550, 847, 855, 1001, 1012, 1015, 1016, 1038, 1485], 'SBC044.trn': [113, 128, 129, 194, 266, 440, 448, 1348, 1366, 1394], 'SBC045.trn': [156, 157, 328, 347, 896, 919, 922, 1010, 1058, 1071, 1100, 1103, 1109, 1157], [96, 414, 415, 417, 838, 842, 908, 926, 1000, 1002, 1011, 1025], 'SBC049.trn': [1182], 'SBC050.trn': [19, 29, 90, 117, 175, 186, 273, 483, 486, 722, 798, 799, 8, 1229, 1509, 1669], 'SBC056.trn': [89, 273, 282, 344, 345, 346, 411, 522, 533, 619, 229, 1238, 1470], 'SBC057.trn': [117, 193, 300, 330, 526, 716, 967, 974, 990], [9, 564, 751, 863], 'SBC059.trn': [567, 569, 625, 628, 637, 696, 700, 713, 715, 716, 7384, 1607, 1637], 'SBC060.trn': [652, 654, 672, 800, 808, 995, 996]}
```

```
targets = [likes, loves, wells, fines, goods, greats, rights, imeans]
targets_str = ['likes', 'loves', 'wells', 'fines', 'goods', 'greats', 'rights',
count = 0
```

```
for target in targets:
    # establish indices
    indices = {key: [tpl[0] for tpl in value] for key, value in target.items()}

    # write out file
    with open(targets_str[count]+"_output.txt", 'w') as file:
```

```
import sys
sys.stdout = file
```

```
for index in indices:
    df = dfs[index]
    print(index)
    key = index
    values = indices[key]
    # give 20 lines of context
    for v in values:
        first = v - 10
        last = v + 10
        print(df.iloc[first:last])
```

```
count+=1
```

```
sys.stdout = sys._stdout
```


Output files

	time	speaker	text
983	842.52 843.37		... [Yeah <X I know X>],
984	842.94 845.04	PHIL:	[(H)] !Jack and I and !Jim .. met with him.
985	845.04 845.34	BRAD:	.. Oh,
986	845.34 845.94		you mean his,
987	845.94 847.86		.. !Jack's [financial] friend that --
988	846.63 847.25	PHIL:	[!Jack's friend].
989	847.86 848.81	BRAD:	.. is retired.
990	848.81 849.26	PHIL:	Yeah.
991	849.26 850.06		.. [He's a] banker,
992	849.23 849.58	BRAD:	[Yeah].
993	850.06 851.41	PHIL:	.. and he'd be good on one hand.
994	851.41 851.78		.. [But],
995	851.56 851.81	BRAD:	[Yeah].
996	851.81 852.50	PHIL:	I would like,
997	852.50 854.37		(H) ideally XX I'd want em both.
998	854.37 854.87	BRAD:	.. Yeah.
999	854.87 855.54		... [Yeah].
1000	855.38 855.80	BRAD:	[You know].
1001	855.80 856.25		.. And,
1002	856.25 858.29		... [I'm hoping],

	time	speaker \
486	532.42 533.20	
487	533.20 534.27	
488	534.27 535.29	
489	535.29 536.80	
490	536.80 538.15	
491	538.15 539.39	
492	539.39 540.01	
493	540.01 541.32	
494	541.32 543.29	
495	543.29 544.42	
496	544.42 547.07	
497	547.07 547.37	
498	547.37 548.37	
499	548.37 550.02	
500	550.02 550.92	
501	550.92 552.07	
502	552.07 552.87	
503	552.87 553.82	
504	553.82 554.04	
505	554.04 554.36	

Some got
disjointed, but the
indices are correct

	text
486	they have to go,
487	.. a long ways to go get em,
488	like back East somewhere,
489	to get these .. horse hooves.
490	(H) For the college.
491	(H) They go back % .. East,
492	and they get em.
493	and they freeze em you know?
494	.. (H) So we have this frozen horse hoof,
495	that we have to start out on,
496	cause you don't want to .. cripple up a .. (H)...
497	and like,
498	my first hoof,
499	(H) .. that horse would have been,
500	.. lame,
501	@ @ <@ like cra=zy.
502	@ @ (H) But @>,
503	(H) it was just,
504	and,
505	oh,

03

Findings

Transcript – “fine”

	time_start	time_end	speaker \	text
41	58.688	59.596	MELISSA:	[One side] of a page?
42	59.596	60.598		It takes me a long time,
43	60.598	62.461		because I've got to go over the sentences,
44	62.461	63.323		(H) figure out,
45	63.323	65.768		.. if I'm gonna rewrite them or leave them the...
46	65.768	66.909		(H) [and just] write them out.
47	65.768	66.280	JAN:	[Well y-] --
48	66.909	68.772	MELISSA:	I can't write them exactly the way they are,
49	68.772	69.681		because they stink.
50	69.681	71.730	JAN:	.. Then you need to go downstairs and finish it.
51	71.730	72.755	MELISSA:	(H) I'm fine.
52	73.989	74.781	FRANK:	... ~Melissa,
53	74.711	75.829		[it's nine o'clock].
54	74.804	76.807	MELISSA:	[I'm not gonna do any better] downstairs.
55	76.761	78.880	FRANK:	It's [2nine o'clock .. in the evening2].
56	77.063	79.509	JAN:	[2You also have algebra2] to do.
57	79.509	80.953	MELISSA:	.. I can skip algebra.
58	80.953	82.024	FRANK:	... No you [can't].
59	81.465	82.024	JAN:	[No].
60	82.024	82.420		.. [2you- (Hx)2] --

SBC019.trn: *Doesn't Work in this Household*

"A family conversation, recorded in Michigan. Frank and Jan (a married couple) are talking with Ron--Jan's brother who is visiting from California. Brett and Melissa are Frank and Jan's junior-high-age children, who are doing homework and also taking part in the conversation."

Transcript – “I mean”

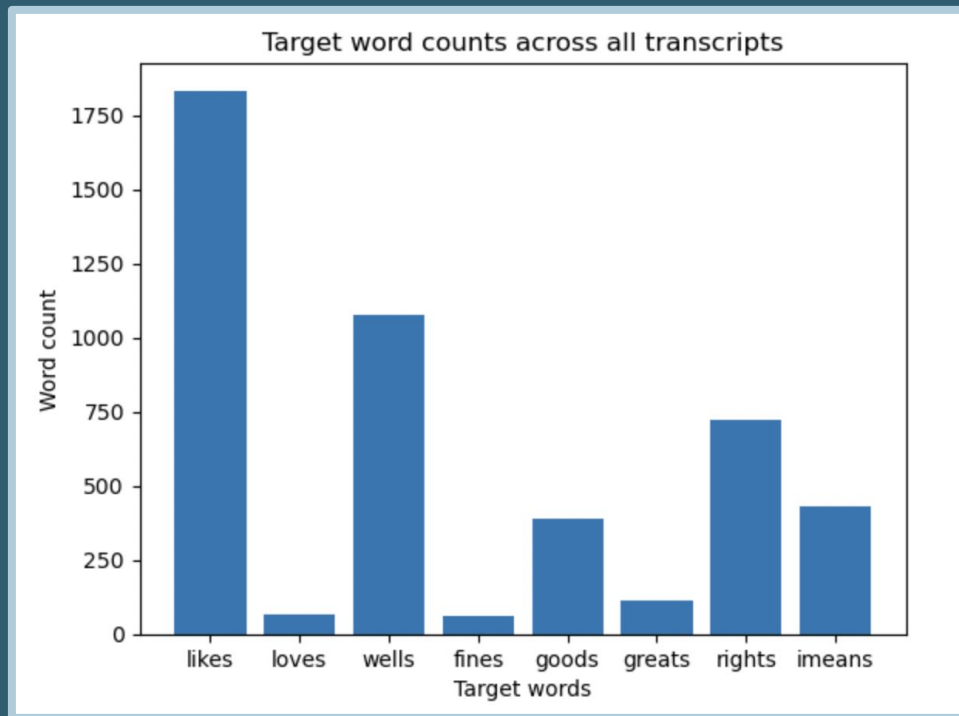
SBC002.trn

	time	speaker	text
45	45.91 47.53	MILES:	and I guess he really goes fa=st.
46	47.53 47.86		[<X And X>],
47	47.59 47.94	JAMIE:	[Yeah].
48	47.94 48.49	MILES:	(H) he k- --
49	48.49 50.09	JAMIE:	... He does[n't explain anything precisely].
50	48.69 52.20	MILES:	[He has to double it dow=n to] like...
51	52.20 53.31		before they can g=- pick it up-,
52	53.31 54.23		... @@
53	54.23 55.91	HAROLD:	Well I'm sure !Thomas is all over it.
54	55.91 61.26	JAMIE:	... Prob[ably XX] [2XXX2] --
55	59.81 60.56	HAROLD:	[I mean he] [2has a bro-2] --
56	60.26 61.36	MILES:	[2XXXX could have2] see=n him.
57	61.36 63.58	HAROLD:	I guess that means his broken leg is [3@doing ...
58	62.97 63.86	PETE:	[3I was w...
59	63.86 65.25		I was imagining [4he had broke an arm4] or som...
60	64.24 64.99	JAMIE:	[4<HI Oh yeah= HI>4].
61	65.25 66.05	PETE:	But it was his leg?
62	66.05 66.70	HAROLD:	.. Yeah[=].
63	66.35 66.76	PETE:	[That's like],
64	66.76 68.71		.. <X I guess X> that he was being hauled arou...

SBC002.trn: *Lambada*

"After-dinner conversation among four friends in San Francisco, California. Participants are in their late twenties or early thirties. Harold and Jamie are a married couple, Miles is a doctor, and Pete is a graduate student from Southern California."

Figure



04

Reflection and Future Steps

Reflection

→ What went well

- ◆ Topic
- ◆ Theoretical basis

→ What could've been better

- ◆ Data processing
- ◆ Time management

→ What I learned

- ◆ Licensing
- ◆ Python libraries

Future Steps

- Analyze more of the data
- Do value add
- Make more figures
- Draw concrete conclusions



Thanks!

Questions?

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