Speaker recognition using keywords and key hypernyms

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Introduction

Question

 Can we recognize the speaker in a spoken corpus using keywords and hypernyms?

???: I love my job at the <u>Krusty Krab</u>.

I like <u>Jelly Fishing</u> and <u>Bubble Blowing</u>.

I've never been late for work!



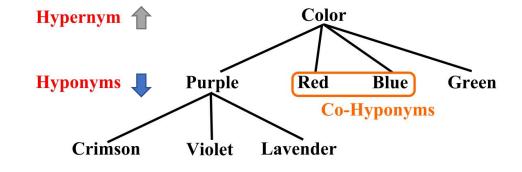
KEYWORDS VS KEY HYPERNYM

KEYWORDS

- Comparing the word frequency.
- Noun, Pronoun, Verb, Adjective, Adverb

KEY HYPERNYM

- Abstract and conceptual words, key content.
- Not about the frequency but the diversity of hyponyms.
- We have to know synonym sets (e.g. Is, Are, Was, Were → Be.v)
- Noun (Normally Verbs don't have enough hypernyms to construct a network)

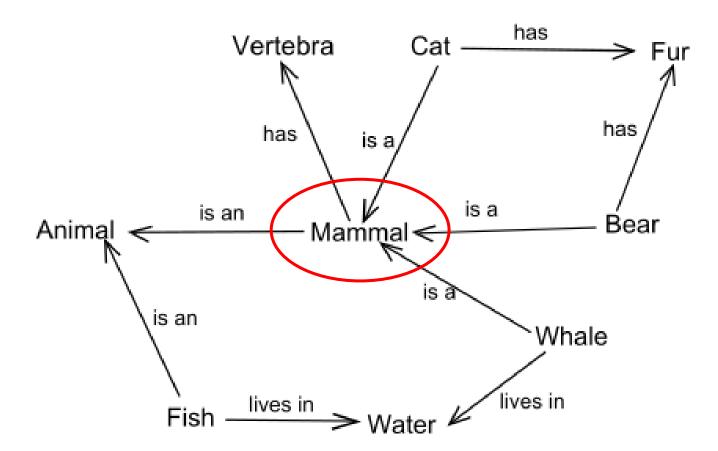


Used NLTK Tagger and TAGANT for POS Tagging

```
>>> text = word_tokenize("They refuse to permit us to obtain the refuse permit")
>>> nltk.pos_tag(text)
[('They', 'PRP'), ('refuse', 'VBP'), ('to', 'TO'), ('permit', 'VB'), ('us', 'PRP'),
('to', 'TO'), ('obtain', 'VB'), ('the', 'DT'), ('refuse', 'NN'), ('permit', 'NN')]
```

Characteristics of Key Hypernym

- Likely to have many linked hyponyms.
- Using different words in the similar category will increase the keyness.



Spoken Corpus

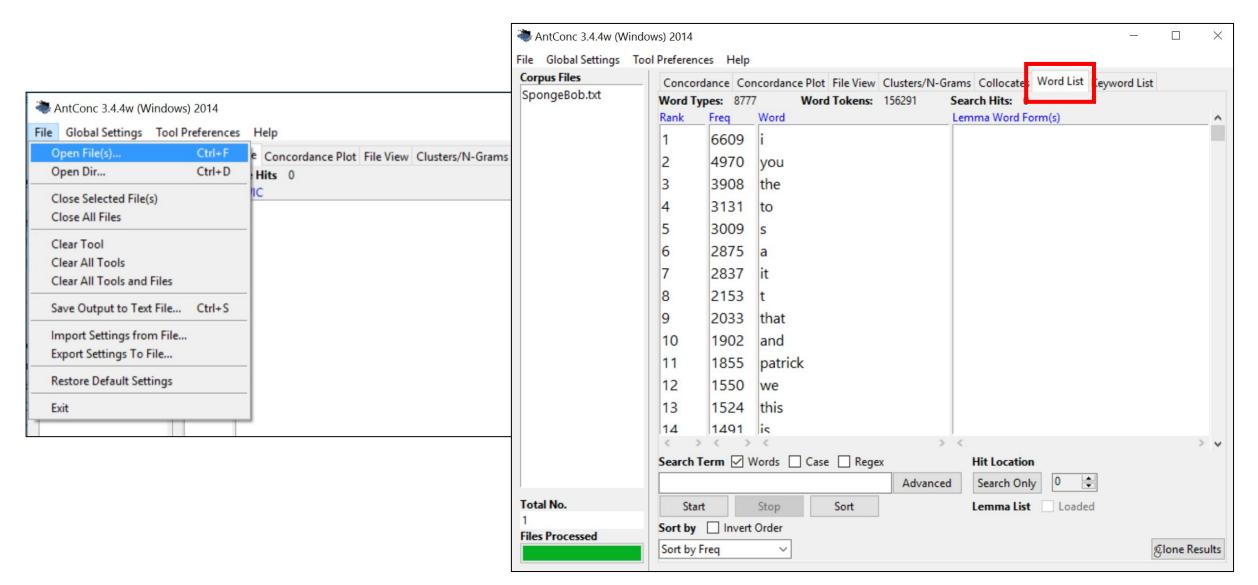
- http://spongebob.wikia.com/wiki/List_of_transcripts
- Transcripts of "SpongeBob"
 - SpongeBob: [Jumps on the diving board] Look at me, I'm... [Jumps up, and leaves his underwear behind] ...naked! [Lands inside pants, walks over to exercise room. His head pops out of the top of his pants]

 Gotta be in top physical condition for today, Gary.
 - . Gary: Meow.
 - SpongeBob: [He goes inside his small gym that has a sign that says, "I Love Pain." Taking deep breaths, he prepares to lift a barbell that is balanced by two lightweight stuffed animals. He sticks out his chest, but almost passes out because he can barely lift it. He drops it, and it makes a 'squeak' noise] I'm ready! [Runs outside] I'm ready, I'm
 - Patrick: Go, SpongeBob! [Patrick falls] Whoa! [Crash!]
 - SpongeBob: [Runs down the street to the Krusty Krab] There it is. The finest eating establishment ever established for eating. The Krusty Krab, home of the Krabby Patty. With a 'Help Wanted' sign in the window! For years I've been dreaming of this moment! I'm gonna go in there, march straight to the manager, look 'im straight in the eye, [breaks the fourth wall and looks the audience in the eye] lay it on the line and... I can't do this! [He starts to run home but Patrick stops him] Uh, Patrick!
 - Patrick: Where do you think you're going?
 - SpongeBob: I was just.
 - Patrick: No you're not. You're going to the Krusty Krab and get that job!
 - SpongeBob: I can't, don't you see? I'm not good enough!
 - Patrick: Whose first words were "may I take your order"?
 - SpongeBob: Mine were.
 - Patrick: Who made a spatula out of toothpicks in wood shop?
 - SpongeBob: I did.
 - Patrick: [Grimaces and contorts twice while trying to come up with a good third line] Who's a, uh who's uhh, oh!
 Who's a big yellow cube with holes?
 - SpongeBob: I am!
 - Patrick: Who's ready?

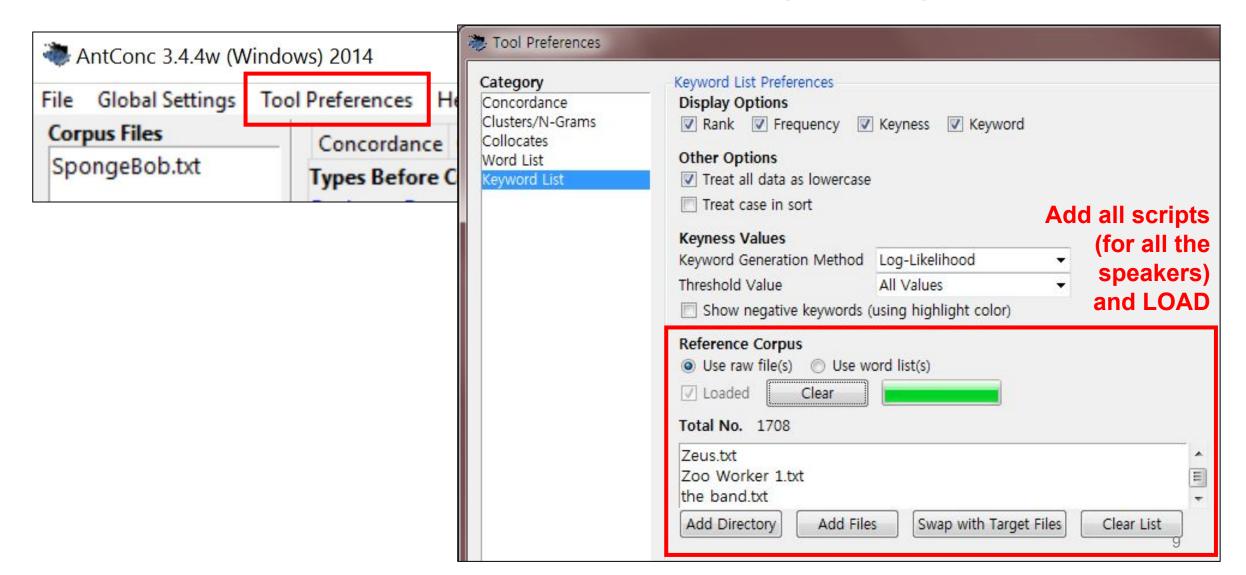
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8a Sandy's Rocket View trans	7b	Jellyfish Jam	View transcr
1.1 PM 10 1 PM 10	8a	Sandy's Rocket	View transcr
	9a	Nature Pants	View transcr

Keyword Approach

METHOD – KEYWORDS



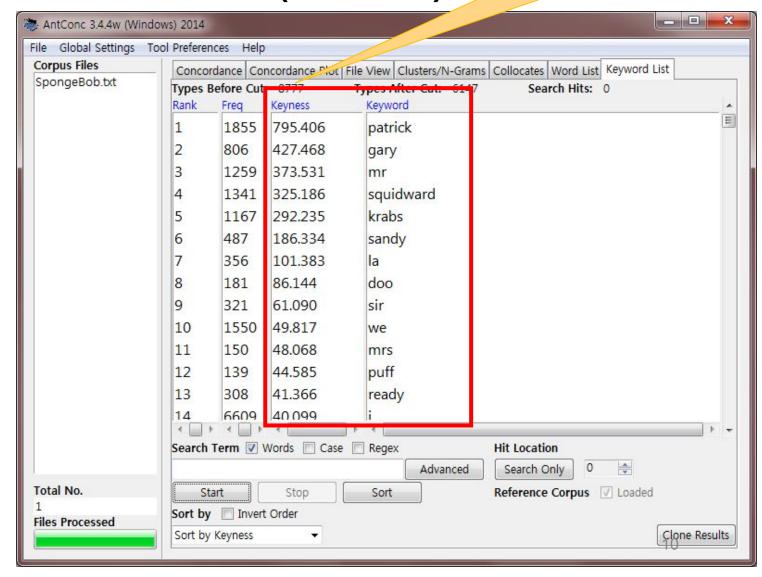
METHOD - KEYWORDS (cont.)



What is Keyness?

METHOD – KEYWORDS (Cont.)

- Sort by <u>Keyness</u>:
 - What is keyness?



METHOD - KEYWORDS (Cont.)

Help >> View Readme Help File

Keyword List

This tool shows the which words are unusually frequent (or infrequent) in the corpus in comparison with the words in a reference corpus. This allows you to identify characteristic words in the corpus, for example, as part of a genre or ESP study.

The following steps produce a keyword list and demonstrate the main features of this tool.

- 1) Select a set of target files.
- 2) Go to the 'Preferences' menu and chose the 'Keyword Preferences' option.
- Choose the keyword generation method (a statistical measure) to calculate the 'keyness' of the target file words. The default setting of Log Likelihood is recommended. When using either Log Likelihood or Chi-squared as the statistical measure, the following significance values apply (see:

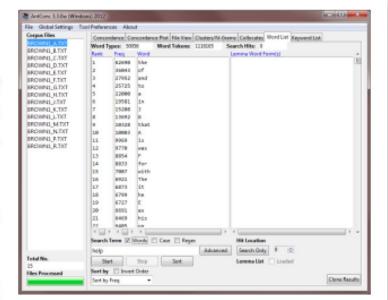
http://ucrel.lancs.ac.uk/llwizard.html):

95th percentile; 5% level; p < 0.05; critical value = 3.84

99th percentile; 1% level; p < 0.01; critical value = 6.63

99.9th percentile; 0.1% level; p < 0.001; critical value = 10.83

99.99th percentile; 0.01% level; p < 0.0001; critical value = 15.13



METHOD - KEYWORDS (Cont.)

http://ucrel.lancs.ac.uk/llwizard.html

How to calculate log likelihood

Log likelihood is calculated by constructing a contingency table as follows:

table as lengths.				
	Corpus 1	Corpus 2	Total	
Frequency of word	а	b	a+b	
Frequency of other words	c-a	d-b	c+d-a-b	
Total	С	d	c+d	

Note that the value 'c' corresponds to the number of words in corpus one, and 'd' corresponds to the number of words in corpus two (N values). The values 'a' and 'b' are called the observed values (O), whereas we need to calculate the expected values (E) according to the following formula:

$$E_i = \frac{N_i \sum_i O_i}{\sum_i N_i}$$

In our case N1 = c, and N2 = d. So, for this word E1 = $c^*(a+b) / (c+d)$ and E2 = $d^*(a+b) / (c+d)$. The calculation for the expected values takes account of the size of the two corpora, so we do not need to normalize the figures before applying the formula. We can then calculate the log-likelihood value according to this formula:

$$-2\ln \lambda = 2\sum_{i} O_{i} \ln \left(\frac{O_{i}}{E_{i}}\right)$$

This equates to calculating log-likelihood G2 as follows G2 = 2*((a*ln (a/E1)) + (b*ln (b/E2)))

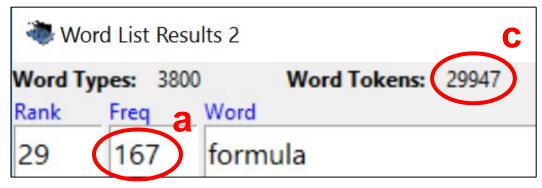
METHOD – KEYWORDS (Cont.)

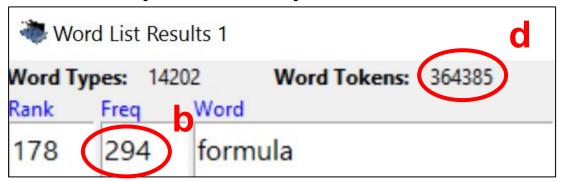
[token]	Corpus 1	Corpus 2
Frequency of Word	а	b
Corpus size	C	d

Summary:

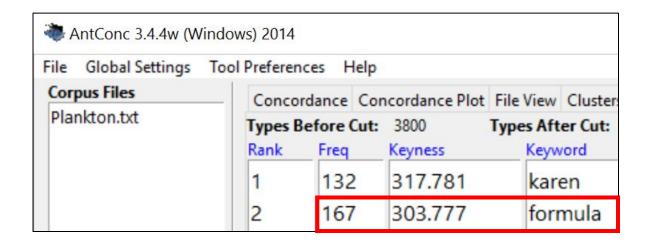
- E1 = c*(a+b) / (c+d)
- E2 = d*(a+b) / (c+d)
- Keyness = 2*((a*In (a/E1)) + (b*In (b/E2)))

METHOD - KEYWORDS (Cont.)





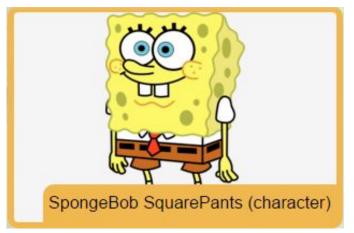
Verification



```
>>> def keyness(a, b, c, d):
        a = float(a)
        b = float(b)
        c = float(c)
        d = float(d)
        E1 = c*(a+b) / (c+d)
        E2 = d*(a+b) / (c+d)
        ka = (a*math.log(a/E1))
        kb = (b*math.log(b/E2))
        return 2*(ka+kb)
>>> kevness(167. 294, 29947, 364385)
303.7765602866808
```

Not showing: Pronouns(Patrick, Gary, Mr. Krabs), Interjection(oh, wow)

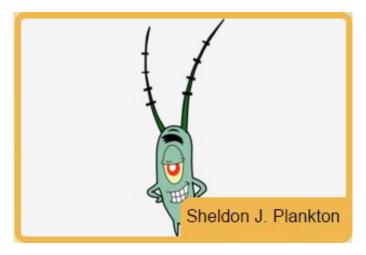
Keyword Result



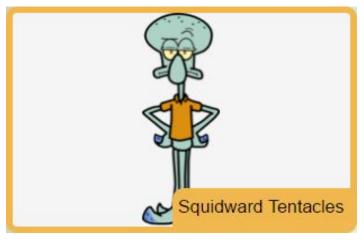
ncy	Rank	Keyness	Word
	13	41.366	ready
	18	27.577	sorry
	19	25.872	guess
	20	24.558	best
	21	23.572	friend
	22	20.901	okay
	25	18.692	jellyfish
	26	17.694	worry
	34	13.799	spatula
	36	12.487	buddy



Rank	Frequency	Keyness	Word
1	354	467.163	money
2	408	338.615	boy
3	1027	226.042	me
4	90	149.547	lad(젊은이)
5	105	115.565	customers
6	162	84.869	ya
7	78	68.870	boys
10	138	42.642	patties
11	44	39.369	dollar
13	80	37.533	free



Rank	Frequency	Keyness	Word
2	167	341.692	formula
3	98	172.930	chum
4	102	130.935	secret
5	50	82.089	bucket
7	26	59.169	wife
8	47	56.949	mine
9	36	55.701	plan
10	122	53.999	patty
11	33	52.922	recipe
12	34	51.706	steal



Rank	Frequency	Keyness	Word
2	611	94.085	no
3	174	75.149	two
5	748	57.677	what
7	45	45.993	art
8	37	43.834	clarinet
9	23	39.927	morons
10	26	37.625	squilliam
12	42	35.634	whatever
13	43	31.021	stupid
14	30	30.996	quiet



Word	Keyness	Frequency	Rank
critter	74.906	20	6
karate	56.301	28	7
air	56.189	27	8
rodeo	56.160	17	9
critters	51.327	13	10
nuts	45.009	16	14
experiment	43.866	12	15
science	36.561	14	16
tarnation	31.115	8	18
helmet	29.212	11	20



Word	Keyness	Frequency	Rank
meow	5156.911	521	1
mooowww	40.943	4	2
reow	30.707	3	3
mah	29.467	3	4
meoooow	28.446	3	5
meowow	20.471	2	6
moooowww	20.471	2	7
mrloooow	20.471	2	8
			• • •
20			

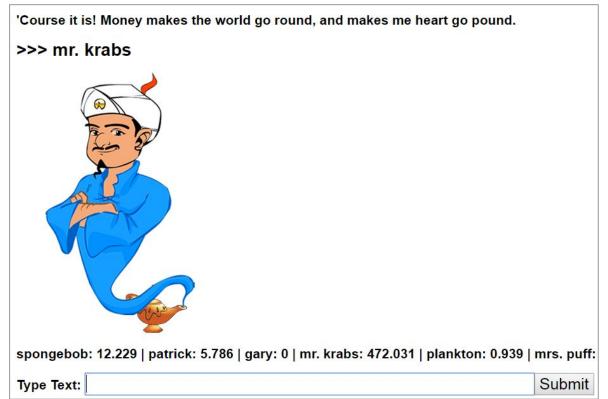
METHOD - KEYWORDS (Cont.)

- Keyness shows the importance of the word in the text, based on the impact in the full text.
- So, I used this metric to construct the speaker recognition system.
 - Score for each token.
 - Sum up and find result.

```
::: I never been late for work.
['I', 'never', 'been', 'late', 'for', 'work', '.']
why?
     gary 0
why? mr. krabs 15.359
why?
     mrs. puff 6.877
why?
     narrator 0
why?
      patrick 0.081
why?
      plankton 3.148
why?
      sandy 0
     spongebob 17.839
why? squidward 2.327
spongebob 17.839
```

LIVE DEMO

- Gets score when the word hits.
- Ignore Names (e.g. SpongeBob, Patrick, Krusty Krabs ...)
- Ignore **Stop words** (e.g. i, me, my, myself, you, yourself ...)
- Run!



LIVE DEMO

Test sets:

Plankton: Why couldn't I see it before? The way to get the Krabby Patty formula was so obvious! Spend an inordinate amount of time training several dozen sea bears to take over your restaurant and force you to give it up! can turn them from their central purpose!

SpongeBob: Yoo-hoo! Who wants their tummies tickled?

Plankton: No... My weapons! Ouch!

SpongeBob: Sea bears aren't weapons, Plankton. They're furry buckets of love. See? And what do sea bears love more than

tummy tickles? Jellyfish honey!

SpongeBob: Come and get it!

Plankton: No! Come back!

Mr. Krabs: Why do you keep doing this, Plankton?

Plankton: Heh-heh-heh...

Mr. Krabs: When you mess with me business, ya mess with me money!

Plankton: Er, money's not everything, you know.

Mr. Krabs: 'Course it is! Money makes the world go round, and makes me heart go pound.

Higher Value: Strong Identity

Testing Result

Character	Correct	Total	Percentage
SpongeBob	3150	13348	23.60 %
Gary	415	420	98.81 %
Mr. Krabs	2168	4812	45.05 %
Patrick	2421	5206	46.50 %
Plankton	848	2164	39.19 %
Sandy	517	1403	36.85 %
Squidward	1416	4796	29.52 %

However, POS keyword approach failed

- Tagging the words, categorizing into Nouns, Pronouns, Verbs,
 Adverbs, Adjective, and analyzing keyness was not effective.
- Tagger problem
 - Lots of scripts were difficult to tag for the program.
 - MAN OVERBOARD! Climb, Mr. Squidward! Climb!
 - N N NP NP N

 - VVG RP VVG RP NP RB
 - Both NLTK tagger and TAGANT was not working correctly.
 - e.g. SpongeBob was categorized into noun, verb, adverb...
 - Imperative (명령형) sentences are all categorized into noun...

Hypernym Approach

Hypernym in NLTK

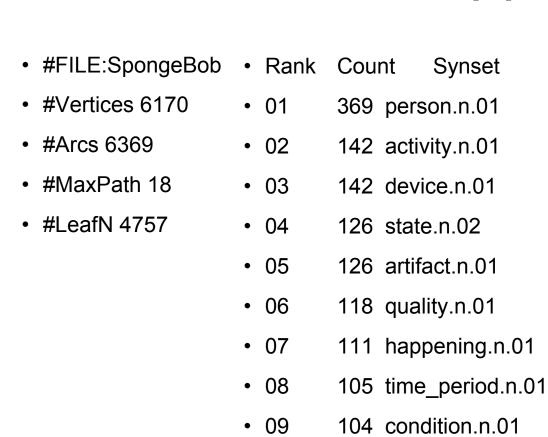
- 1. Tokenize the sentence and **pos tag**. (possible tagger problem)
 - a. words = nltk.pos_tag(nltk.word_tokenize(line))
- 2. Consider the word which is 'NN' tag: Noun.
- 3. Find synsets (synnonym sets).
 - a. syns = wn.synsets(w, pos=wn.NOUN)
 - b. e.g. is, was, am, are \rightarrow "be.v.01, be.v.03 ..."
- 4. Choose the first synset.
 - a. ws = syns[0]
- 5. Follow up the hypernym path.
 - a. ws.hypernym_paths()[0] ← sometimes many paths are available
- 6. Create the network.

Hypernym in NLTK

- 1. Testing: I never been late for work.
- 2. >>> nltk.pos_tag(nltk.word_tokenize('I never been late for work.'))
 [('I', 'PRP'), ('never', 'RB'), ('been', 'VBN'), ('late', 'JJ'), ('for', 'IN'), ('work', 'NN'), ('.', '.')]
- >>> wn.synsets('work', pos=wn.NOUN)
 [Synset('work.n.01'), Synset('work.n.02'), Synset('study.n.02'), Synset('work.n.05'), Synset('workplace.n.01'), Synset('oeuvre.n.01')]
- >>> wn.synsets('work', pos=wn.NOUN)[0].hypernym_paths()[0]
 [Synset('entity.n.01'), Synset('abstraction.n.06'),
 Synset('psychological_feature.n.01'), Synset('event.n.01'), Synset('activity.n.01'),
 Synset('activity.n.01'), Synset('work.n.01')]
- 5. Using the hypernym path, draw the network.

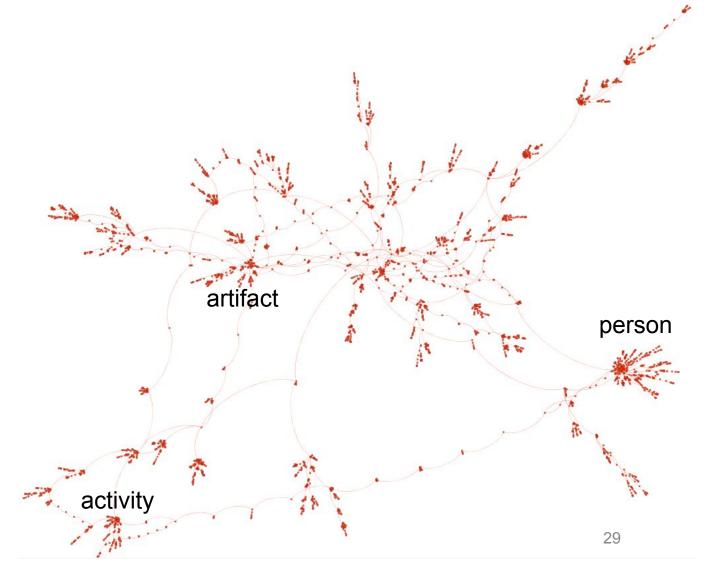
HYPERNYM Approach (Nouns)

material.n.01



• 10

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HYPERNYM Approach (Verbs)

Verbs don't cluster that much.

#FILE:SpongeBob	 Rank 	Count Synset		×	
• #Vertices 2130	• 01	205 change.v.01	73 .		
• #Arcs 1916	• 02	173 travel.v.01			*
• #MaxPath 12	• 03	122 change.v.02		N.	7-
• #LeafN 3504	• 04	101 be.v.01			*
	• 05	98 move.v.02	***	•	•
	• 06	77 move.v.03		***	(4)
	• 07	67 act.v.01	à*	<i>f</i> • • • • • • • • • • • • • • • • • • •	
	• 08	60 make.v.03			8
	• 09	55 inform.v.01		76	¥
	• 10	53 communicate.v.02	+		*

Keyness? (Nouns)

```
while True:
   V = []
           '''I love my job at the Krusty Krab.
 like Jelly Fishing and Bubble Blowing. I've never been late for work!'''
     sent = raw_input("Chat: ")
    sent = ssss
    words = nltk.pos_tag(nltk.word_tokenize(sent))
    score = {}
    for fname in file_names:
        score[fname] = 0
    for (w, tag) in words:
        w = w.lower()
        if tag == 'NN':
            syns = wn.synsets(w, pos=wn.NOUN)
            if len(syns) > 0:
                ws = syns[0]
                for path in ws.hypernym_paths()[0]:
                    addToList(V, path.name())
    for v in V:
        for fname in file_names:
            if ddic[fname].has_key(v):
                score[fname] += mykeyness(fname, v)
    for who in sorted(score, key=score.get, reverse=True):
        print who, score[who]
    if sent == 'quit()': break
    break
```

I tested many other sentences, but they didn't show the speaker well.

```
>>>
= RESTART: C:\Users\user\[E\]
Squidward 7817.31940878
SpongeBob 5105.76130151
Sandy 3638.24666524
Plankton 2343.77224926
Patrick 1178.81146147
Mr. Krabs 87.6181579499
Gary 70.8001529575
```

Failures

- Tagger Problem
 - Not working correctly, especially for imperative form.
- Synset and Hypernym Path problem
 - In fact, we should choose the right synset and hypernym path manually.
- Hypernym Keyness Speaker Detector
 - Difficult to test by each sentence. (Sentence has too few word to make an analizable hypernym network)
 - Difficult to find out the meaning from the graph.

Comparison with my previous project.

- I used Brown Corpus into those categories:
 Then tested with the real news articles (Chosun english news, Reuters)

- A. PRESS: Reportage (44 texts)
- B. PRESS: Editorial (27 texts)
- C. PRESS: Reviews (17 texts)
- D. RELIGION (17 texts)
- E. SKILL AND HOBBIES (36 texts)
- F. POPULAR LORE (48 texts)
- G. BELLES-LETTRES Biography, Memoirs, etc. (75 texts)

- H. MISCELLANEOUS: US Government & House Organs (30 texts)
- J. LEARNED (80 texts)
- K. FICTION: General (29 texts)
- L. FICTION: Mystery and Detective Fiction (24) texts)
- M. FICTION: Science (6 texts)
- N. FICTION: Adventure and Western (29 texts)
- P. FICTION: Romance and Love Story (29 texts)
- R. HUMOR (9 texts)

Conclusion

Conclusion

- Successfully built speaker recognition system.
 - Keywords were detected very well.
- For the spoken corpus, it is better to calculate the **keyness** of the word directly.
 - POS Tagging and categorization approach is inefficient.
- POS Tagging fails for spoken text in many case.
- Hypernym approach might be better for written corpus.
 - We don't use various words when we are speaking.
 - Mostly, the size of the spoken data is small to make network.
 - But it might be effective for written data, since we try to avoid to use same word frequently, which makes easy to recognize the key hypernym.

Future research

- NLTK also provides the chat corpus (NPS):
 - originally collected by the Naval Postgraduate School
 - The corpus contains over 10,000 posts, anonymized by replacing usernames with generic names of the form "UserNNN", and manually edited to remove any other identifying information.
 - The corpus is organized into 15 files, where each file contains several hundred posts collected on a given date, for an age-specific chatroom (teens, 20s, 30s, 40s, plus a generic adults chatroom)

Age Detection?

```
>>> from nltk.corpus import nps_chat

>>> chatroom = nps_chat.posts('10-19-20s_706posts.xml')

>>> chatroom[123]

['i', 'do', "n't", 'want', 'hot', 'pics', 'of', 'a', 'female', ',',

'l', 'can', 'look', 'in', 'a', 'mirror', '.']
```

Thank you!