Texts, Treebanks, Networks descriptions and analysis of languages

XINYING CHEN

University of Ostrava, Nov. 8, 2017

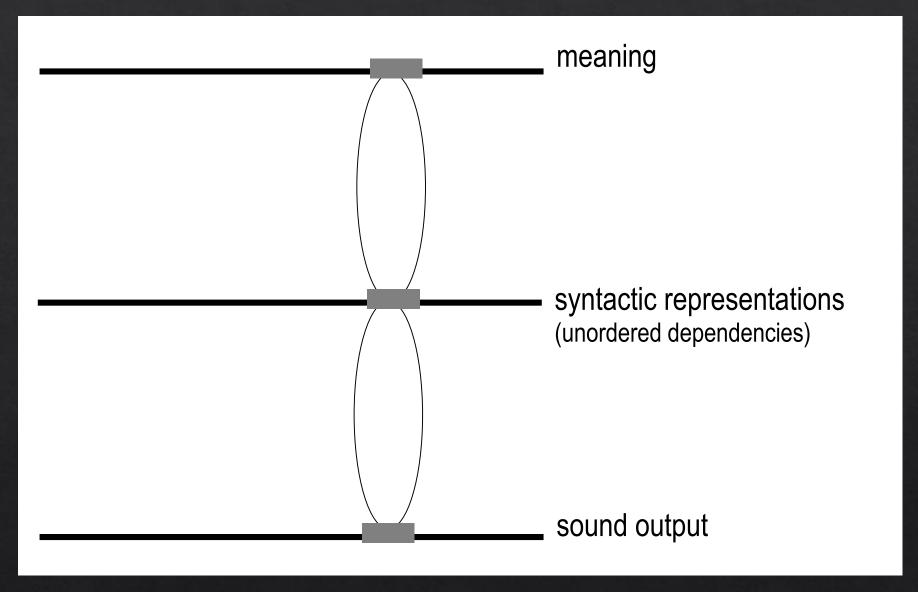
Outline

- Why different language descriptions and analysis
- ♦ Texts
- ♦ Treebanks
- Networks
- Questions & comments

Outline

Why different language descriptions and analysis

language is a system of signs, each of which is an arbitrary union of sound and meaning... Any given sign, is defined by its relationships with the others. (Saussure 1959)



Saussure's signified and signifier (2011)
Mel'čuk's meaning and text (1981)
Chomsky's logical and phonetic structure (2002)

Conceptual Layer (abstract, language-independent)

Semantic Layer (abstract, language-dependent)

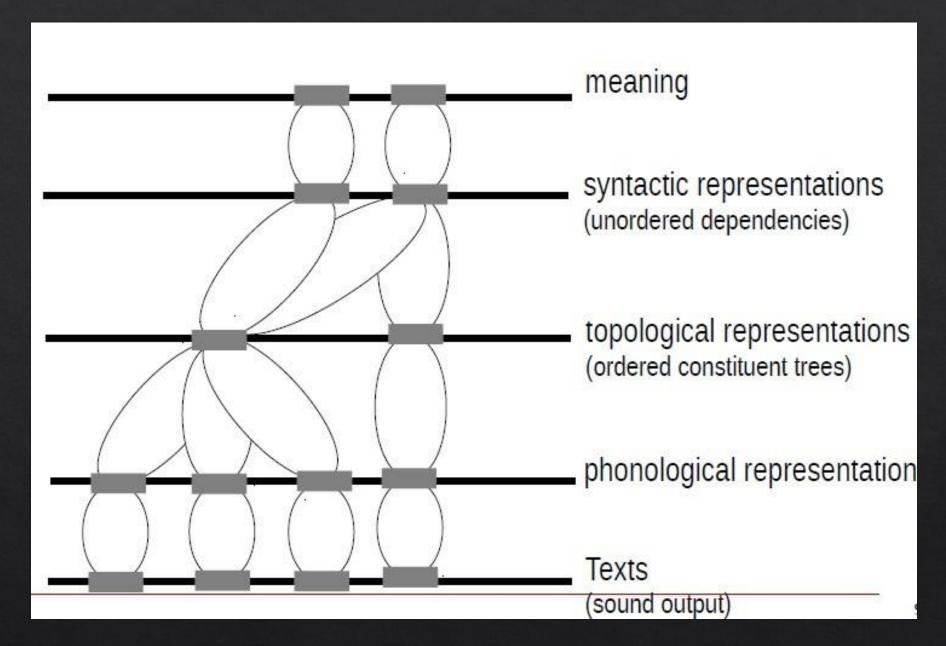
Deep-syntactic Layer (sentence structuring, lexicalization)

Surface-syntactic Layer (introduction of grammatical units)

Deep-morphological Layer (ordering, inflecting)

Surface-morphological Layer (morphologic interactions)

Text (full-fledged sentences)



Language understanding and generation require deconstructions or constructions of units on different language layers. Different language descriptions, therefore, yield to this nature of language communications.

Outline

- Why different language descriptions and analysis
- ♦ Texts

Texts

- You may have heard
 - word, token, type, lemma, phrase, ngram, sentence, paragraph...
 - word length, frequency, size...
 - ♦ genre, style, authorship...

Texts

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 - ⋄ word, token, type, lemma, phrase, ngram, sentence, paragraph...
 - word length, frequency, size...
 - ⋄ genre, style, authorship...
- Advantages
 - ♦ Relatively objective (word vs phrase...)
 - More available (OCR, Google books...)
 - ♦ Easier to test hypothesis (Zip'f law...)

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- Disadvantages
 - One dimensional description (linear modals)
 - Missing details of language understanding or generation

♦ Hypothesis

♦ The relationship between the frequency of Chinese characters and structural complexity of characters should be in accordance with Zipf's law due to the principle of least effort

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♦ Frequency

- Chinese Characters' Frequency Dictionary
 - the most frequent 3,061 different Chinese characters and their frequency in People's Daily, a famous newspaper in China, Corpus. (99.43%)

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⋄ Frequency

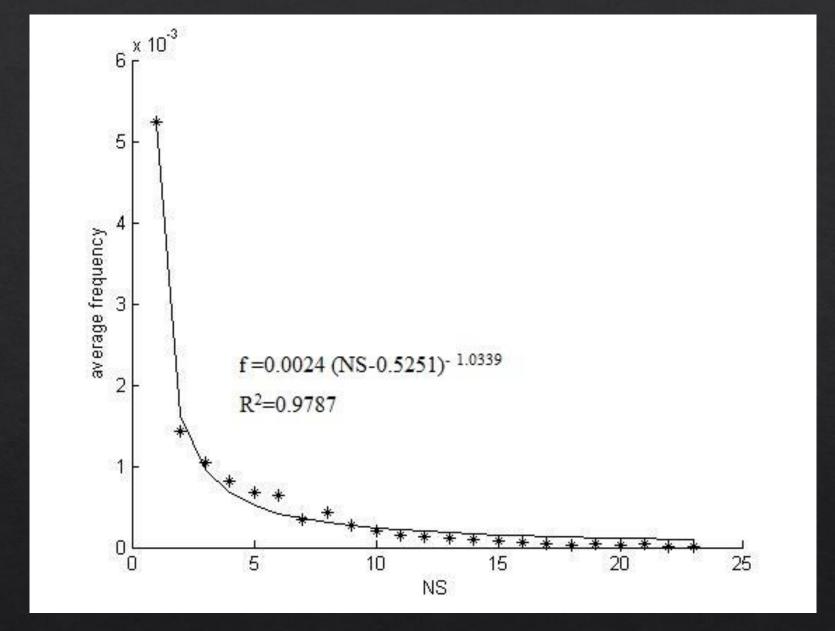
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♦ Structural Complexity

- Number of strokes
- Number of components
 - Both counted based on dictionaries and issued formal documents.

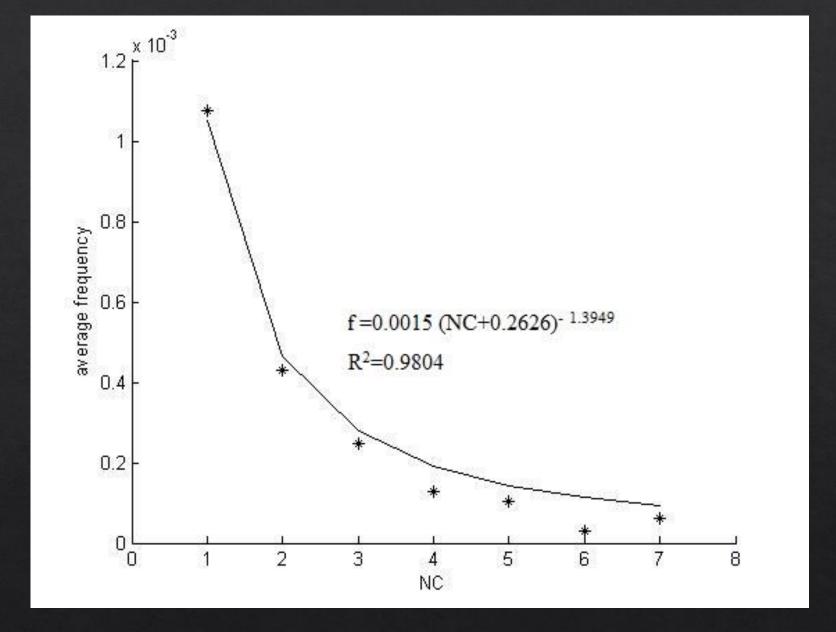


Number of Stroks	Number of Chracters	Examples	Cumulative frequency	Average frequency
1	2	一,乙	0.0104818889	0.0052409444
2	19	人,了,力	0.0270823814	0.0014253885
3	51	大,上,工	0.0535283278	0.0010495751
4	113	中,不,为	0.0924586324	0.0008182180
5	145	发,业,民	0.0989700908	0.0006825524
6	237	在,有,年	0.1514154083	0.0006388836
7	307	作,这,来	0.1035224004	0.0003372065
8	379	的,国,和	0.1611448260	0.0004251842
9	368	是,要,政	0.1006057580	0.0002733852
10	343	家,部,展	0.0675444718	0.0001969227
11	290	理,得,基	0.0437326729	0.0001508023
12	276	就,等,提	0.0366750136	0.0001328805
13	177	新,意,解	0.0215748566	0.0001218918
14	124	道,管,赛	0.0111661176	0.0000900493
15	100	题,增,德	0.0084463093	0.0000844631
16	55	整, 融,器	0.0033965338	0.0000617552
17	33	藏,繁,疑	0.0012677566	0.0000384169
18	11	翻,覆,藤	0.0002244518	0.0000204047
19	13	警,疆,攀	0.0005720697	0.0000440054
20	11	籍,灌,耀	0.0002580577	0.0000234598
21	4	露,霸,髓	0.0001694056	0.0000423514
22	2	囊,镶	0.0000231615	0.0000115808
23		罐	0.0000151569	0.0000151569



The relationship between NS and the average frequency of Chinese characters that share the same NS

Number of Components	Number of Characters	Examples	Cumulative frequency	Average frequency
1	187	一,中,人	0.2008274347	0.0010739435
2	969	的,国,和	0.4164280467	0.0004297503
3	1186	在,是,发	0.2922298614	0.0002463995
4	534	高,能,说	0.0682191084	0.0001277511
5	150	题,领,解	0.0153812537	0.0001025417
6	32	歌,疑,衡	0.0010030980	0.0000313468
7	3	疆,凝,颤	0.0001869466	0.0000623155



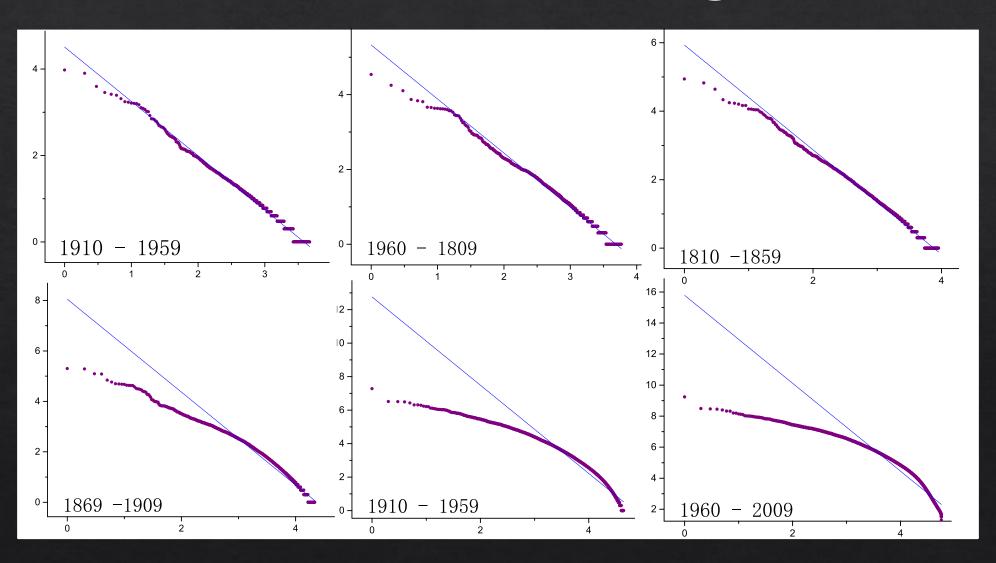
The relationship between NC and the average frequency of Chinese characters that share the same NC

- ♦ Observation
 - the lexicon changes of a language during a time period
- Hypothesis verification
 - There is a multi-syllabification trend of Chinese words during Chinese evolution (Chen, H., Liang, J., & Liu, H. 2015 PLoS ONE)

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- The Chinese Google 1-gram data (segmented-words frequency)
 - ♦ Symbols, Japanese, Roman letters, Arabic numerals
 - ♦ 1710 ~ 2009 (300 years)
 - ⋄ × Before 1710 (sparse, segmentation accuracy)
 - ♦ 6 * 50 (years)
 - ♦ 1710~59, 1760~1809, 1810~59, 1869~1909, 1910~59, and 1960~2009

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- ♦ Power law fitting

Power Law Fitting



Power Law Fitting: $y = a^*x^{-b}$

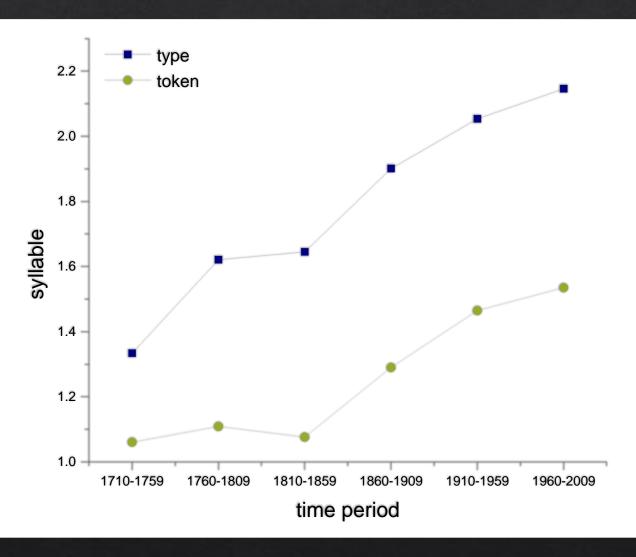
Time Period	а	b	R^2
1760-1809	34,791	-0.98122	0.98755
1810-1859	96,530	-0.94701	0.96039
1860-1909	249,946	-0.82656	0.92599
1910-1959	16,579,000	-1.09064	0.91207
1960-2009	1,512,270,000	-1.05617	0.90595

all the six frequency lists fit the Power Law well (the minimum R2 is 0.90595) despite the huge data scale (the largest list includes more than 21 billion tokens)

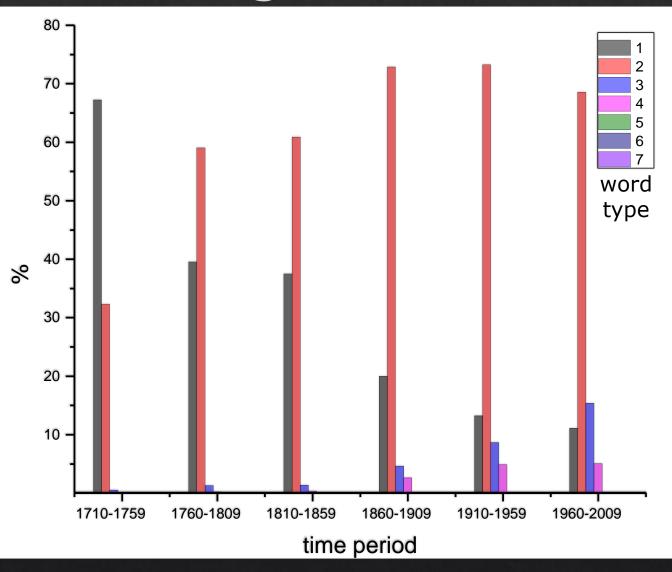
[♦] The Google Ngram data should be suitable for doing some linguistic studies

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Word length distribution



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- ♦ Less available
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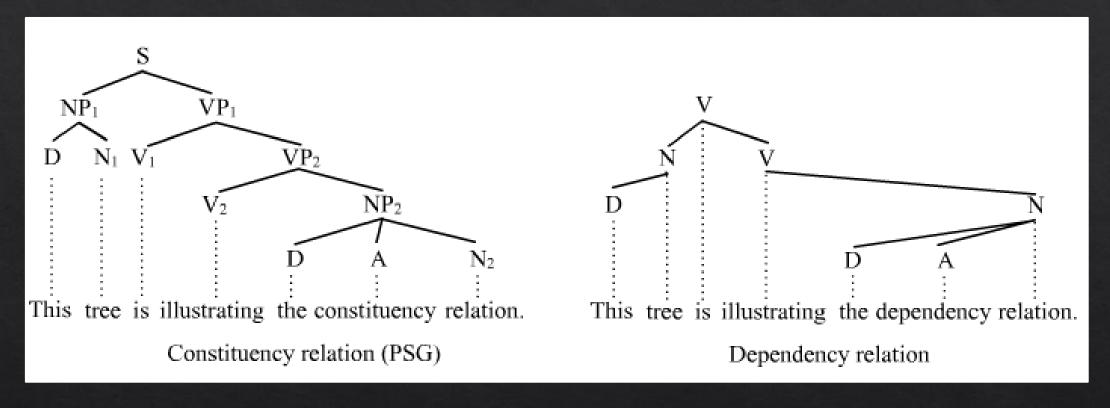
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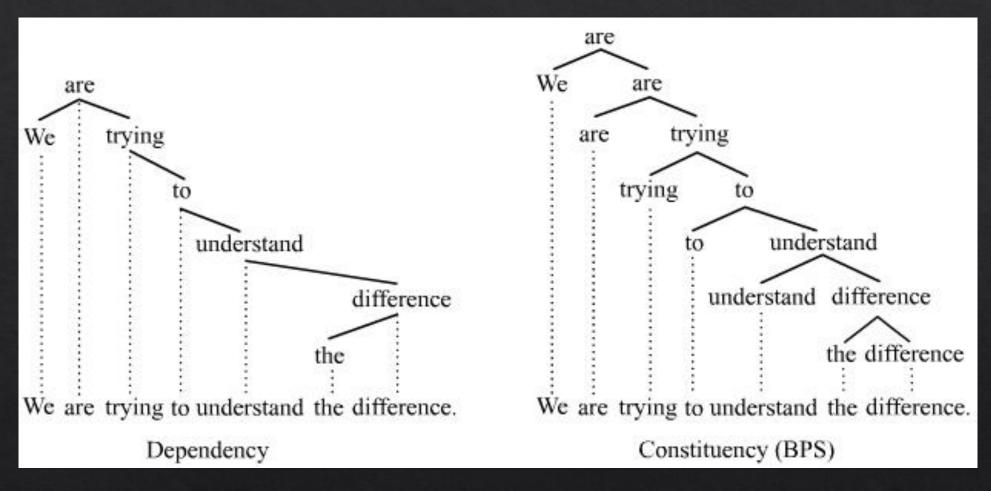
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What is treebank

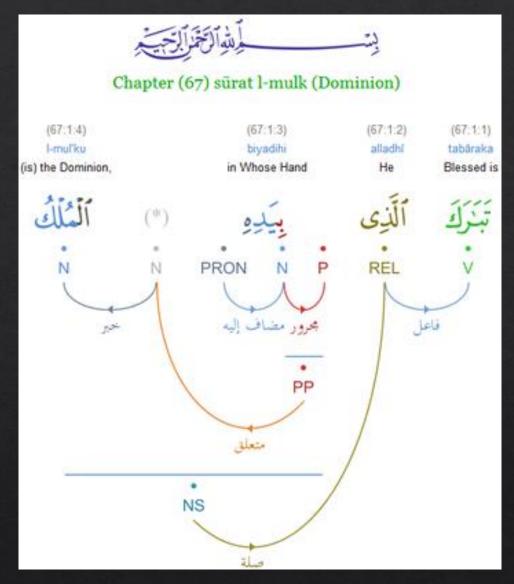


http://en.wikipedia.org/wiki/Phrase_structure_grammar



http://en.wikipedia.org/wiki/Dependency_grammar

Hybrid constituency/dependency tree from the Quranic Arabic Corpus



http://en.wikipedia.org/wiki/Treebank

1 2 3 4 5 6 7	Pricing details were n't immediately available	pricing detail be not immediately available	NN NNS VBD RB RB JJ	 2 3 0 3 6 3 3	NMOD SBJ ROOT ADV AMOD PRD P
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CoNLL-X

```
Other format (xml)
```

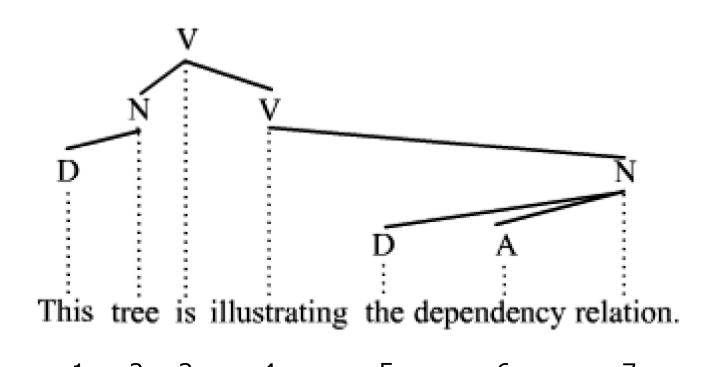
https://code.google.com/p/clearparser/wiki/DataFormat

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- ♦ Question
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- Dependency distance & word order (head initial, head final, mixed)
- Distributions of frequencies
- Clustering algorithm

Dependency Length



DL(this-is)=3-1=2

DL(is-illustrating)=3-4=-1

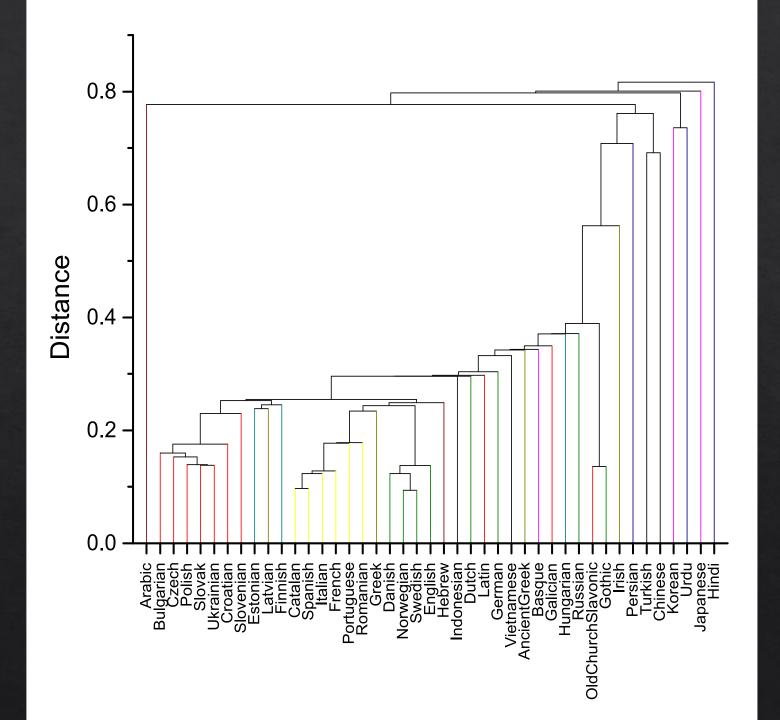
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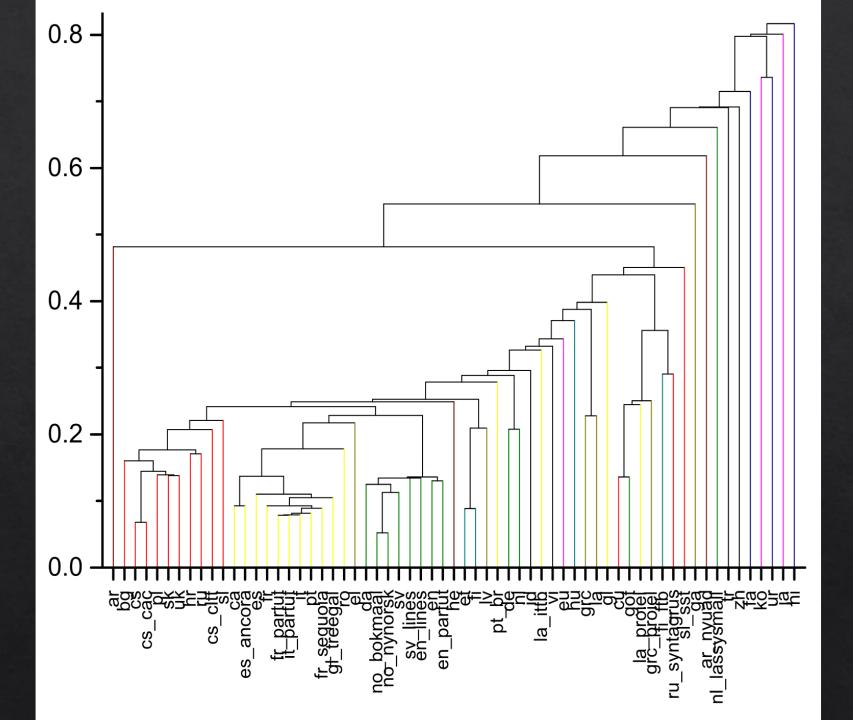
$$DDD(R) = \frac{\sum_{r \in R} distance(r)}{frequency(R)}$$

R: type of dependencies r: dependencies

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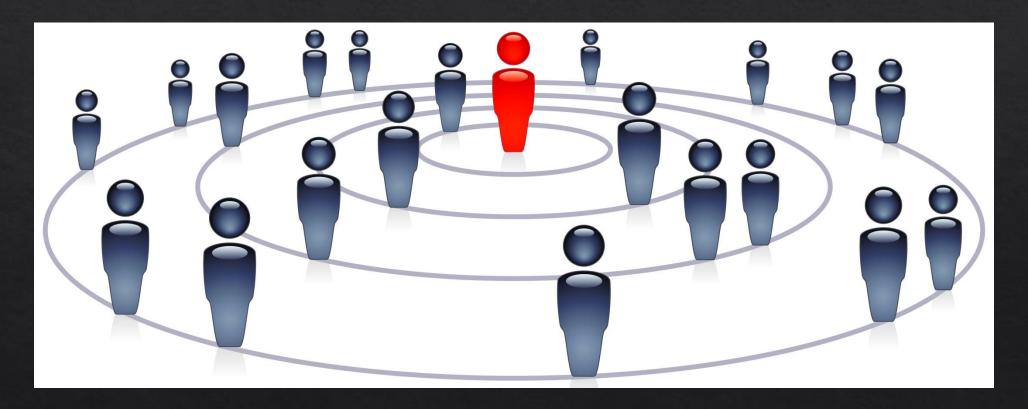
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- What is network

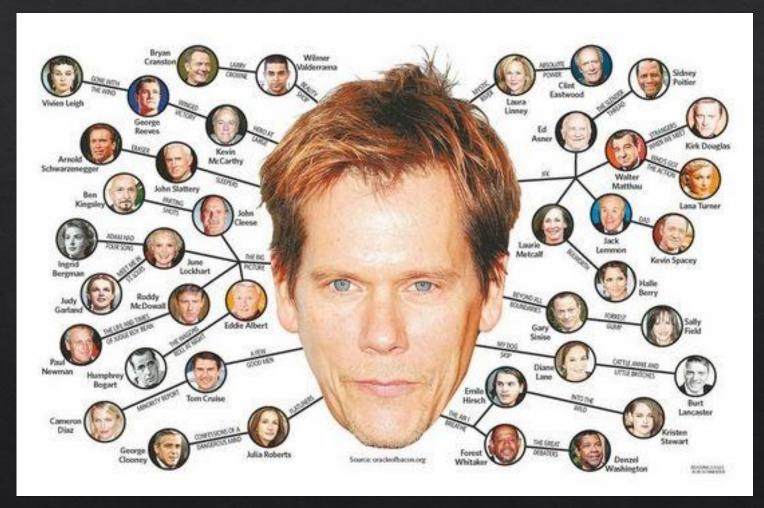
Six Degrees of Separation



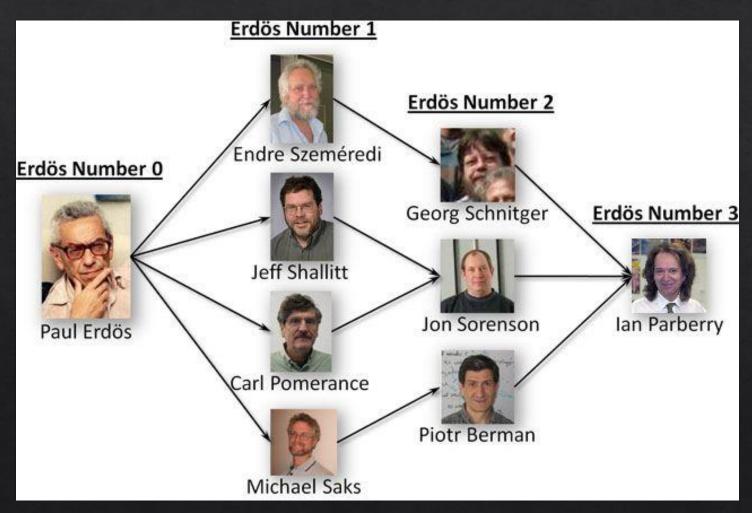
http://indianvox.com/articles/news/143667.php

- Six Degrees of Separation
 - Harvard psychologist Stanley Milgram 1967
 - ♦ U.S. / 300 letters / >60 / 6

Bacon number

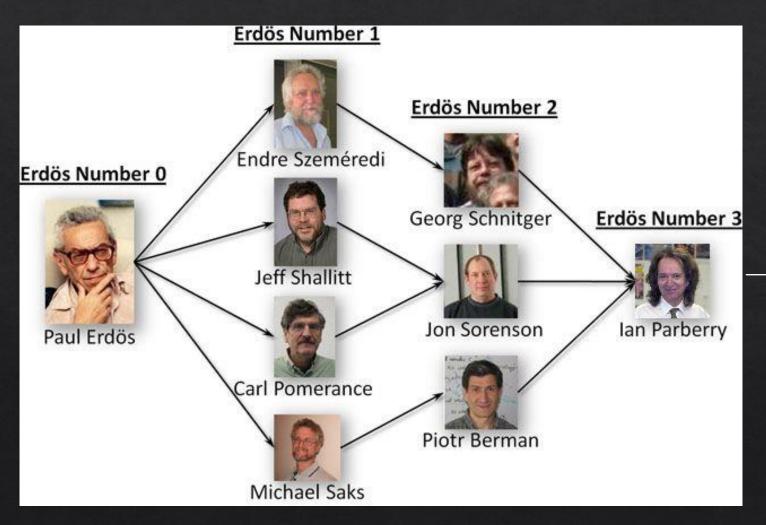


Erdős number



https://larc.unt.edu/ian/claimtofame.html

Erdős number



Erdős number 4

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 - Microsoft researchers Jure Leskovec and Eric Horvitz 2007
 - ♦ 30 billion Skype messages / 240 million people / 6

- Six Degrees of Separation
- Big Data Area & Quantitative Method
 - ♦ UCINET (Borgatti et al. 2002)
 - ♦ PAJEK (Nooy et al. 2005)
 - ♦ NETDRAW (Borgatti 2002)
 - ♦ CYTOSCAPE (Shannon 2003)
 - **◇** ...

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 - Language network

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- Different language networks
 - Syntactic dependency networks, language development or language evolution, language clustering and linguistic categorization, manual and machine translation, word sense disambiguation, communication and interaction, semantic networks, phonetics, morphology, parts of speech, knowledge networks, cognitive networks ...

Why do we need the network approach in linguistic research?

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 - General advantages of the network approach
 - Graphic view, intuitive and easy to understand
 - Overall perspective, not only focusing on the sentence level
 - General method, easier to communicate and to combine with other disciplines, such as neurology
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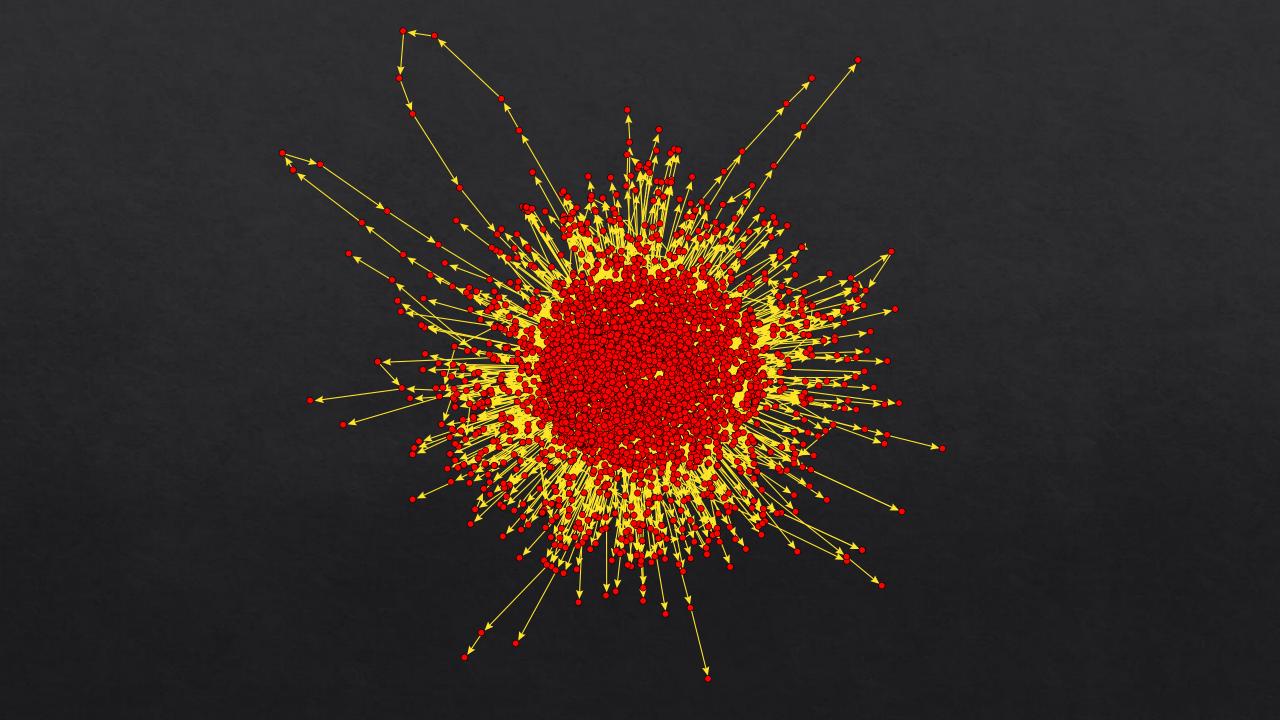
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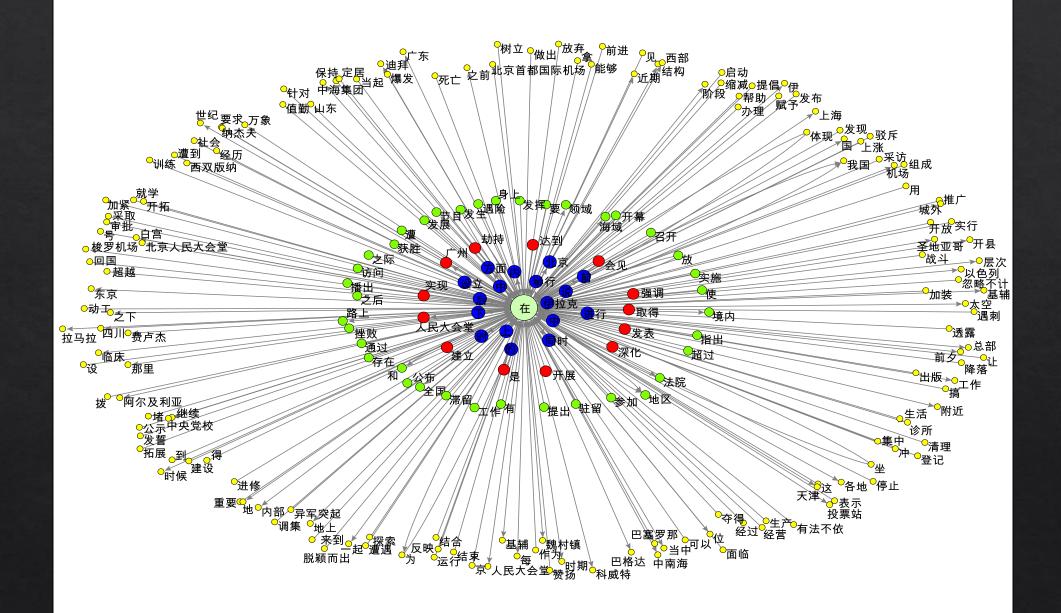
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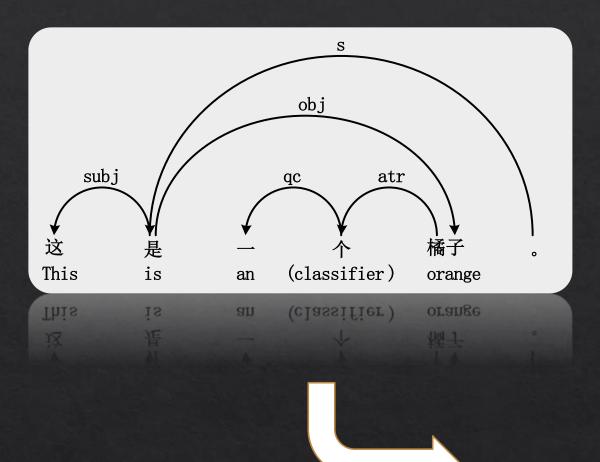
whole ≠ sum of parts

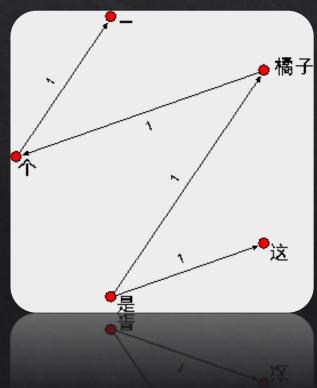
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How to Build the Networks



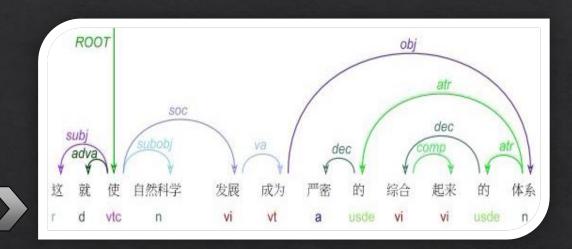


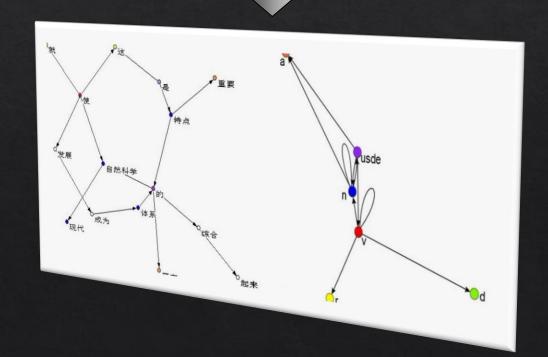
Annotation of a sample sentence in the Treebank. 这是一个橘子 zhe-shi-yi-ge-ju-zi `this is an orange'

Sente		Depende	nt		Govern	or	
nce Order	Order	Charact er	POS	Order	Charact er	POS	Dependency Type
S1	1	这 zhe	pronoun	2	是 shi	verb	Subject
S1	2	是 shi	verb	6	•	punctuati on	main governor
S1	3	— yi	numeral	4	个 ge	classifier	complement of classifier
S1	4	个 ge	classifier	5	橘子 juzi	noun	Attributer
S1	5	橘子 juzi	noun	2	是 shi	verb	Object
S1	6	o	punctuati on				

sno	Wno	wd					
S1	1		Wtag	gno	gwd	aton	
S1	2	 就	r	3	使	gtag Vtc	syt
S1	3		d	3	使	vtc	subj
S1	4		vtc	0	S	S	adva
S1	5	自然科学	n	3	使	vtc	s subobj
S1	6	发展	vi	3	使	vtc	SOC
S1	7	成为	٧t	5	发展	vi	va
S1	8	严密	à	8	的	usde	dec
S1	9	的	usde	12	体系	n	atr
S1	10	综合	vi	11	的	usde	dec
S1		起来	vi	9	综合	vi	comp
S1	11	的	usde	12	体系	n	atr
	12	体系	n	6	成为	vt	obj
S2	1	这	r	2	是	vl	subj
S2	2	是	vl	0	S	S	S
S2	3	现代	n	4	自然科学	n	atr
S2	4	自然科学	n	5	的	usde	dec
S2	5	的	usde	7	特点	n	atr
S2	6	重要	à	7	特点	n	atr
S2	7	特点	n	2	是	vl	obj

52 7 特点 n 2 卓		-
00 1000		
TV 0 1 30	IJ	ati
27 6 重要 。 。 4 年 -		0.0





Chinese Function Characters/Words Evolution

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STEPS:

Get networks

- 4 periods
- Ancient Chinese
- Middle Ancient Times Chinese
- Modern Times Chinese
- Modern Chinese
- 2 single-character words
- •在 zai 'to exist, be living, to stay or remain, (to be located) in, at, '
- •人 ren 'people'

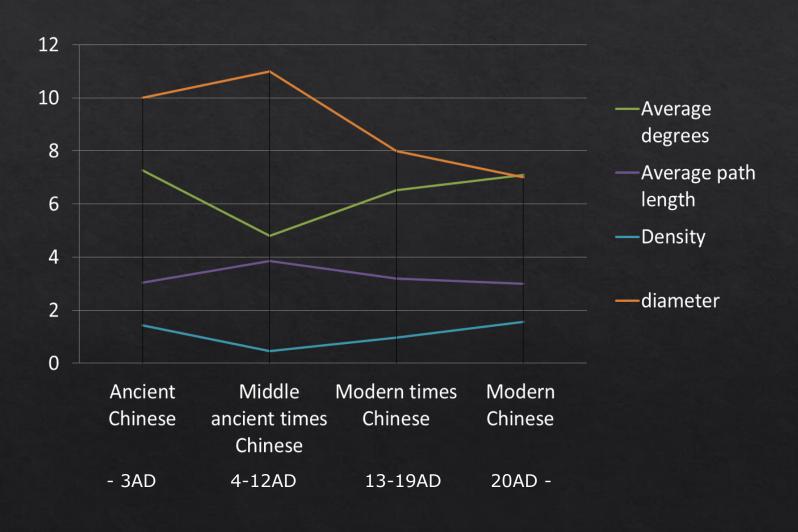
Network characteristics

- average degrees
- average path length
- density
- diameter
- degrees of 'zai'
- degrees of 'ren'

Evolution description

- quantitatively describe the evolution process of Chinese in 4 periods
- quantitatively describe the evolution process of 2 single-character words of Chinese in 4 periods

Co-occurrence Character Networks



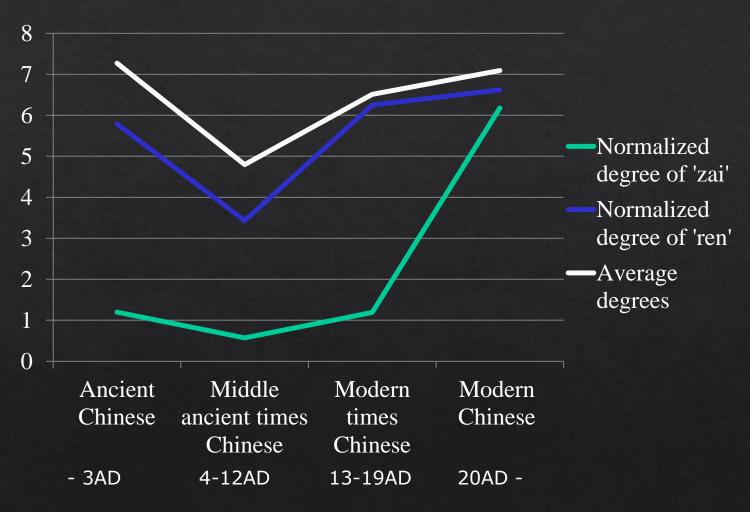


Chinese is an isolating language: syntactic structure relies primarily on function words and word order rather than on rich morphological information to encode functional relations between elements (Levy & Manning 2003).

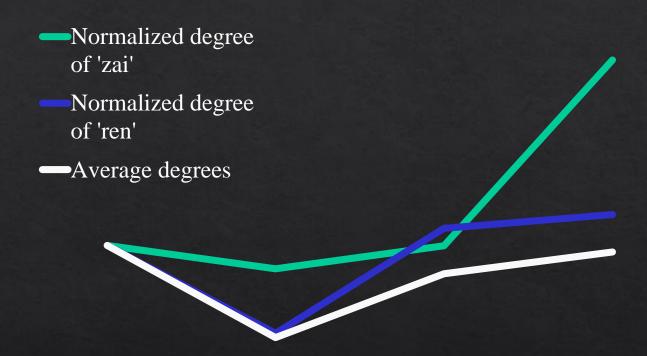
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- ♦ 2 single-character words
 - Frequent characters/words both in the corpus and in general
 - zai underwent a grammaticalization process, whereas ren remains a content word

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- 2 single-character words
 - Frequent characters/words both in the corpus and in general
 - zai underwent a grammaticalization process, whereas ren remains a content word
- Hubs could indicate the grammaticalization process and its starting points. Hubs could be functional or potential functional words to undergo future grammaticalization. (Solé et al. 2002)

Grammaticalization Process in Networks



Grammaticalization Process in Networks



Conclusion

 The network features offer a new source of information to distinguish evolutions of different characters

Observing evolution process in networks allows us to consider the evolutions of some specific units and the evolution of the whole system simultaneously.

Questions & Comments

Thank you!