

### Text Analysis for Big Data

Magnus Sahlgren

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Size

Style (editorial vs social)



# Text analysis Data

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Style (editorial vs social)



### Text analysis Data Size small data \_\_\_\_\_\_big data text analysis

Style (editorial vs social)



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(editorial vs social)



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#### Language



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#### Language







#### Text analysis Technologies

Knowledge-based (use resources like Wikipedia

Supervised machine learning (use annotated data)

Unsupervised machine learning (use unstructured data)



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Supervised machine learning (use annotated data)

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#### Semantic memories



#### Semantic memories

(systems that learn language by reading large amounts of text)



#### Semantic memories

(systems that learn language by reading large amounts of text)





# 



Identify and extract items (e.g. entities and events)

Find relations (e.g. synonyms and associations)



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### Find relations



### Find relations





#### Find relations (lexicon.gavagai.se)

V.

econom

00000

economy's

economía

economia economie

economic

ectomy

ecology

ekonomi

#### GAVAGAI LIVING LEXICON

#### L. certhqueke in English 2. bethery in English 3. charite hebda in Regilian

economy to the 858 most frequent term in our English Lexicon, which puts in

at the tep 8.035% of the vecebulary that contains some 2.599.000 different

Try our Orone Extension to be able to your own applications? Sign up for a

in economy	These terms are frequently used before economy	These terms are economy
	eurozone	
	slowing	(india's)
	bitcoins	manoractorin
	the greek	defence budo
	russia's	japan's econo
	the chinese	foreign exch
	greece's	industrial so
		the greek, gree banking secto financial and
		banking syste
		economic out
		financial set
		anid reserve
		economic ori
		banking indu
		ruling elite
		debt crisis
		stock market
		pension syste
		finance mini-
		agricultural
		tourism sect
		agriculture

MANTICALLY SIMILAR ese terms are used with the	same other terms as	RIGHT SIDE NEIGHBOURS These terms are frequently used after economy
ndia's) manufacturing sector gross domestic product defence budget	(contracted)	contracted is it you
foreign exchange reserves industrial sector		
he greek, greece's) banking sector		
banking system economic output financial sector		
economic woes gold reserves economic crisis banking industry		
ruling elite debt crisis stock markets real economy		
pension system finance ministry		
unisia's) agricultural sector tourism sector agriculture sector	(government-held)	
energy mix tourism industry second-largest economic base		
nuclear industry lowing, sluppish)		
domestic economy local economy global economy economic growth		
property market housing market	(personal finance)	

esseary is used as part of these terms china's economy the global ecceenty fuel economy world's second-largest economy largest economy global economy the world economy chinese economy sharing economy indian economy digital economy warld's second-biggest economy demostic eccentry political economy canada's economy slowing chinese economy economy class fapan's economy local economy india's economy warld economy presius economy circular economy market economy rural economy national economy second-largest economy british economy ania's third-largest economy new economy

economy

noneyweek headlines vahoo the global economy **FUTO 2008** ftse 100 recession canada's economy consecutive quarters canada's world's second-largest economy 53.A 49.7 fastest pace stock indexes china's economy stephes harper cenediana world's most populous spurred shrank dethrosed prowth model



hard landing

market forces

larching

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Compress and refine the information (e.g. summarization and topic detection)



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### Compress and refine



#### Compress and refine (monitor.gavagai.se)



Hillery Clin Occurrence Time	ton ® 1032 mentions September 04	Jeb Bush Ecourrence Time	S93 mentions September D4	Rand Paul Occurrence Time	281 mentions September 34	Mike Huckabee Docurrence Time	220 mentione September D4
WASHINGTON (A presidential candie	P) — Aides to late Hilla	LACONIA, N.H. (P Republican presid	teuters) - U.S. lential	(WASHINGTON) - candidate Mand P	- Presidential auf unveil	Former Arkansas Gor (M), a 2016 presiden.	. Mike Huckaber -

Ted Cruz	۲	Elizabeth Wa	erren 👁	Rick Perry	۲	Ben Carson	۲
Occurrence Time	346 mentions September 04	Dccurrence Time	103 mentions September D4	Occurrence Time	101 mentions September 04	Decurrence Time	321 mentions September D4
Republican presic Texas Sen. Ted C	lential candidate	But the vice presid signaled that he w	lent has also ould	Commitments also campaigns of Ohio	came in from the Gov	Retired neurosurge gaining in several p	ion Ben Carson is

Scott Walker Occurrence Time	September 04	Marco Rubio Decurrence Time	857 mentions September D4	Carly Fiorina Occurrence Time	ITS mentions September 04	Bernie Sanders Occurrence Time	387 mention September 0
Gov. Scott Walker of appeared less clear of	Wisconsin on h	CHATTANOOGA, Tr presidential candidat	enn Republican III	He said towans are a political establis	so tired of the	U.S. Sen. Bernie San for president, spea	ders, candidate

Chris Christie	۲	Donald Trump	۲	
Occurrence Time	1221 mentions September 04	Decurrence Time	91917 mentions September D4	
T EPolation Chris	Christie blames	Donald Trump signed a pledge on Thursday stating that h		



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Measure things (e.g. attitudes and opinions)



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Sentiment analysis



Sentiment analysis



Positivity vs negativity wrt the global economy in English online media



Sentiment analysis



Worry wrt the global economy in English online media



#### Sentiment analysis



Negativity towards China in English online media



#### Sentiment analysis



Attitude towards Sweden in Russian online media



Predict



Predict



Rönnqvist & Sarlin (2015): Detect & Describe: deep learning of bank stress in the news

