

위키피디아 기반 의미식별 연구동양

InSciTe 서비스 제작을 통한 경험을 중심으로

2013. 06. 26. 안국과악기술정보연구원 왕명권



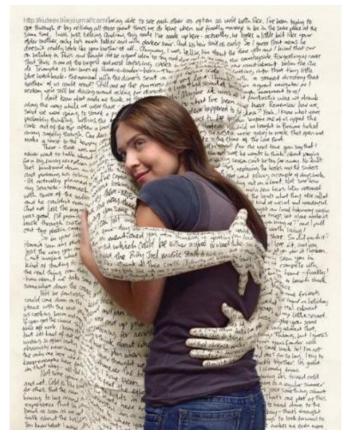
Intro – Word Sense Disambiguation

WSD

문맥의 의미를 파악하여 문맥 내의 단어들의 정확한 의미를 파악하는 것

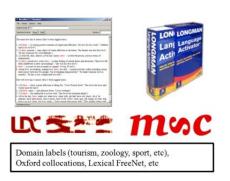
• WSD의 중요성

이미지, 비디오, 텍스트, 그래프,
 소리 등 모든 개체를 강호 분석하기
 위안 필수
 빅 데이터, 데이터 큐레이션 등

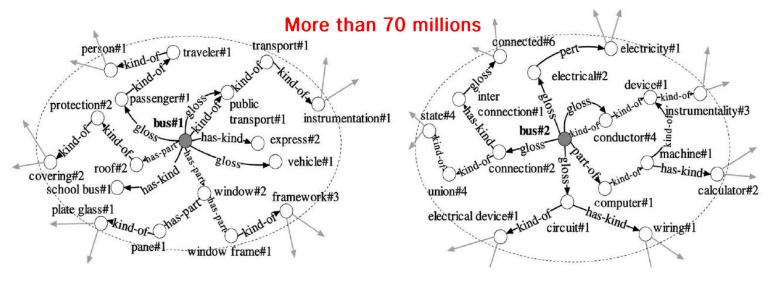




WordNet based WSD – SSI Algorithm



Resources	Explanations
WordNet 2.0	110,00 concepts
Annotated corpora	SemCor, LDC-DSO, WordNet glosses, WordNet usage examples
Dictionaries of collocations	Oxford Collocations, Longman Language Activator, Lexical FreeNet
Etc	Domain labels



R. Navigli, et. al, "Structural Semantic Interconnections: A Knowledge-Based Approach to Word Sense Disambiguation," IEEE PAMI, Vol. 27, No. 7, pp. 1075-1086, 2005.



WordNet based WSD – SSI Algorithm

Retrospective: "an exhibition of a representative selection of an artist's life work."

Retrospective#1, statue#1, artist#1, exhibition#2, object#1, art#1, patinting#1, life#12

Work, selection, representative

```
\begin{split} T &= [retrospective, work, object, exhibition, life, statue, \\ & artist, selection, representative, painting, art] \\ I &= [retrospective\#1, -, -, -, -, -, -, -, -, -, -] \\ P &= [work, object, exhibition, life, statue, artist, selection, \\ & representative, painting, art]. \end{split}
```

```
I = [retrospective\#1, statue\#1, artist\#1]

P = [work, object, exhibition, life, selection, representative, painting, art]
```

$$\begin{split} retrospective \#1 & \xrightarrow{kind-of^2} exhibition \#2, \\ statue \#1 & \xrightarrow{kind-of^3} art \#1 \text{ and } statue \#1 & \xrightarrow{kind-of^6} object \#1. \end{split}$$

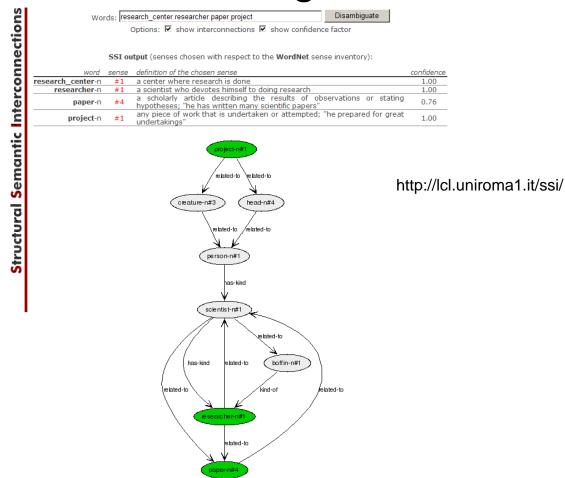
$$I = [retrospective\#1, statue\#1, artist\#1, exhibition\#2, \\ object\#1, art\#1]$$

P = [work, life, selection, representative, painting].

R. Navigli, et. al, "Structural Semantic Interconnections: A Knowledge-Based Approach to Word Sense Disambiguation," IEEE PAMI, Vol. 27, No. 7, pp. 1075-1086, 2005.



WordNet based WSD – SSI Algorithm

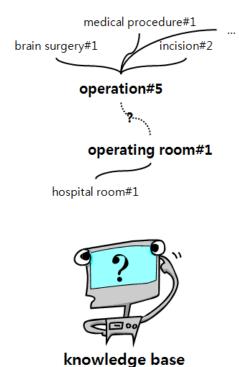


R. Navigli, et. al, "Structural Semantic Interconnections: A Knowledge-Based Approach to Word Sense Disambiguation," IEEE PAMI, Vol. 27, No. 7, pp. 1075-1086, 2005.



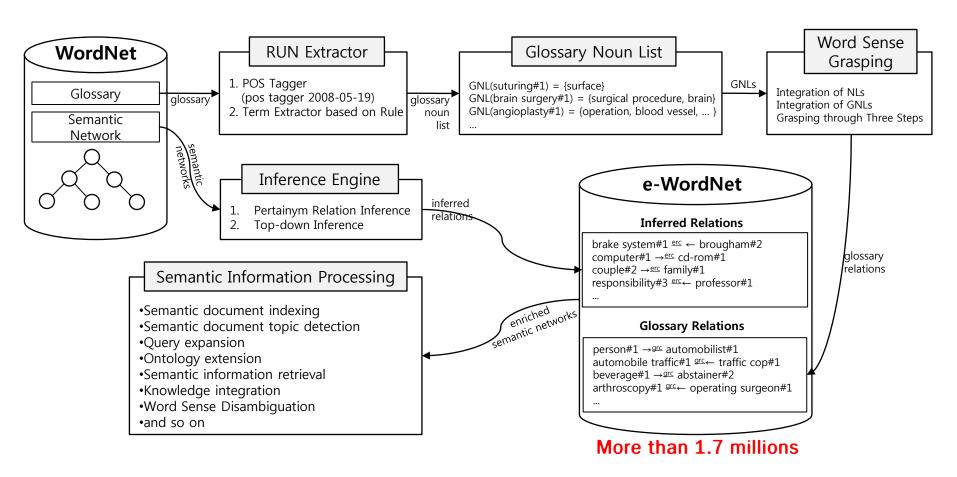
WordNet based WSD – WSD SemNet







WordNet based WSD – WSD SemNet





WordNet based WSD – WSD SemNet

Coverage Test

KB	Basic KB	Light KB	Heavy KB	SSI
Coverage	74.25	82.11	89.13	85.45
(%)	444/598	491/598	533/598	511/598

Concept-pairs	The	The	The	SSI
Concept-pairs	basic	light	heavy	KB
love#1-family#1	X	O	O	O
book#1-desk#1	X	O	O	O
travel#1-tour_guide#1	X	X	O	O
police#1-traffic#1	X	X	X	O
brougham#2-fan_belt#1	X	X	O	X
fire_engine#1-fireman#4	X	X	О	X
fireman#4-fire#1	X	X	O	O
home#1-marriage#1	О	О	O	X
crop#1-growing_season#1	X	О	O	X
emergency#1-fire#1	О	О	О	X
bank#1-money#1	X	О	О	О
buddhist#1-India#1	X	О	О	X
education#1-school#1	X	О	О	О
captain#3-ship#1	X	О	О	О
liquid#1-water#1	X	О	О	О
parent#1-love#1	X	X	X	О
captain#3-ship#1 liquid#1-water#1	X X	0	0	0

WSD based on SemCor

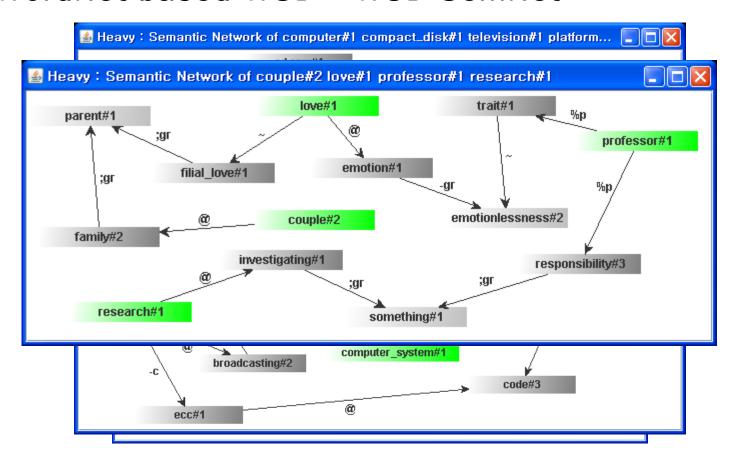
Method	Total Count of Concept Pairs	Pre.	Recall	F1
SSI	70,005,325	56.21	95.08	70.65
WSD-SemNet with the Basic KB	203,760	56.20	82.33	66.24
WSD-SemNet with the Light KB	318,160	59.82	86.84	70.84
WSD-SemNet with the Heavy KB	1,748,627	57.94	92.69	71.31

WSD based on Senseval-3

КВ	Pre.	Recall	F1
The basic	70.3	74.3	72.2
The light	75.7	78.9	77.2
The heavy	71.3	85.2	77.7
The heavy without 'gr'	71.2	83.3	76.8



WordNet based WSD – WSD SemNet





Intro – Wikipedia

• 강점

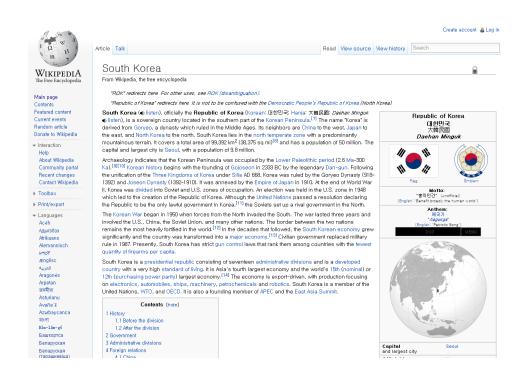
- 전세계 모든 개념 커버, 시맨틱 네트워크 영성
- _ 지속적인 개념 증가
- 신뢰성 보장





Intro – Wikipedia

- Components
 - Title (concept)
 - Abstract
 - Info box
 - Contents
 - In/out link
 - Article—category
 - _ …



Categories: South Korea | Divided regions | East Asian countries | G20 nations | Republics | States and territories established in 1948 | Member states of the United Nations

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Context based WSD

Academic major

From Wikipedia, the free encyclopedia

In the United States and Canada, an **academic major** or **major concentration** (informally, **major** or **concentration**) is the academic discipline to which an undergraduate student formally commits. A student who successfully completes the courses prescribed in an academic major qualifies for an undergraduate degree.

Abbott Lawrence Lowell introduced the *academic major* system to Harvard University in 1910, during his presidency there. It required students to complete courses not only in a specialized discipline, but also in other subjects.^[11] Variations of this system are now definitive among tertiary education institutions in the United States and Canada.

Today, an academic major typically consists of a core curriculum, prescribed courses, a liberal arts curriculum, and several elective courses. The amount of latitude a student has in choosing courses varies from program to program. [1] Typically, the courses of an academic major are portioned in several academic terms.

A major is administered by select faculty in an academic department. A major administered by more than one academic department is called an **interdisciplinary major**. In addition, some students earn individually designed majors. [2]

Whereas some students choose a major when first enrolling as an undergraduate at a school, others choose one after beginning their studies. Some schools forbid students from declaring a major until the end of their second academic year.

A student who declares two academic majors is said to have a **double major**. A **coordinate major** is an ancillary major designed to complement the primary one. A coordinate major requires
fewer course credits to complete. (Compare with academic minor and joint honours.)

Noun (tf)	Student(0.12), university(0.08), major(0.08), core(0.06), field(0.04), study(0.04), education(0.04), college(0.04), term(0.02), portion(0.04), year(0.04), number(0.04), curriculum(0.04), concentration(0.04), art(0.02), program(0.02), framework(0.02),
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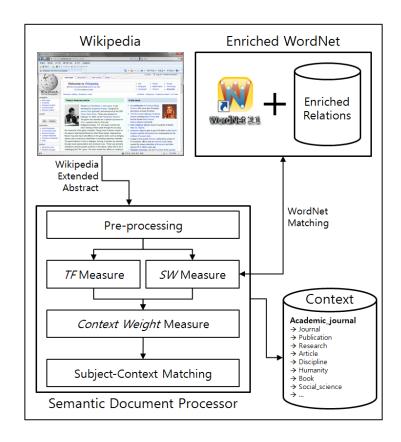


	Context Word#Sense	Context_CW
	student#1	0.654
	university#2	0.344
	major#4	0.312
	core#4	0.271
Context Weight	field#4	0.259
	study#6	0.259
	education#1	0.222
	college#1	0.181
	curriculum#1	0.147

D. Choi, et. al, "Semantic Context Extraction from Wikipedia Document," In Proceedings of The 2010 International Conference on Semantic Web & Web Services, pp. 38-41, 2010.



Context based WSD



D. Choi, et. al, "Semantic Context Extraction from Wikipedia Document," In Proceedings of The 2010 International Conference on Semantic Web & Web Services, pp. 38-41, 2010.



Context based WSD

$$relatedness(s_{ia}, s_{jb}) = \frac{1}{\underset{s_{ia} \in SL_{i}, s_{jb} \in SL_{j}}{\arg\min(dist(s_{ia}, s_{jb}))}}, i \neq j$$

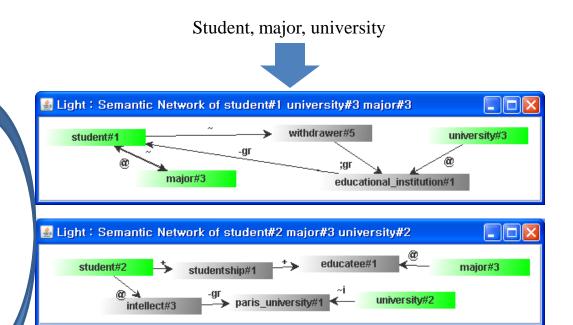
$$sw(s_{ia}) = \sum_{j=1}^{n} \underset{s_{jb} \in S_{j}}{\arg\max(relatedness(s_{ia}, s_{jb}))}, i \neq j$$

$$sw(t_{i}) = \underset{s_{ia} \in S_{i}}{\arg\max(sw_{ia})}$$

$$cw(t_{i}) = tf(t_{i}) \times (sw(t_{i}) + 1)$$

Academic major

Context Word#Sense	Context_CW
student#1	0.654
university#2	0.344
major#4	0.312
core#4	0.271
field#4	0.259
study#6	0.259
education#1	0.222
college#1	0.181
curriculum#1	0.147
•••	•••



relatedness(student#1, major#3) = 0.5 relatedness(student#1, university#3) = 0.33 cw(student#1) = 0.83

relatedness(student#2, major#3) = 0.25 relatedness(student#2, university#2) = 0.25 cw(student#2) = 0.5

D. Choi, et. al, "Semantic Context Extraction from Wikipedia Document," In Proceedings of The 2010 International Conference on Semantic Web & Web Services, pp. 38-41, 2010.



Article/Category based WSD (Similarity)

Similarity (natural language processing, word sense disambiguation)

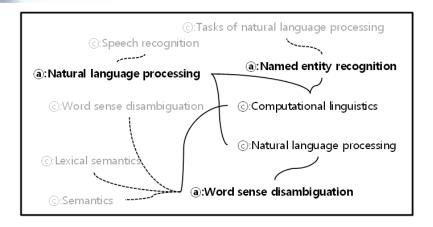
Similarity (natural language processing, named entity recognition)

Pair	Method	Similarity measure
	sim_C	(2*1)/(3+2) = 0.400
NLP, NER	sim_I	(3*3)/(2*(212+152)+(2*43)/(3*(212+152)=0.041
	sim_H	(0.4+0.041)/2 = 0.221
	sim_C	(2*2)/(3+5) = 0.500
NLP, WSD	sim_I	(2*3)/(2*(212+26)+(2*10)/(3*(212+26)) = 0.= 0.087
	sim_H	(0.5+0.087)/2 = 0.294

M. Hwang, et. al, "Measuring Similarities between Technical Terms Based on Wikipedia," In Proceedings of IEEE International Conferences on Things, and Cyber, Physical and Social Computing, pp. 533-536, 2011.



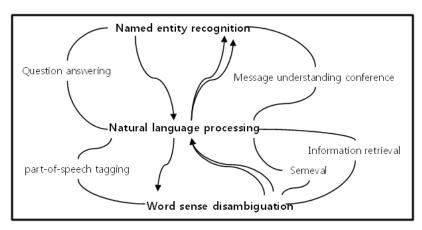
Article/Category based WSD



$$sim_C(a,b) = \frac{2 \times |C_a \cap C_b|}{\left(|C_a| + |C_b|\right)}$$

$$n_{-}sim_{C}(a,b) = \frac{sim_{C}(a,b)}{\max \arg_{art \in cat_{a}, art \neq a} sim_{C}(a,art)}$$

art means an article and cat_a is a category of an article a.



SNA: 'NLP' - 'NER'

 SN_B : 'NLP' – 'WSD' based on Wikipeida interlink Actually 'NLP,' 'NER,' and 'WSD' has 212, 152, and 26 interlinks

(DL, IL) of SN_A and SN_B is (3, 43) and (3, 10) respectively in Wikipedia.

$$sim_{I}(a,b) = \frac{2 \times |DL(a,b)|}{2 \times (|\vec{a}| + |\vec{b}|)} + \frac{2 \times |IL(a,b)|}{3 \times (|\vec{a}| + |\vec{b}|)}$$

 $a \rightarrow$ and $b \rightarrow$ is out-link of the article. The results are also normalized by the maximum value. Hereafter, *sim* means the normalized similarity for each measure.

M. Hwang, et. al, "Measuring Similarities between Technical Terms Based on Wikipedia," In Proceedings of IEEE International Conferences on Things, and Cyber, Physical and Social Computing, pp. 533-536, 2011.



Article/Category based WSD

$$sim_H(a,b) = (1-\alpha) \times sim_C(a,b) + \alpha \times sim_I(a,b)$$

Pair	Method	Similarity measure	
	sim_C	(2*1)/(3+2) = 0.400	
NLP, NER	sim_I	(3*3)/(2*(212+152)+(2*43)/(3*(212+152)=0.041	
	sim_H	(0.4+0.041)/2 = 0.221	
	sim_C	(2*2)/(3+5) = 0.500	
NLP, WSD	sim_I	(2*3)/(2*(212+26)+(2*10)/(3*(212+26)) = 0.= 0.087	
	sim_H	(0.5+0.087)/2 = 0.294	

M. Hwang, et. al, "Measuring Similarities between Technical Terms Based on Wikipedia," In Proceedings of IEEE International Conferences on Things, and Cyber, Physical and Social Computing, pp. 533-536, 2011.

Iranian POW negotiator holds talks with Iraqi ministers

The head of Iran's prisoner of war commission met with two Iraqi Cabinet ministers Saturday in a bid to glean information about thousands of Iranian POWs allegedly in Iraq, the official Iraqi News Agency reported.

Iraqi Foreign Minister Mohammed Saeed al-Sahhaf told Abdullah al-Najafi that the two states needed to "speed up the closure of what remains from the POW and Missing-In-Action file," INA said.

The issue of POWs and missing persons remains a stumbling block to normalizing relations between the two neighbors.

Iraq has long maintained that it has released all Iranian prisoners captured in the 1980-88 Iran-Iraq War. The countries accuse each other of hiding POWs and preventing visits by the International Committee of the Red Cross to prisoner camps.

The ICRC representative in Baghdad, Manuel Bessler, told The Associated Press that his organization has had difficulty visiting POWs on both

sides on a regular basis.

In April, Iran released 5,584

since **1990**.

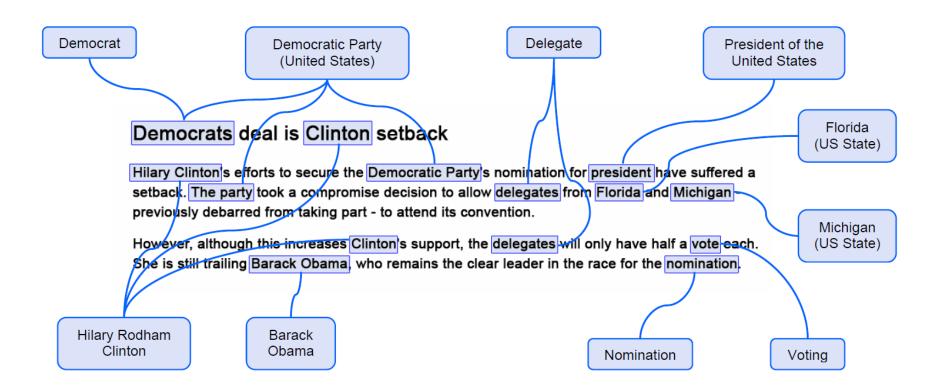
Baghdad

Baghdad is the capital of Iraq and of Baghdad Governorate. With a metropolitan area estimated at a population of More than 1 million people w 7,000,000, it is the largest city in Iraq. It is the second-largest city in the Arab world (after Cairo) and the second-largest city in southwest Asia (after Tehran).

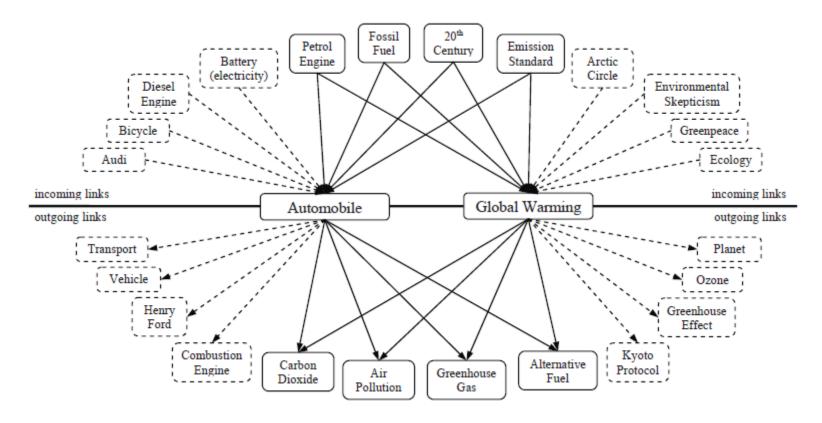
open in wikipedia

fied as civil law detainees in the largest exchange

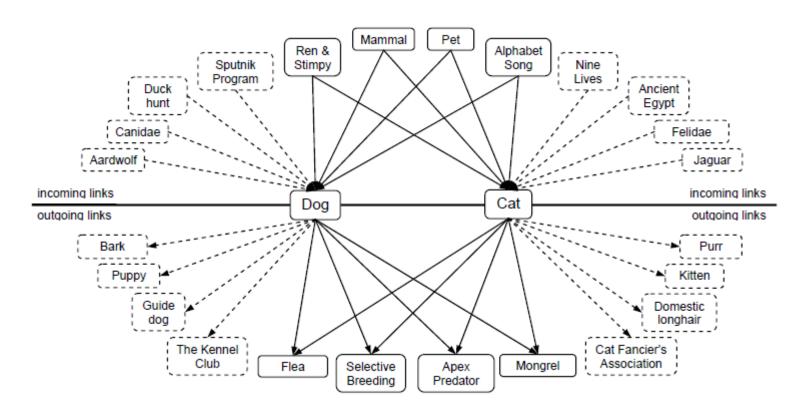
David Milne, "Learning to Link with Wikipedia," In Proceedings of CIKM 2008.



David Milne, "Learning to Link with Wikipedia," In Proceedings of CIKM 2008.



David Milne and Ian H. Witten, "An Effective, Low-Cost Measure of Semantic Relatedness Obtained from Wikipedia Links," In Proceedings fo AAAI 2008



David Milne, "An Open-Source Toolkit for Mining Wikipedia," In Proceedings of New Zealand Computer Science Research Student Conference, 2009.

$$sr(a,b) = \frac{\log(\max(|A|,|B|)) - \log(|A \cap B|)}{\log(|W|) - \log(\min(|A|,|B|))}$$

Depth-first search

From Wikipedia, the free encyclopedia

Depth-first search (**DFS**) is an algorithm for traversing or searching a tree tree structure or graph. One starts at the root (selecting some node as the root in the graph case) and explores as far as possible along each branch before backtracking.

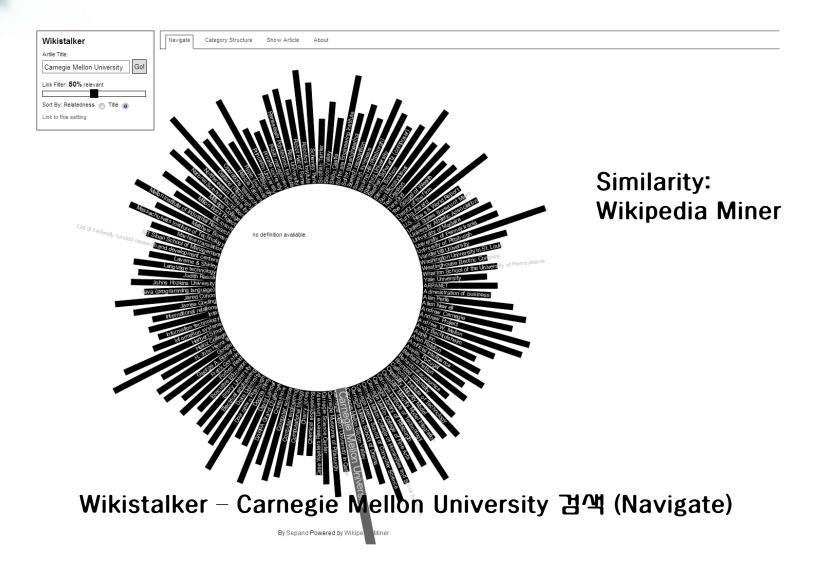
Formally, DFS is an uninformed search that progresses by expanding the first child node of the search tree that appears and thus going deeper and deeper until a goal node is found, or until it hits a node that has no children. Then the search backtracks, returning to the most recent node it hadn't finished exploring. In a non-recursive implementation, all freshly expanded nodes are added to a LIFO stack for exploration.

sense	commonness	relatedness
Tree	92.82%	15.97%
Tree (graph theory)	2.94%	59.91%
Tree (data structure)	2.57%	63.26%
Tree (set theory)	0.15%	34.04%
Phylogenetic tree	0.07%	20.33%
Christmas tree	0.07%	0.0%
Binary tree	0.04%	62.43%
Family tree	0.04%	16.31%
<u></u>		

David Milne, "Learning to Link with Wikipedia," In Proceedings of CIKM 2008.

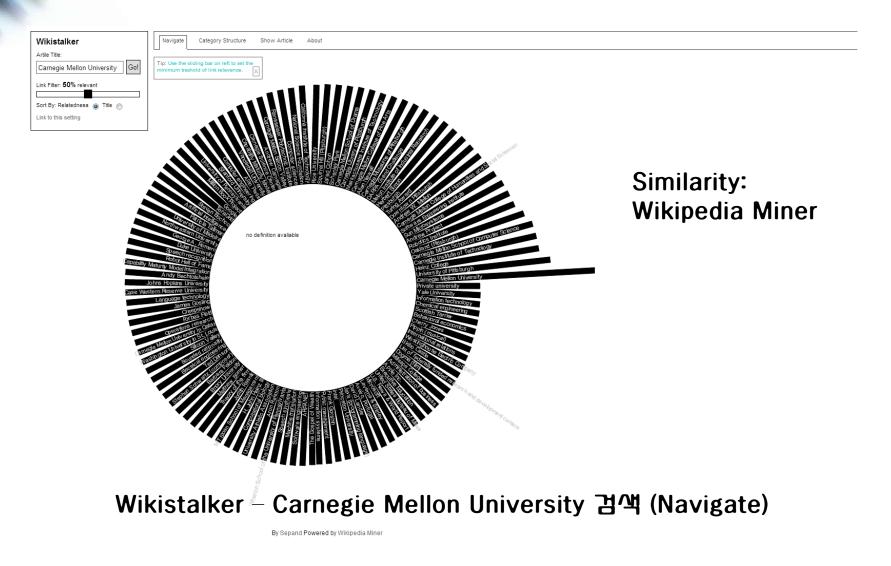


Wikistalker





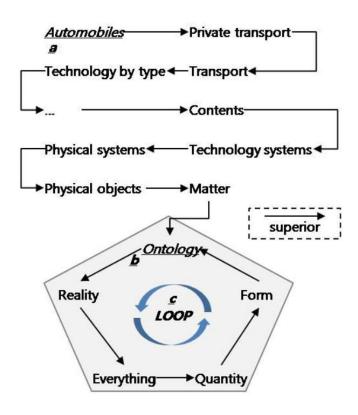
Wikistalker





Outro

Limitations





Thank you for having me.

왕명 권

mgh@kisti.re.kr

http://johnnie.kisti.re.kr