Data and Network Analysis on Wikipedia Outline Computer Science & Artificial Intelligence

Introduction to Digital Humanities Final Project Professor Moses Boudourides Joonha Yu

Goals:

- Visualize network patterns within pages originating from two outlines
- Identify existing communities in the network
- Identify major topics within pages originating from two outlines
- Find relationships between words used in Wikipedia pages

1. Wikipedia Networks of Hyperlinks and Topic Modeling of Pages Originating from the *Outline of computer science* and the *Outline of artificial intelligence*

```
excluded=['International Standard Book Number',
     'International Standard Serial Number',
     'JSTOR'.
     'Library of Congress Control Number',
     "Digital object identifier",
     "Integrated Authority File",
     "PubMed Identifier",
     "PubMed Central",
     "OCLC",
     "Wayback Machine",
     "ArXiv",
     "Bibcode",
     "ACM Computing Classification System",
     "Academic Press",
     "Website",
     "World Wide Web",
     "BioRxiv",
     "CiteSeerX",
     "Telecommunication network",
     "Web sites",
     "Daylight saving time",
     "International Standard Name Identifier",
     "Système universitaire de documentation",
     "Midpeninsula Free University",
     "California",
     "Random House",
     "Associated Press",
     "Computer History Museum",
     "Venture capitalist",
     "List of programmers",
     "List of computer scientists",
     "AFIPS",
     "Electronic publication",
     "Time zone",
     "Scientific American"
```

Collection Process:

- Used python module wikipedia to collect data
- Two outline pages:
 - "Outline of computer science"
 - "Outline of artificial intelligence"
- Excluded 86 links non-relevant to topics
- Verify collected links

Verification Process

- PEAS Performance, Environment, Actuators, Sensors
- Percept (artificial intelligence) –
- Perceptual Computing –
- Rule-based system –
- Self-management (computer science) -
- Soft computing –

[nonexistent page shown in red]

Hardware architect

From Wikipedia, the free encyclopedia (Redirected from Hardware engineer)

[redirect page]

Collection Result:

- "Outline of Computer Science" has 234 functioning hyperlinks
- "Outline of Artificial Intelligence" has 407 functioning hyperlinks
- Two outlines have 26 hyperlinks in common
- In total, there are 615 unique hyperlinks

The directed graph of hyperlinked Wikipedia pages originating from two outlines: 'Outline of computer science and Outline of artificial intelligence'



Graph Diagnostics:

The graph has 477 nodes and 2199 edges

The graph is a simple graph

The graph is an unweighted graph

The graph is a directed graph

The graph is not a bipartite graph

The graph is not a tree

The graph is not strongly connected and it has 429 strongly

connected components

The graph is a weakly connected graph

The graph has no isolates

The density of the graph is 0.010

The transitivity of the graph is 0.287

2. Wikipedia Networks of Reciprocating Hyperlinks of Pages Originating from the *Outline of computer* science and the *Outline of artificial intelligence*

Reciprocation Analysis:

The graph has 2199 hyperlinks. Among those, there are 108 reciprocating hyperlinks.

Graph Diagnostics:

The graph has 47 nodes and 54 edges

The graph is a simple graph

The graph is an unweighted graph

The graph is an undirected graph

The graph is not a bipartite graph

The graph is not a tree

The graph is a disconnected graph and it has 6 connected components

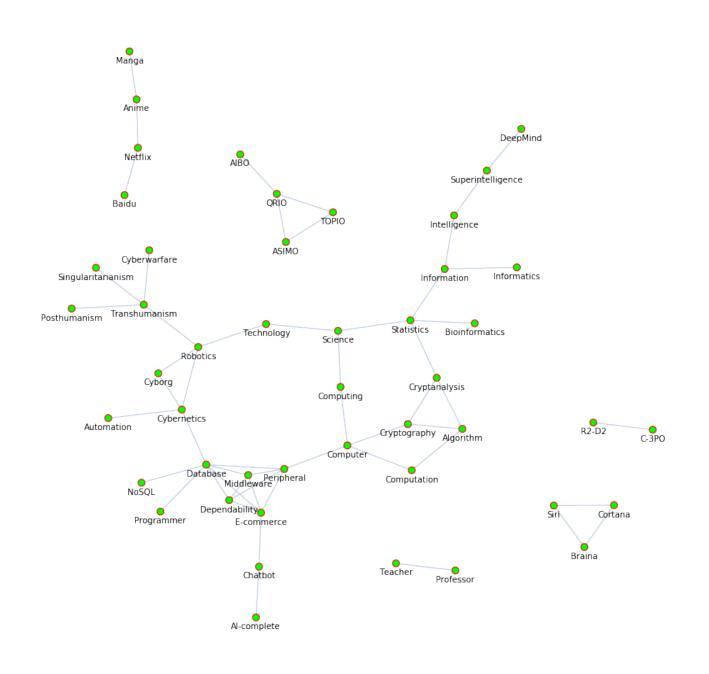
The largest connected component of this graph has 32 nodes and 42 edges

The graph has no isolates

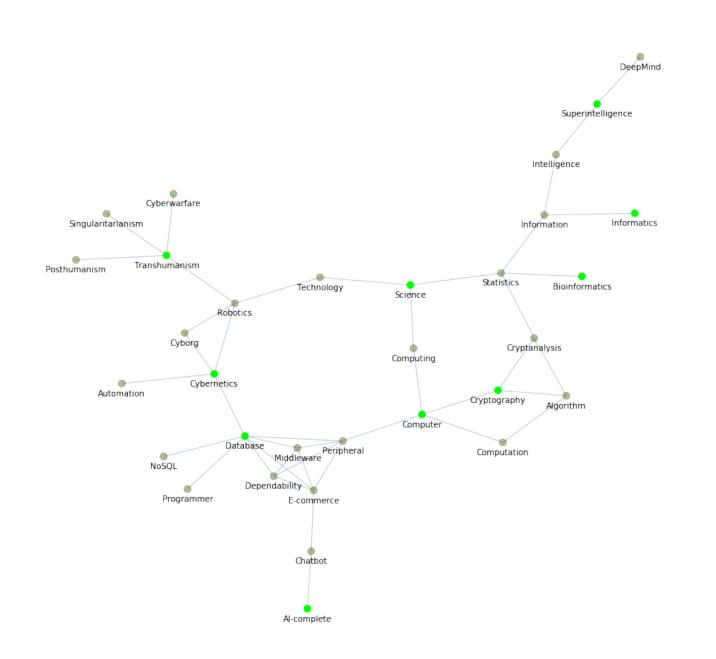
The density of the graph is 0.050

The transitivity of the graph is 0.365

The diameter of the graph is 11

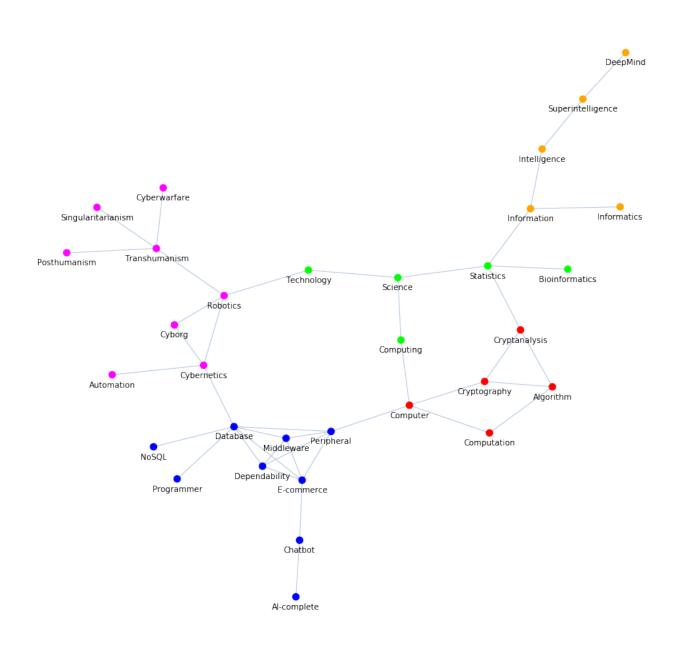


The largest connected component of the network of reciprocated hyperlinks of Wikipedia pages originating from 'Outline of computer science and Outline of artificial intelligence'

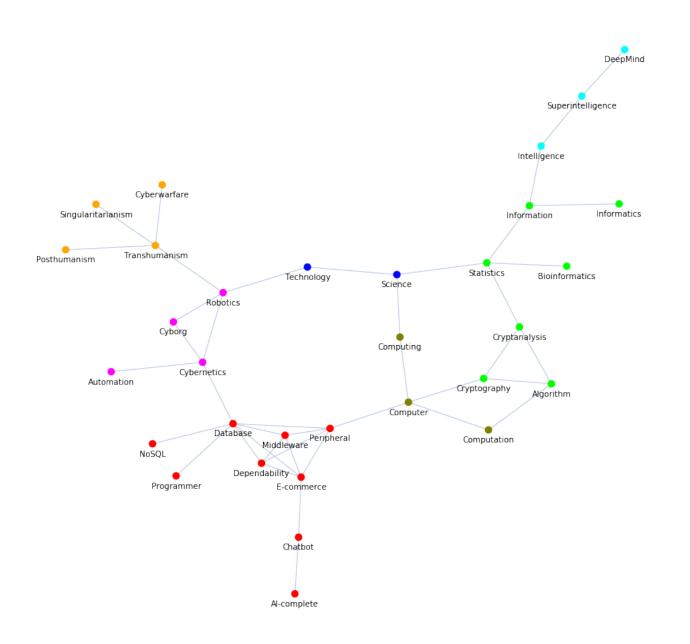


Community Analysis

The largest connected component of the network of reciprocated hyperlinks of Wikipedia pages originating from 'Outline of computer science and Outline of artificial intelligence' colored in 5 Louvain communities



The largest connected component of the network of reciprocated hyperlinks of Wikipedia pages originating from 'Outline of computer science and Outline of artificial intelligence' colored in 7 Fluid communities



3. Topic Modeling of Pages Originating from the *Outline of computer science* and the *Outline of artificial intelligence*

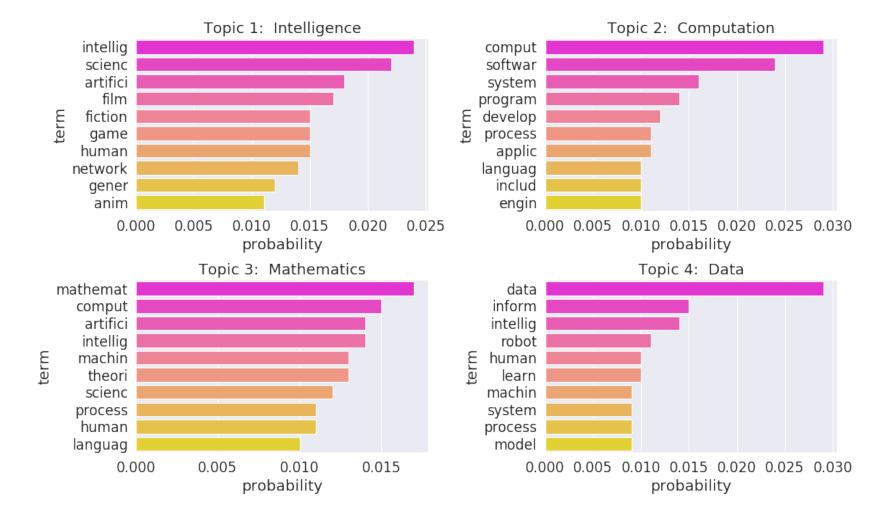
Previous Collection Result:

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Preparing for Topic Modeling:

- Collect Wikipedia summaries from 615 pages.
- Tokenize 11890 words from the summaries.
- Preprocess by lemmatize stemming -> 6237 words.
- To focus on relevant words, filter vocabs which appear at least on 20 pages -> 60 words

Topic Modeling (TM) of the 615 Wikipedia hyperlinks of 'Outline of computer science and Outline of artificial intelligence'



```
Topic 1: Intelligence
   scienc<sub>game</sub>
network
  filmintellig anim human
      fiction
    gener artifici
     Topic 3: Mathematics
  process
       mathemat
human
comput
languagscienc
```

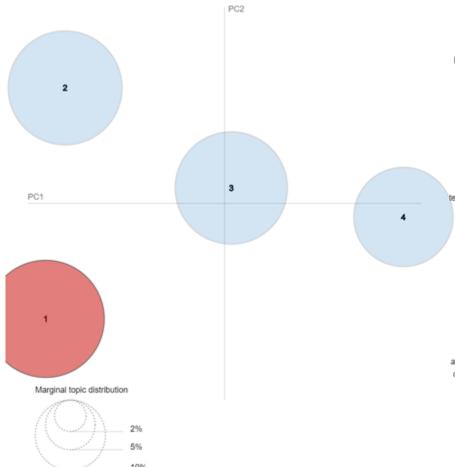
intellig

```
Topic 2: Computation
 process program
includ comput
    develop
          softwar
  applic system
enginlanguag
      Topic 4: Data
       data<sub>inform</sub> intellig
 model
                 robot
  learn
             human
    machin
        system
  process
```

Out[29]: Selected Topic: 0 Previous Topic | Next Topic | Clear Topic Slide to adjust relevance metric:(2) 0.0 0.2 0.4 0.6 0.8 $\lambda = 1$ Intertopic Distance Map (via multidimensional scaling) Top-30 Most Salient Terms¹ 0 100 200 300 500 PC2 data softwar film game fiction anim 2 comput intellig seri mathemat artifici scienc network novel PC1 charact award robot stori star inform franchis googl write type theori hardwar user neural play agent Marginal topic distribution Overall term frequency Estimated term frequency within the selected topic 1. saliency(term w) = frequency(w) * [sum_t p(t | w) * log(p(t | w)/p(t))] for topics t; see Chuang et. al (2012) 2. relevance(term w | topic t) = $\lambda * p(w \mid t) + (1 - \lambda) * p(w \mid t)/p(w)$; see Sievert & Shirley (2014)

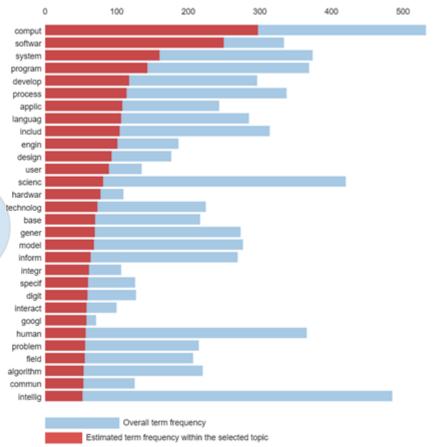


Intertopic Distance Map (via multidimensional scaling)





Top-30 Most Relevant Terms for Topic 1 (27.9% of tokens)



1. saliency(term w) = frequency(w) * [sum_tp(t|w) * log(p(t|w)/p(t))] for topics t; see Chuang et. al (2012) 2. relevance(term w|topic t) = λ * p(w|t) + (1 - λ) * p(w|t)/p(w); see Sievert & Shirley (2014)

Out[29]: Selected Topic: 0 Previous Topic | Next Topic | Clear Topic Slide to adjust relevance metric:(2) 0.0 0.2 0.4 0.6 0.8 $\lambda = 1$ Intertopic Distance Map (via multidimensional scaling) Top-30 Most Relevant Terms for Topic 2 (26.4% of tokens) 0 100 200 300 PC2 data inform intellig robot human learn machin system process model base agent theori oper PC1 type object applic includ problem structur program time scienc exampl class term technolog develop languag function Marginal topic distribution Overall term frequency Estimated term frequency within the selected topic $\underline{\text{1. saliency}(\text{term w}) = \text{frequency}(w) * [\underline{\text{sum}}\underline{\text{t}}\ p(t\mid w) * log(p(t\mid w)/p(t))]}\ \text{for topics } t; \ \text{see Chuang et. al}\ (2012)$ 2. relevance(term w | topic t) = $\lambda * p(w \mid t) + (1 - \lambda) * p(w \mid t)/p(w)$; see Sievert & Shirley (2014)

Out[29]: Selected Topic: 0 Previous Topic | Next Topic | Clear Topic Slide to adjust relevance metric:(2) $\lambda = 1$ Intertopic Distance Map (via multidimensional scaling) Top-30 Most Relevant Terms for Topic 3 (25.6% of tokens) 0 100 200 300 500 PC2 mathemat comput artifici intellig machin theori 2 scienc process human languag control develop function algorithm PC1 system program research includ problem model seri technolog studi award term univers formal method Marginal topic distribution Overall term frequency Estimated term frequency within the selected topic $\underline{\text{1. saliency}(\text{term w}) = \text{frequency}(w) * [\underline{\text{sum}}\underline{\text{t}}\ p(t\mid w) * log(p(t\mid w)/p(t))]}\ \text{for topics } t; \ \text{see Chuang et. al}\ (2012)$ 2. relevance(term w | topic t) = $\lambda * p(w \mid t) + (1 - \lambda) * p(w \mid t)/p(w)$; see Sievert & Shirley (2014)

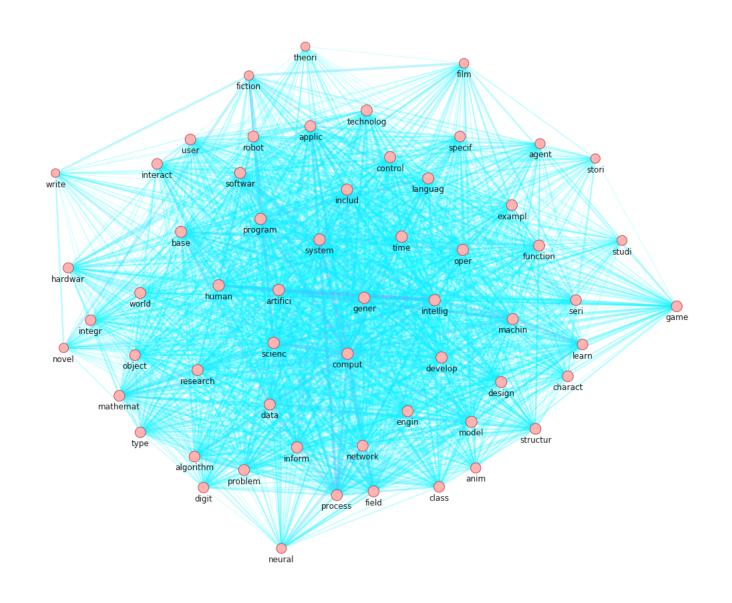
Out[29]: Selected Topic: 0 Previous Topic | Next Topic | Clear Topic Slide to adjust relevance metric:(2) $\lambda = 1$ Intertopic Distance Map (via multidimensional scaling) Top-30 Most Relevant Terms for Topic 4 (20% of tokens) 0 100 200 300 400 500 PC2 intellig scienc film fiction game 2 human network gener seri program write charact PC1 novel includ model machin control stori neural world robot comput book creat time research Marginal topic distribution Overall term frequency Estimated term frequency within the selected topic 1. saliency(term w) = frequency(w) * [sum_t p(t | w) * log(p(t | w)/p(t))] for topics t; see Chuang et. al (2012) 2. relevance(term w | topic t) = $\lambda * p(w \mid t) + (1 - \lambda) * p(w \mid t)/p(w)$; see Sievert & Shirley (2014)

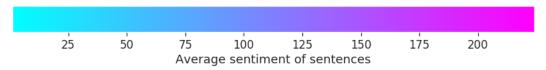
Visualization of Sententially Co-occurrent terms

```
['agent',
'algorithm',
'anim'.
'applic',
'artifici',
'base'.
'charact',
'class'.
'comput',
'control',
'data'.
'design',
'develop',
'problem',
'process'.
'program',
'research'.
'robot',
'scienc',
'seri'.
'softwar',
'specif',
'stori',
'structur',
'studi',
'system',
'technolog',
'theori',
'time',
'type',
'user',
'world'.
'write']
```

Preparing for Visualization:

- Among 615 nodes, in total of 60 words without aliases cooccur in at least 20 pages.
- Sentiment scores (polarity score) divided by their average. -1 to 1





Graph Result:

- The graph is weighted
- The graph is a connected graph
- It has 59 nodes and 1536 edges.
- The term that does not cooccur is 'googl'
- The density of this graph is 0.898

Possible Future Works:

Applying the similar analysis to different Wikipedia outlines Applying the similar analysis on other wiki sources Finding more efficient ways to conduct analysis

Thank You