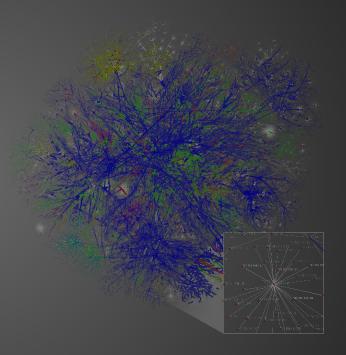
# Human Navigation of Information Systems: Wikispeedia



5/5/17 - APAM 4990 - Jake Hofman Andres, Jayati, and Tomer

#### Our goal

- Wikispeedia Data Set
- Exploratory Data Analysis
- Modeling Task
  - o Topic Modeling
  - Regression



### Motivation

How do humans navigate through information systems?

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Can systems be designed to be easier for humans to navigate?

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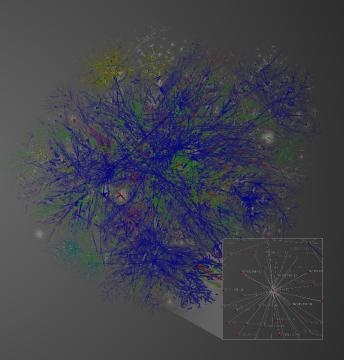
Can systems be designed to be easier for humans to navigate?

This is especially relevant today!

Our goal

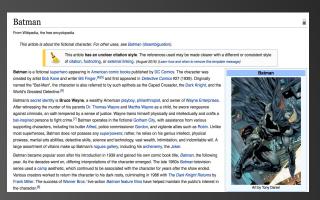
#### Wikispeedia Data Set

- Exploratory Data Analysis
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  - o Topic Modeling
  - Regression



# How do you play?

- How do you get from Article to A and B with the shortest amount of hyperlinks possible?
- Example
  - o Batman => Vitamin D
  - Batman => Superman => Sun => Sunlight =>Vitamin D
- Example
  - The Beatles => Beer
  - The Beatles => United States => Household income in the United States=> Oregon => Wine => Beer





discovery was due to effort to find the dietary substance lacking in rickets (the childhood form of osteomalacia). [4]

Beyond its use to prevent osteomalacia or rickets, the evidence for other health effects of vitamin D supplementation in

#### Raw Data

- Source: <a href="http://snap.stanford.edu/data/wikispeedia.html">http://snap.stanford.edu/data/wikispeedia.html</a>
- User paths Incomplete
- User paths Complete
- Category mapping for articles
- Links each article contains
- Plaintext files of all the articles

Total number of articles: ~4,600 Total number of games played: ~51,000

### Previous Research

WWW 2012 - Session: Web User Behavioral Analysis and Modeling

April 16-20, 2012, Lyon, France

#### **Human Wayfinding in Information Networks**

Robert West Computer Science Department Stanford University west@cs.stanford.edu Jure Leskovec Computer Science Department Stanford University jure@cs.stanford.edu

#### ABSTRACT

Navigating information spaces is an essential part of our everyday lives, and in order to design efficient and user-friendly information systems, it is important to understand how humans navigate and find the information they are looking for. We perform a large-scale study of human wayfinding, in which, given a network of links between the concepts of Wikipedia, people play a game of finding a short path from a given start to a given target concept by following hyperlinks. What distinguishes our setup from other studies of human Web-browsing behavior is that in our case people navigate a graph of connections between concepts, and that the exact goal of the navigation is known ahead of time. We study more than 30,000 goal-directed human search paths and identify strategies people use when navigating information spaces. We find that human wayfinding, while mostly very efficient, differs from shortest paths in characteristic ways. Most subjects navigate through high-degree hubs

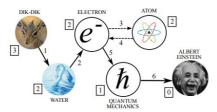
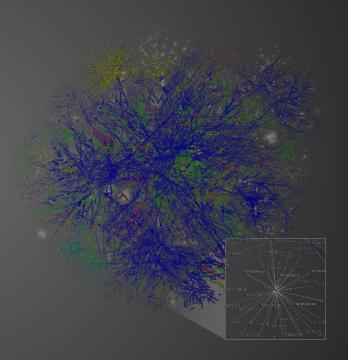


Figure 1: A human example path between the concepts DIK-DIK and ALBERT EINSTEIN. Nodes represent Wikipedia articles and edges the hyperlinks clicked by the human. Edge labels indicate the order of clicks, the framed numbers the shortestpath length to the target. One of several optimal solutions would be (DIK-DIK, WATER, GERMANY, ALBERT EINSTEIN).

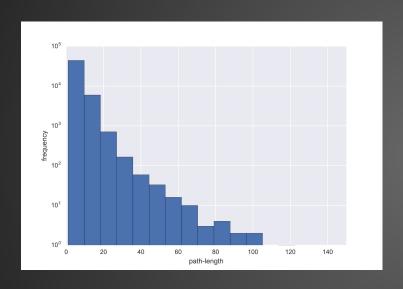
- Our goal
- Wikispeedia Data Set

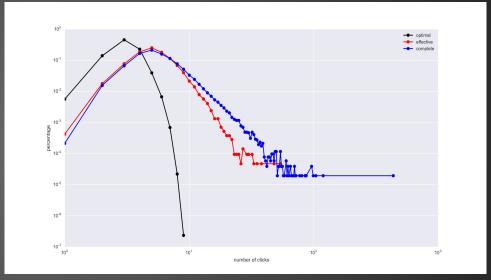
#### **Exploratory Data Analysis**

- Modeling Task
  - Clustering
  - Regression



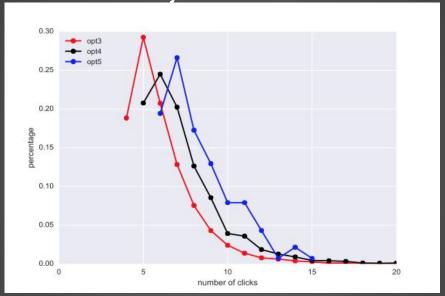
# Exploratory Data Analysis: Basic Statistics





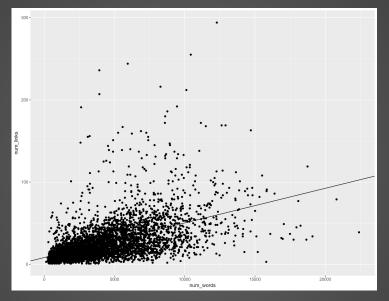
For all games, users tend to do worse than the optimal.

# Exploratory Data Analysis: User Performance



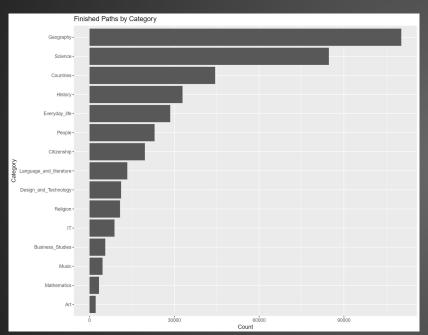
Distribution of optimal paths shows this is a small world network

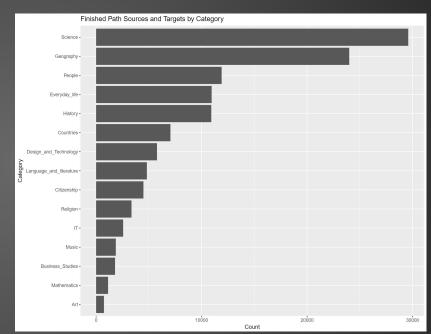
# Exploratory Data Analysis: Regression



No significant correlation exists between article size and number of hyperlinks it contains ( $R^2 = 0.27$ )

# Exploratory Data Analysis: Categories



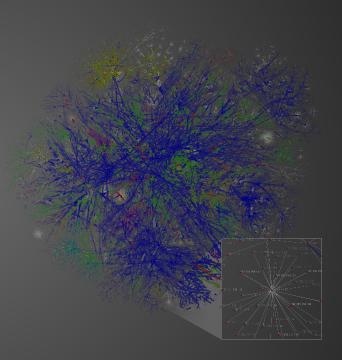


Path category distribution maps well to source/target category distribution

- Our goal
- Wikispeedia Data Set
- Exploratory Data Analysis

#### Modeling Task

- Topic Modeling
- Regression



# Clarifying Our Goal

- Humans likely play Wikispeedia by using conceptual connections. However, it may be faster to go from source to target through a non intuitive path.
- Latent semantic analysis defines a semantic distance between source and target articles to measure the conceptual connection between them.
- Shortest path length defines a topological distance between source and target articles to measure the optimal path between them.

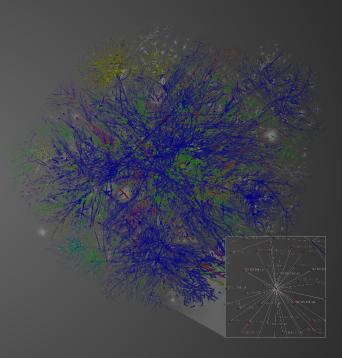
Can semantic distance be a more effective way to organize Wikipedia categories?

Can we better predict human Wikispeedia performance by using semantic distance rather than topological distance between source and target?

- Our goal
- Wikispeedia Data Set
- Exploratory Data Analysis
- Modeling Task

#### **Topic Modeling**

o Regression



Can semantic distance be a more effective way to organize Wikipedia categories?

# Topic Modeling using Latent Dirichlet Allocation

#### Motivation

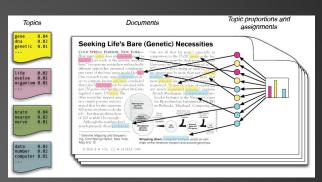
- Is there room for improvement in the categories defined by Wikipedia?
- Higher level, can these categories reveal information about the semantics/clustering of different categories?

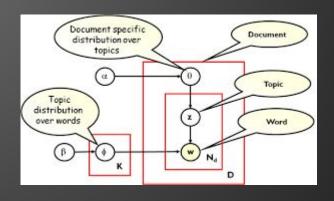
#### How it works

- Topic is a distribution over words
- Document is a mixture of corpus-wide topics
- Word is drawn from one of those topics

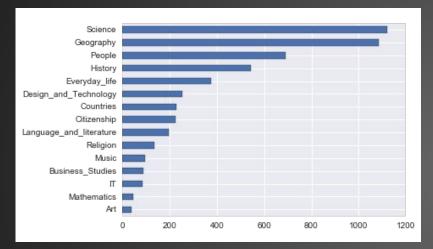
#### What we did

LDA across all 4,000 articles in dataset with 15 topics

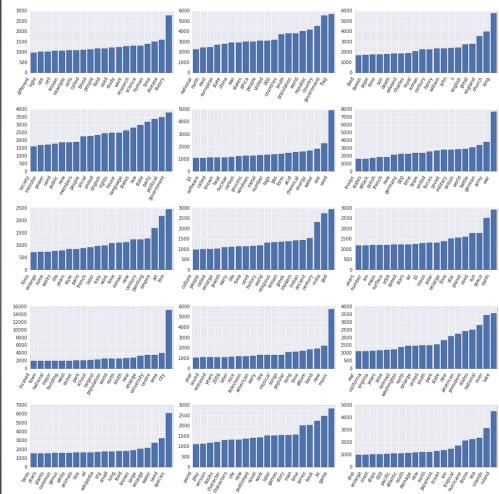




# Topic Modeling: Results

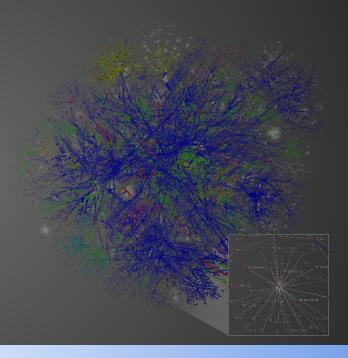


LDA map to categories and provides semantic meaning



- Our goal
- Wikispeedia Data Set
- Exploratory Data Analysis
- Modeling Task
  - o Topic Modeling

Regression



Can we better predict human Wikispeedia performance by using semantic distance rather than topological distance between source and target?

## Latent Semantic Analysis to Measure Document Similarity

#### Motivation

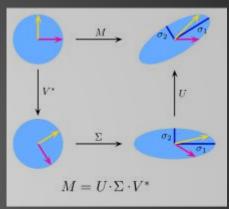
 Proxy for human intuition of navigating network by going through documents that are related

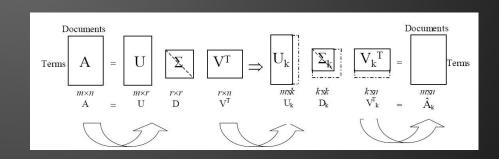
#### How it works

- Truncated Singular Value Decomposition
- o Dimensionality Reduction Technique
- Applied to Bag of Words generated from CountVectorizer

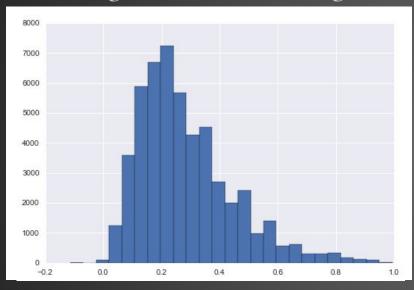
#### • What we did

 Computed semantic distance between every article using LSA



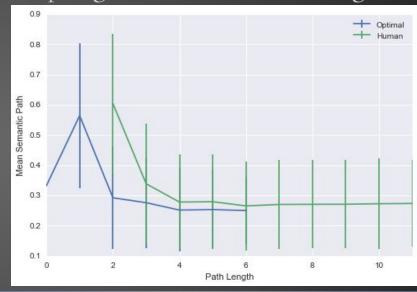


Latent Semantic Analysis: Results
Histogram of Semantic Lengths
Topolog



Slightly skewed normal distribution

Topological versus Semantic Lengths



Higher optimal path lengths have weaker semantic connection

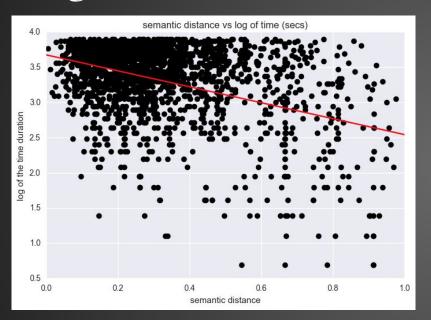
# Latent Semantic Analysis: Review

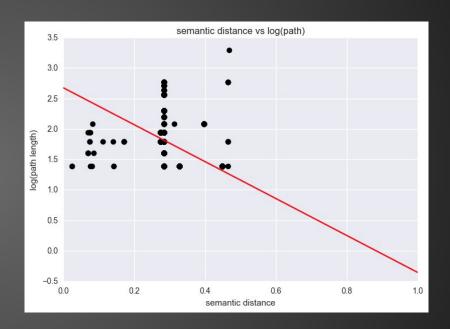
- Examples
  - Same article to same article is 1
  - Glasgow to London is .589
  - o 14th Century to Ancient Greece is .447
  - Weaknesses
- LSA creates a symmetric matrix, but topics may not be symmetric
  - Example: Minneapolis and Minnesota





# Regression Tasks: Results

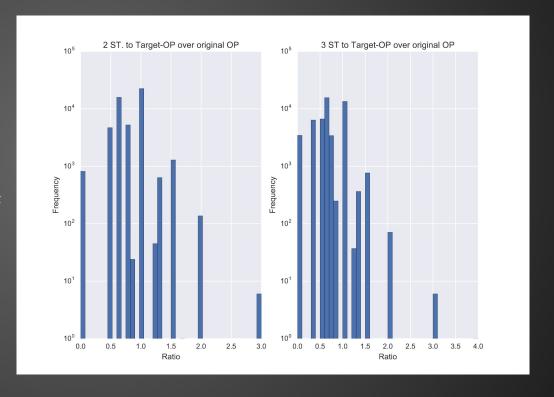




Semantic distance does not correlate to either game duration or path length (R^2 of .18 and .13 respectively)

# Summary

- Topic Modeling
  - LDA provides insights into source/target categories which helps us understand relationship between different clusters
- Regression
  - No correlation between semantic path and game duration
- Next steps
  - Do players get better or worse as they navigate the path?



### References

- http://snap.stanford.edu/data/wikispeedia.html
- <a href="http://infolab.stanford.edu/~west1/pubs/West-Leskovec\_WWW-12.pdf">http://infolab.stanford.edu/~west1/pubs/West-Leskovec\_WWW-12.pdf</a>
- <a href="http://infolab.stanford.edu/~west1/pubs/West-Pineau-Precup\_IJCAI-09.pdf">http://infolab.stanford.edu/~west1/pubs/West-Pineau-Precup\_IJCAI-09.pdf</a>
- https://cseweb.ucsd.edu/~jmcauley/cse255/reports/wi15/Shelby\_Thomas\_Moein\_ Khazraee.pdf

# Thank you! Questions?

