

Using Text Analytics to Spot Fake News

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Text Analytics and Fake News

- “Civilization is in a race between education and catastrophe. Let us learn the truth and spread it as far and wide as our circumstances allow. For truth is the greatest weapon we have.”
 - H. G. Wells
- OED word of the year for 2016: Post-Truth

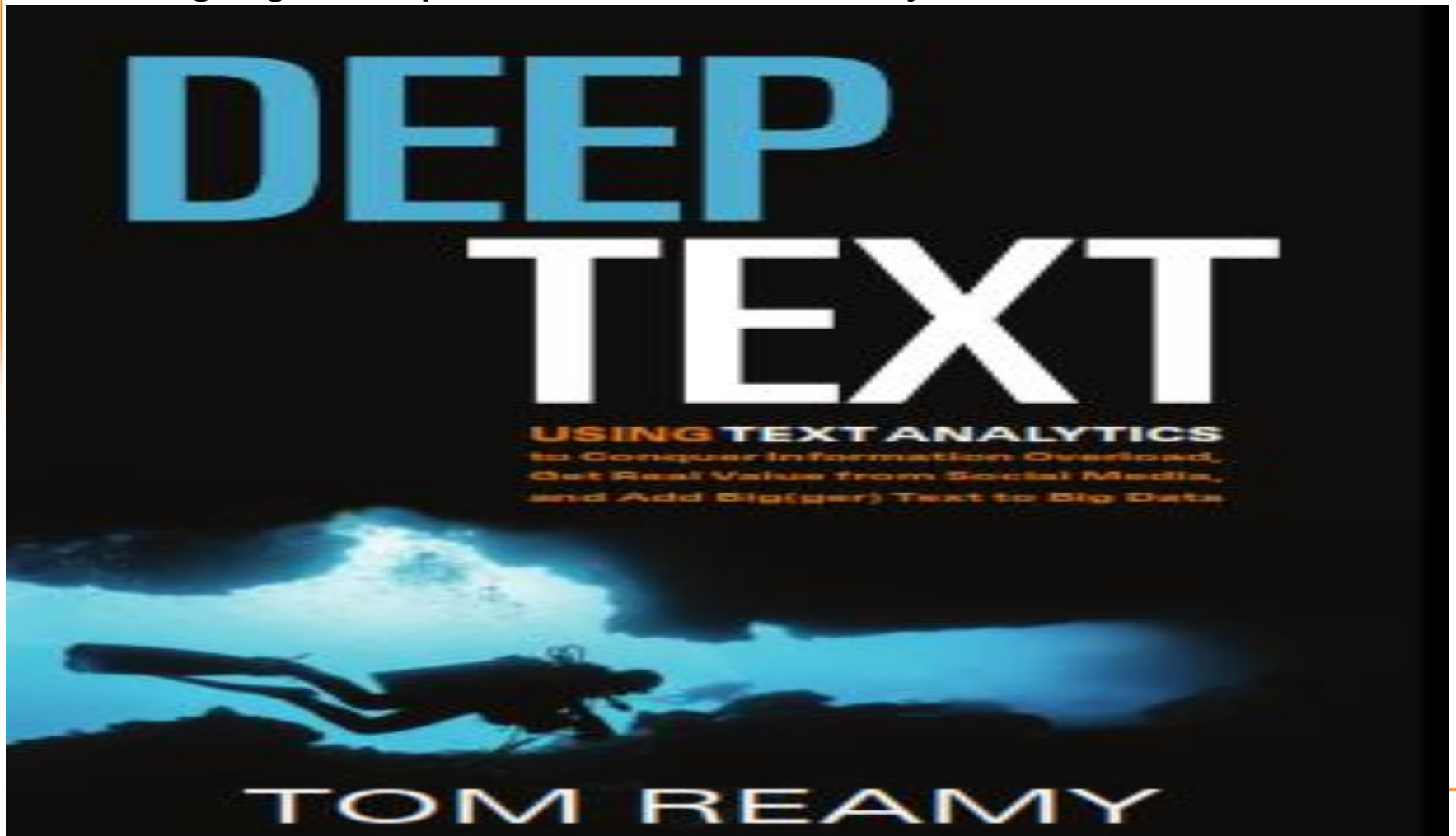
Agenda

- Introduction
- Types of Fake News
- Techniques of Fake News
- Fake News in Context
- Proposed Solutions
- Solutions that Work
- Conclusion

Introduction: KAPS Group

- Network of Consultants and Partners – “Hiring”
- Strategy consulting – Text Analytics, Social Media, Integration
- Text Analytics Smart Start, Next Level
- Development - Taxonomy/Text Analytics, Social Media
- TA Train (1 day to 1 month)
 - Strategic personalized overview to hands on training
- TA Audit –Content, semantic resources, tech, info needs & behaviors
- Partners –Synaptica, SAS, IBM, Smart Logic, Expert Systems, Clarabridge, Lexalytics, BA Insight, BiText
- Clients: Genentech, Novartis, Northwestern Mutual Life, Financial Times, Hyatt, Home Depot, Harvard, British Parliament, Battelle, Amdocs, FDA, GAO, World Bank, Dept. of Transportation, etc.
- Presentations, Articles, White Papers – www.kapsgroup.com

A treasure trove of technical detail, likely to become a definitive source on text analytics – *Kirkus Reviews Information Today Table Book Signing at Reception – 17:15-18:30 Info Today table*



Text Analytics and Fake News

What is Text Analytics?

- Text Mining – NLP, statistical, predictive, machine learning
- Entity / Fact Extraction
- Sentiment Analysis
- Auto-categorization
 - Training sets, Terms, Rules
 - Boolean– Full search syntax – AND, OR, NOT
 - Advanced – DIST(#), ORDDIST#, PARAGRAPH, SENTENCE
- Deep Learning – neural nets – big and fast enough for patterns
 - Good on images, not concepts
- Deep Learning is a dead end – black box, tricks, fast data
 - No common sense and no strategy to get there

Text Analytics and Fake News

What is Fake News?

- Types of Fake News – sliding scale
 - Information out of context, Opinion, Misinformation
 - Alternative facts, Lies
- Fake people, automated bots
 - Twitter – most of top 20 accounts are bots – 1,300 a day
 - Hunter bots – impersonate people to discredit them
- Popularity – Google – can be manipulated
 - Search for Holocaust and get Neo-Nazi
- Two drivers: make money and manipulate people

Text Analytics and Fake News

What is Fake News? Stories

- Ad for Giuliani – support me because I took good care of my mistress
- Words change meaning – globalist for Jewish bankers
- Fine line between comedy/satire and fake news
- Las Vegas shooting – Google and Facebook – top stories were fake – some from known sites – liberal anti-Trump
- Search engines more trusted than regular news
- Algorithms designed to favor popular, get most likes and comments

Text Analytics and Fake News

Fake News Techniques

- “Tens of thousands of fraudulent Clinton votes found in Ohio warehouse”
 - Add: made up person – Randall Prince, electrician
 - Photo of a ballot box (from UK), Label person in photo as Prince
 - Add story details – plan to substitute boxes for real ones
- Got 6 mil views, generated \$1,000 hr in ads
- Russians posted to protest a recent election – 2,000 bots attacked – complained posts were porno
- Twitter is worse – short, automated
- Confusion – fake news is news you don’t like

Text Analytics and Fake News

Fake News: In Context

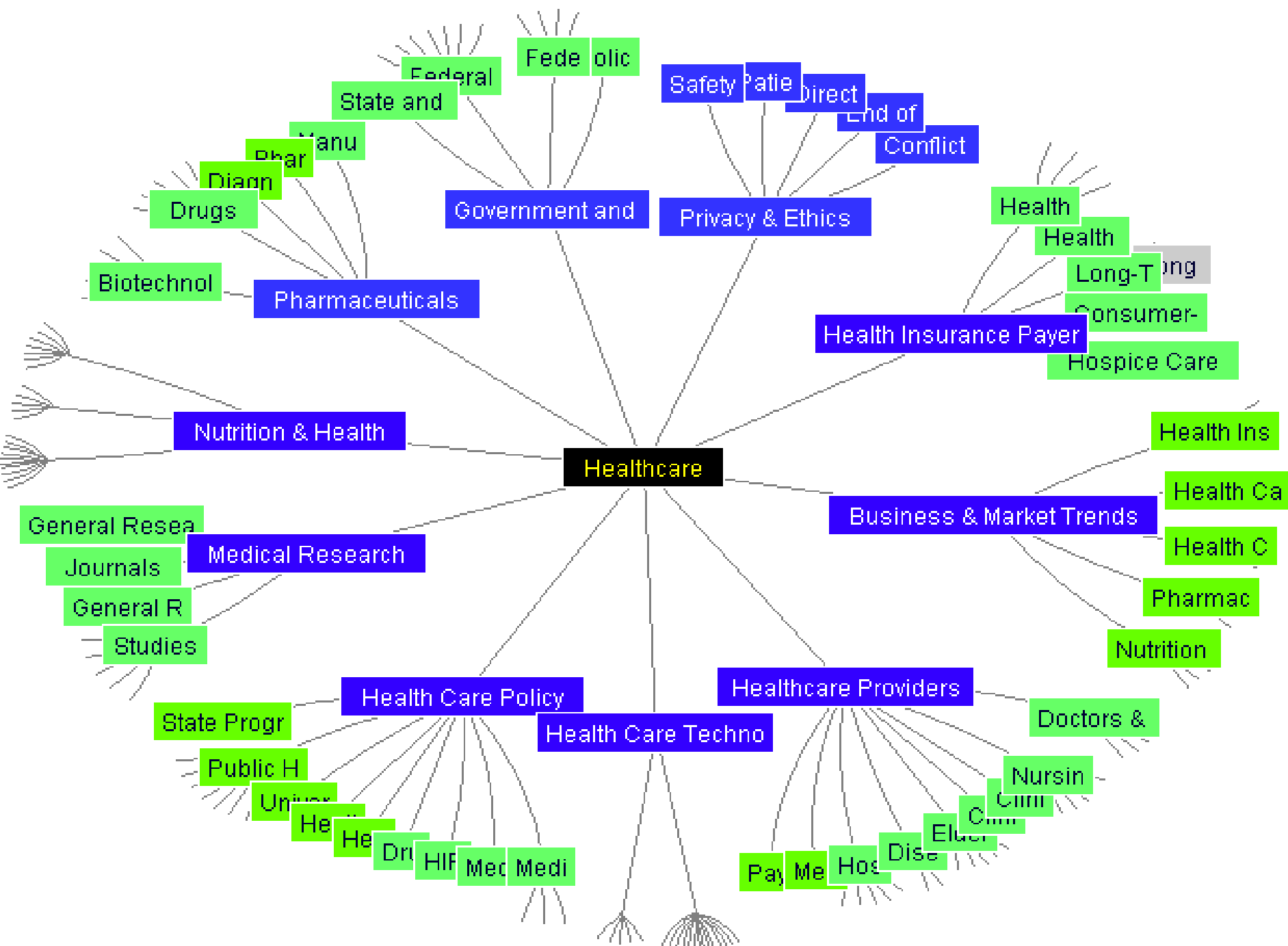
- General – blurring lines opinion & facts, content and ads
- Growth of fringe groups – finding each other, finding “facts” that support them – from hate groups to science deniers
- Polarization and echo chambers
 - Confirmation bias
 - Network effects reward extremists, unfriend = bubble
 - Politics becomes like race/religion – won’t let my daughter marry a liberal
- The Internet is making us stupid
- This could be the first really major crisis of information age
- Getting worse – future = more info, more automation, virtual reality, blurring lines between real and imagined

Text Analytics and Fake News Proposed Solutions - Partial

- Debunking
 - No money – fake news seen by millions, debunk = 1,000's
 - Only facts, but arguments won with emotion and authority
 - Effects linger – George Lakoff – Don't Think of an Elephant
- Financial: block ads
 - Advertisers not shunning like porno and gambling – yet?
 - Doesn't deter political motivations
- Technical: tool to discover “sock puppets”, multiple sites/accounts
 - Track and block known sites – URL based – abcnews.com.co, etc.
- Automated systems, machine learning, algorithms
 - Not smart enough (68% accuracy), can be manipulated
 - Black box – Watson – don't know how it works

Text Analytics and Fake News Case Study – Hybrid Analysis of News

- Inxight Smart Discovery
- Multiple Taxonomies
 - Healthcare – first target
 - Travel, Media, Education, Business, Consumer Goods,
- Content – 800+ Internet news sources
 - 5,000 stories a day
- Combination of Features – Categorization rules, Entity extraction, terms, Boolean, filters, facts
- Application – Editors get categorized results
 - Faster than human review
 - Smarter than automatic solutions



Healthcare

Government and

Privacy & Ethics

Health Insurance Payer

Business & Market Trends

Healthcare Providers

Health Care Techno

Health Care Policy

Medical Research

Nutrition & Health

Pharmaceuticals

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Category Properties

General Definition Exceptions

Category Threshold: 0.01

Terms XDOCs Rules

Use Advanced Rules

"pipeline" AND NOT SENTENCE("oil" OR "gas")

Filter results from above

Must Include Must not include

Apply filter:

Must include all filters

Filter	Case Sensitive	
New England Journal ...	<input type="checkbox"/>	▲
Journal of the America...	<input type="checkbox"/>	☰
The Lancet American	<input type="checkbox"/>	
Journal of Medicine	<input type="checkbox"/>	
British Medical Journal	<input type="checkbox"/>	
Nature	<input type="checkbox"/>	▼

- Boehringer Pilot One Drug Names Disease
 - English
 - Categorizer
 - Top
 - Diseases
 - arthritis
 - Benign Prostatic Hyperpla
 - Cancer
 - Hypertension
 - Deep Vein Thrombosis
 - HIV
 - Pulmonary Disease
 - Drug Names
 - atorvastatin
 - Aspirin

```
(OR,  
  /article/title:"[arthritis]",  
(AND,  /article/mesh:"[arthritis]",  /article/abstract:"[arthritis]"),  
(MINOC_2,  /article/abstract:"[arthritis]"),  
(START_500, (MINOC_2, "[arthritis]"))  
)
```

Text Analytics and Fake News

Pronoun Analysis: Fraud Detection; Enron Emails

- Patterns of “Function” words reveal wide range of insights
- Function words = pronouns, articles, prepositions, conjunctions, etc.
 - Used at a high rate, short and hard to detect, very social, processed in the brain differently than content words
- Areas: sex, age, personality – individuals and groups, power-status,
- Lying / Fraud detection: Documents with lies have
 - Fewer and shorter words, fewer conjunctions
 - More use of “if, any, those, he, she, they, you”, less “I”
 - More positive emotion words
- Current research – 76% accuracy in some contexts
- Text Analytics can improve accuracy and utilize new sources

Text Analytics and Fake News

Deep Text Solution - All of the Above

- Need an all of the above approach – Technical, Financial, Linguistic, Categorical
- Mainstream news works very hard to validate – Facebook needs to do it too
- Facebook Initiatives
 - Need humans – adding 3,000 editors – hybrid solution
 - Also using external organizations – Politifact, Factcheck.org, Snopes
- Text Analytics – Meaning based
 - Depth of intelligence and speed of automated
 - Human-machine partnership = smarter humans.

Text Analytics and Fake News

Deep Text Solution – Filters and Fakeness Score

- Module 1 – database of known sites,
 - Block sites & ads
- Module 2 – Deep Learning – linguistic/social patterns
 - Function words, emotional intensity, abusive language
 - Writing style and posting activity
 - Poorer quality, shorter posts – often voted down
- Module 3 – Flexible categorization rules
 - Subject – political, controversial topics
 - Emotion and motivation taxonomies
- Fakeness Categorization Score – feed to humans

Text Analytics and Fake News Solutions That Work

- All that helps but ultimate solution is education of society
- Who would believe Clinton had a child prostitute ring in a pizza place?
- Need to educate people to spot fake news and give them tools
- Real Time Debunking
 - Automated context
 - Linked Data – quick check – needs to be smarter
- Smart Reader / research assistant– see Deep Text
 - Automate some tasks, enrich others
- Ongoing war – as we develop better techniques, fakes will adapt

Text Analytics and Fake News Resources

- Conferences – Information Today in Wash. DC
 - KMWorld, Taxonomy Boot Camp
 - Text Analytics Forum – New!
- Politifact.org / Google Fact Check / Factcheck.org
- Snopes – from urban legends to fake news
- Books:
 - Weaponized Lies: How to Think Critically in the Post-Truth Era by Daniel J. Levitin
 - Don't Think of an Elephant by George Lakoff
 - Post-Truth: How Bullshit Conquered the World by James Ball

Text Analytics and Fake News

Conclusions

- Fake News is real and really serious
 - Undermine democracy, communication, civilization?
- Multiple factors driving more fake news – money, political, ease of technology & scale, it’s “fun”
- Solutions require all of the above
 - Major initiatives from Facebook, Twitter, etc.
 - Multiple levels – technical, business, government
- Hybrid human-machine solutions – using text analytics
- Ultimate answer is better education

Questions?

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