

Novelty Detection in Text Corpora

An Illustration from Case Study 1

Hillary Clinton Controversies, Spring-Summer 2015

Text Analytics (PREDICT 453)
Northwestern University
School of Professional Studies
Master of Science in Predictive Analytics
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Novelty Detection:

Hillary Clinton and Alleged Classified Emails

How Do We Discern Novelty in the New Document?

Prototype for Precursor Document Set

“Hillary Clinton emails: Did she do anything wrong or not?”



Jeremy Diamond and Elise Labott, CNN
Weds., March 11, 2015

<http://www.cnn.com/2015/03/06/politics/hillary-clinton-emails-was-there-wrongdoing/>

New Document:

“Hillary Clinton emails said to contain classified data”



Michael S. Schmidt and Matt Apuzzo
The New York Times
July 24, 2015

http://www.nytimes.com/2015/07/25/us/politics/hillary-clinton-email-classified-information-inspector-general-intelligence-community.html?_r=0

Step 1: Find the Most-Similar Set of Documents

New document



Compare with SETS of similar documents



Each Set of Similar Documents Has a Prototype Vector

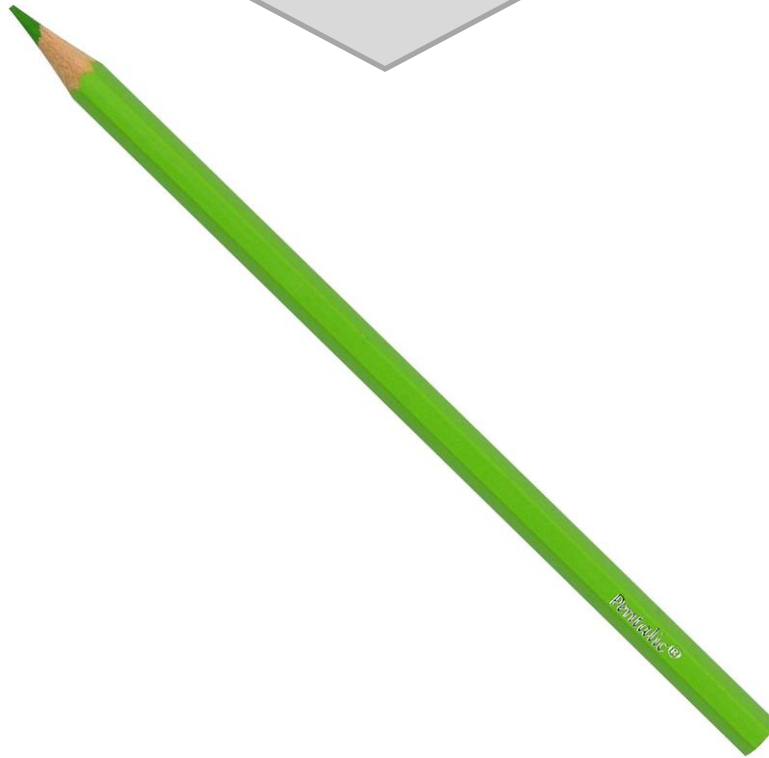
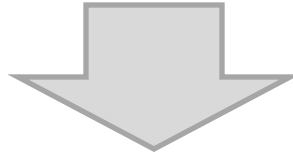
Small Set of
Prototype Vectors



Documents Grouped According
to Best-Matching Prototype



Start by Finding the Best-Matching Prototype Vector



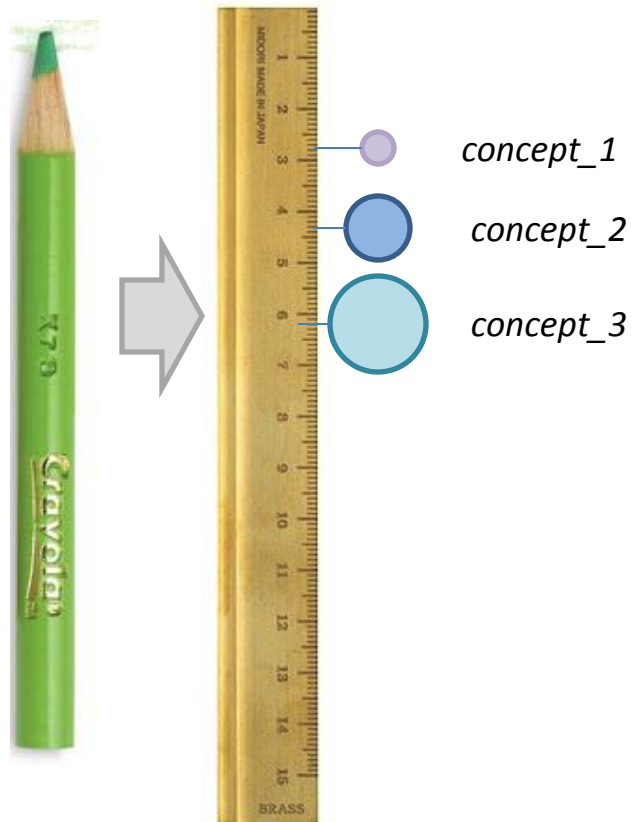
Then: *Find Out What Is New and Different*

Document Matching: Using Vector-Matching to Find Document Similarity

The Big Question: Match Using Terms or Concepts?

- Hundreds of terms per document
- Dozens of concepts per document
- **Concepts are more concise representation**

Vector representation of document content



Document matching uses vector similarity

- **Documents are similar if components are:**
 - Similar in nature
 - Similar in relative strength
- **HOWEVER**, vector matching algorithms require ***matching vector element fields*** => a ***TOUGH CONSTRAINT!***

It is easier to match documents using concepts than terms

- **Concepts condense relevant terms** into more compact and precise units
- **Concepts are more general**, and many terms contribute to each concept
- **Concept-matching has less error** than term-matching when determining document similarity

Example:

Document Matching and Novelty Detection

Starting Point:
An Exemplar Document:

“Hillary Clinton emails: Did she do anything wrong or not?”

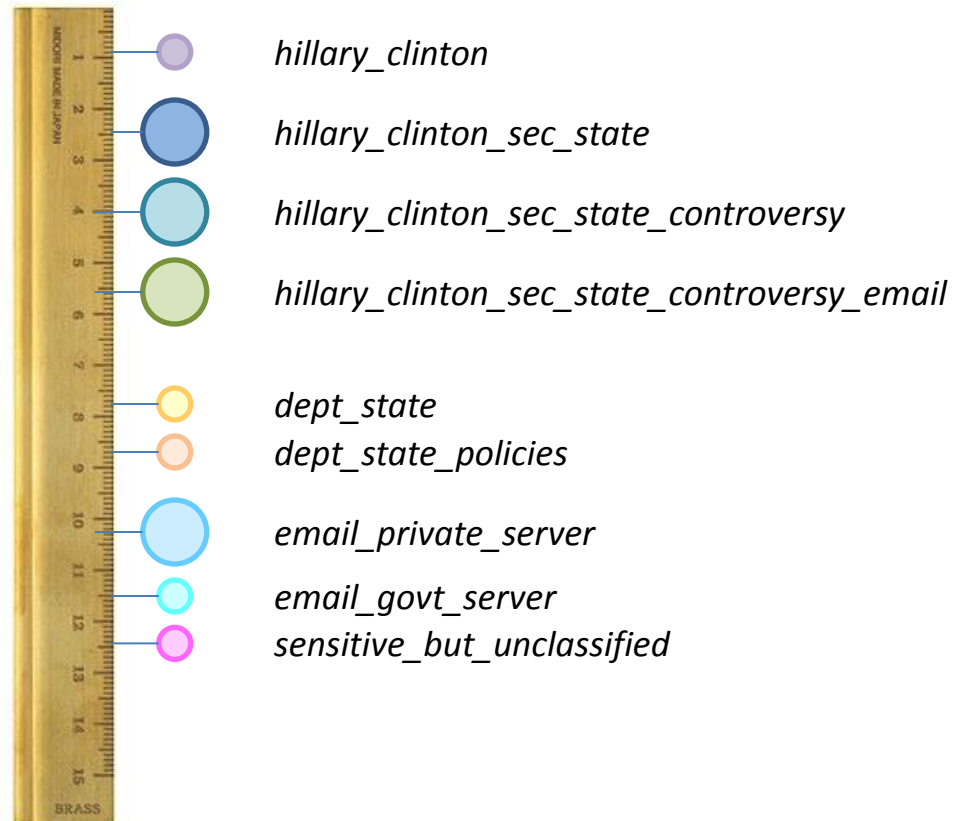


Jeremy Diamond and Elise Labott, CNN
Weds., March 11, 2015

<http://www.cnn.com/2015/03/06/politics/hillary-clinton-emails-was-there-wrongdoing/>

Document Concepts: *Vector representation of document content*

Concepts found in this document
(illustrative subset)



Novelty Detection: Discerning “Significant Newness”

New Document:

“Hillary Clinton emails said to contain classified data”

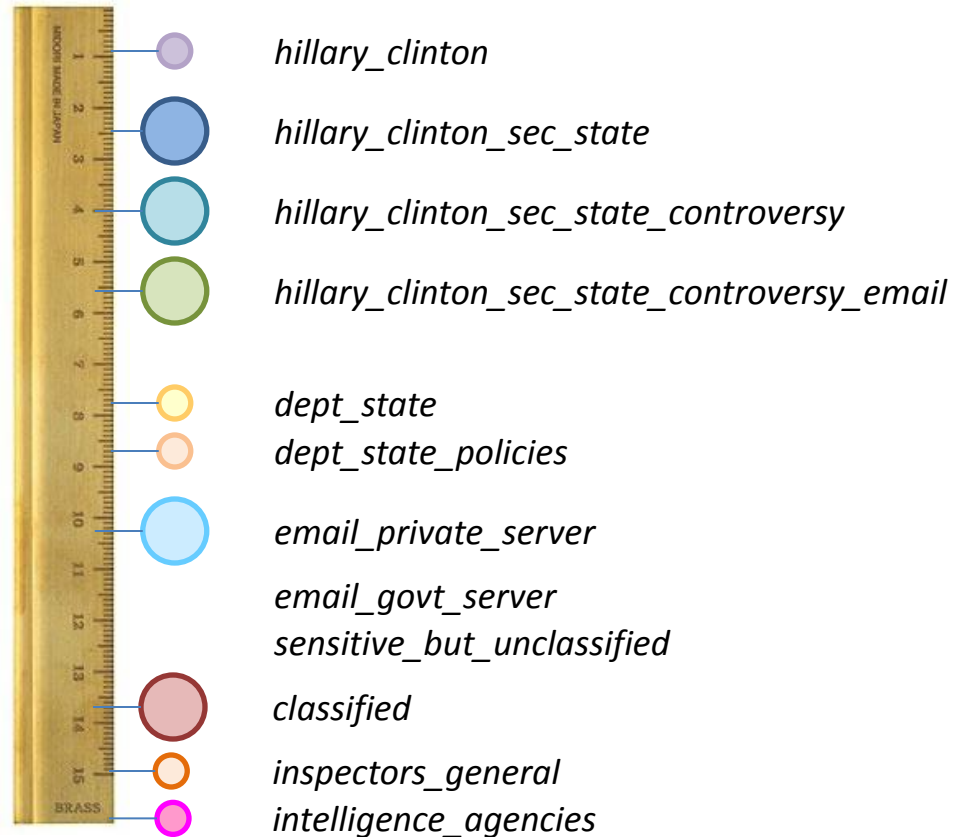


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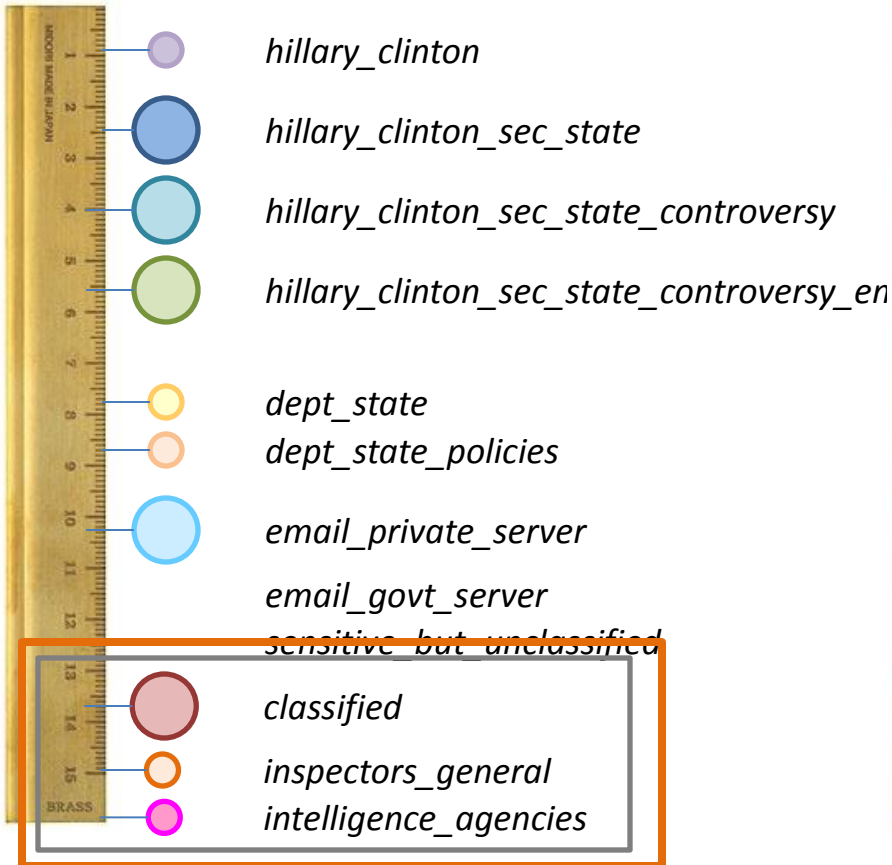
Document Concepts: *Vector representation of document content*

Concepts found in this document
(illustrative subset)

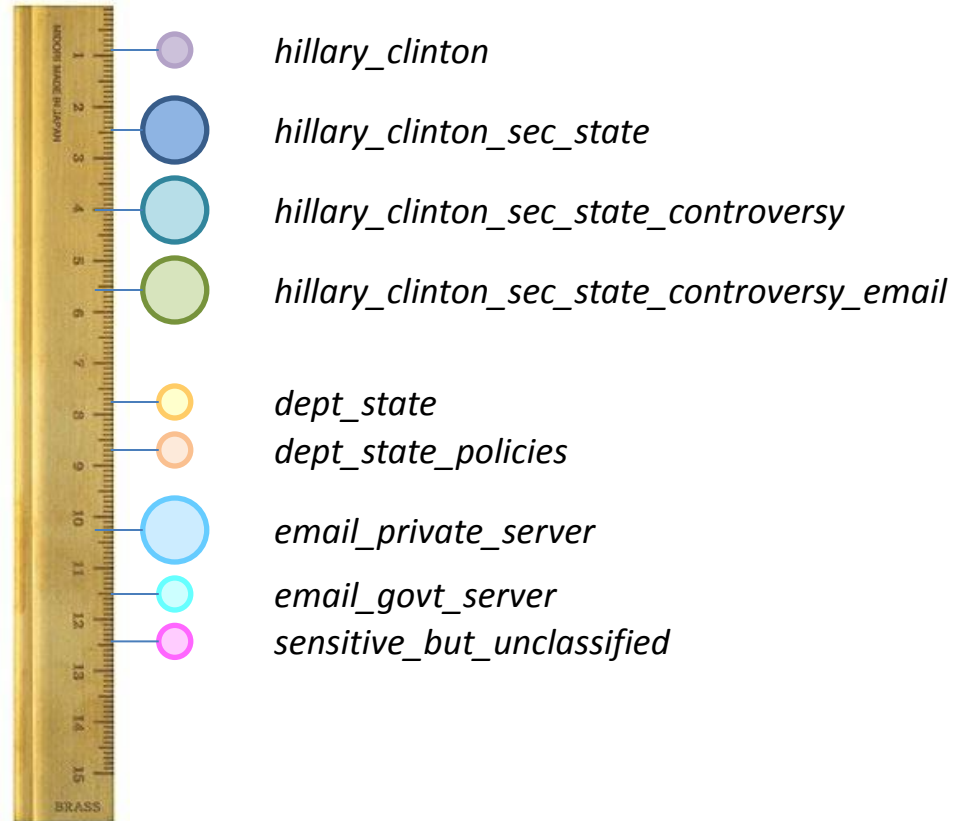


Novelty Detection: Compare New Document Concepts Against Prototype

New Document Concepts



Closest Match: Prototype Document Concepts

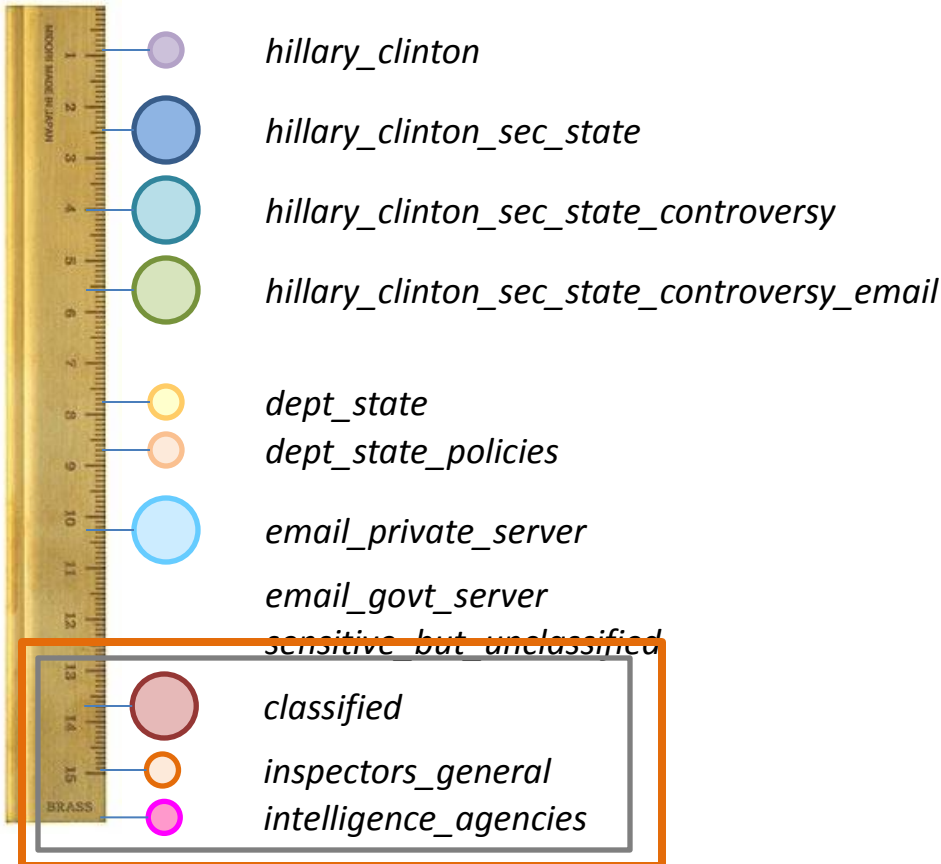


New concepts (or even unassigned terms) versus prototype

Summary

Novelty Detection Happens When We Find Difference Against Known Prototypes

New Document Concepts



Summary of Steps

- Find closest prototype match
- Find key differences
- Test for strength
 - Numbers of similar new documents
 - Strengths of new terms
- **Threshold to flag novelty**

In This Particular Case

- Breaking news, multiple channels
- Consistent use of new terms: “classified,” “inspectors general,” “intelligence agencies”
- Significant difference from prototype: “sensitive but unclassified”

Result: Novel Terms Detected

The Crucial Question

We detected novelty because we matched a document against its best-matching prototype and found differences

We matched using concepts, rather than terms, because it gave us a more accurate prototype match

So the **BIG QUESTION** is:
How do we go from terms to concepts?

The ANSWER:
*This is the toughest task in text analytics.
Look for the SEQUEL – coming soon!*