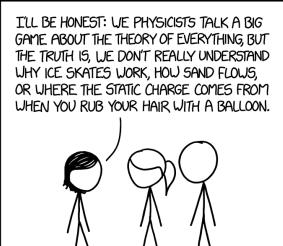
xkcd.com/1867/



#### Otilizing Knowledge Bases for Text Retrieval: A Wishlist to Lext Betrieval: A Mishlist Otilizing Knowledge Bases

Laura Dietz dietz@cs.unh.edu



**University of New Hampshire** 



# KG4IR https://kg4ir.github.io

#### Utilizing Knowledge Graphs in Text-centric Information Retrieval

Laura Dietz (@laur University of New Har Requiring long, complex answers

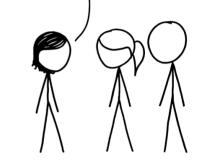
Intended queries:

- how ice skates work
- UK leaving Europe
- cashflow important for investment
- effects of water pollution
- Diesel scandal affect Daimler AG

# If yes, why? If not, why not? Causes? Involvements? Controversy? Backstory? What do I need to know to understand the answer?

Laura Dietz dietz@cs.unh.edu Utilizing Knowledge Bases for Text Retrieval: A Wishlist

I'LL BE HONEST: WE PHYSICISTS TALK A BIG GAME ABOUT THE THEORY OF EVERYTHING, BUT THE TRUTH IS, WE DON'T REALLY UNDERSTAND WHY ICE SKATES WORK, HOW SAND FLOWS, OR WHERE THE STATIC CHARGE COMES FROM WHEN YOU RUB YOUR HAIR WITH A BALLOON.





xkcd.com/1867/

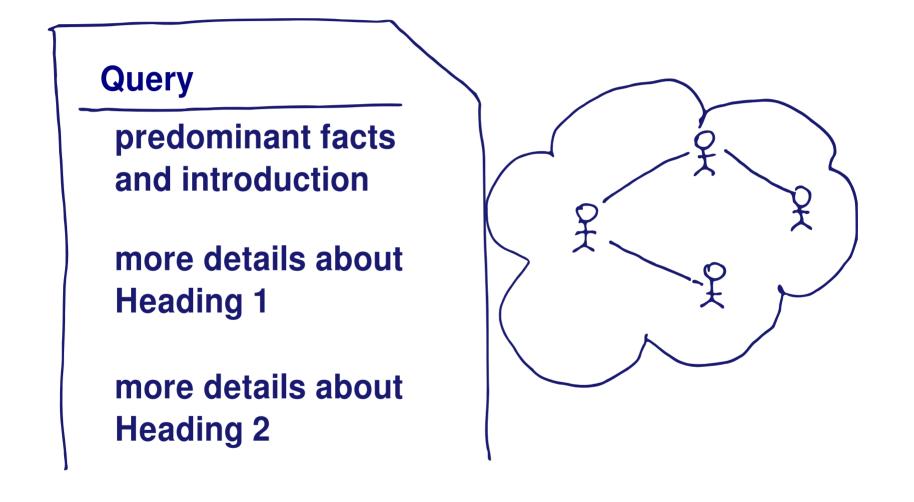


# What is the problem? ...and the solution?

Wikipedia	Web Search
Not enough / recent information	Manually sift through many web pages
Train computers to recycle Web content to write a comprehensive articles in response to a search query	

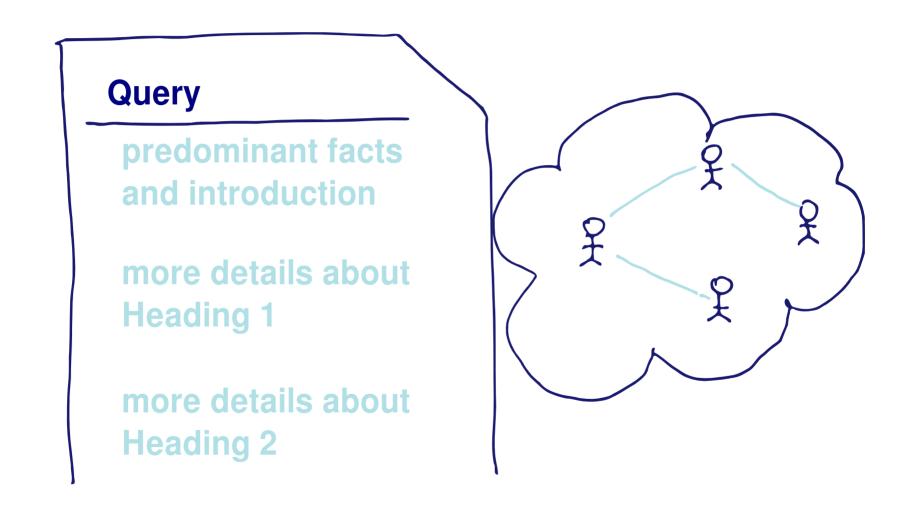


# **Query-specific Article + Knowledge Graph**



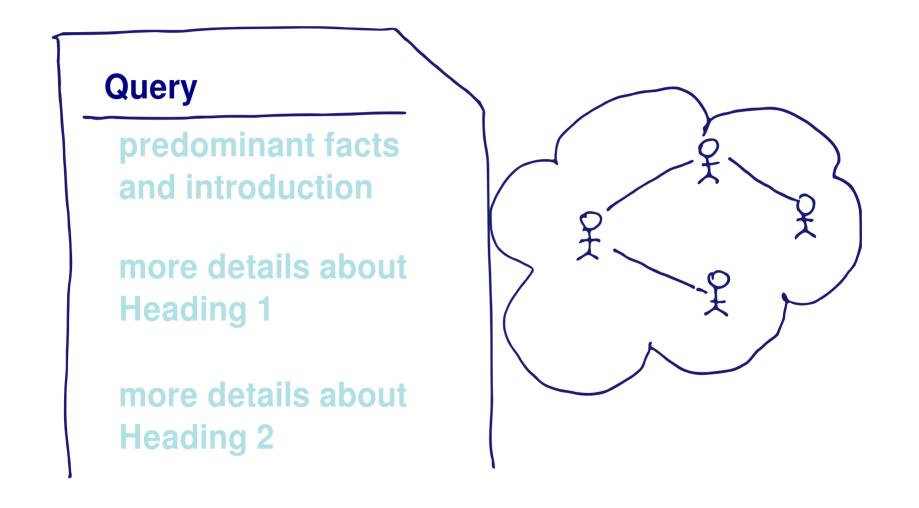


# **Step 1: Find Relevant/Central Entities**



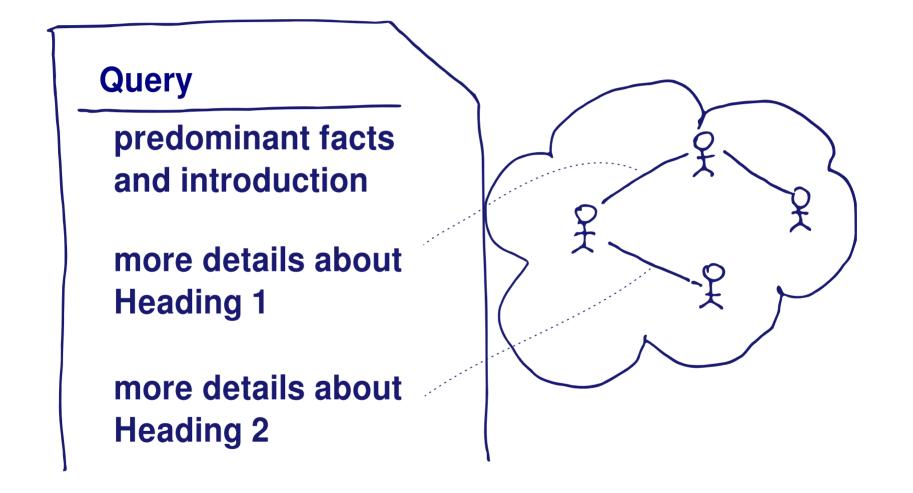


## **Step 2: Find Relevant Relations**





## **Step 3: Find Relevant Text + Consolidate**



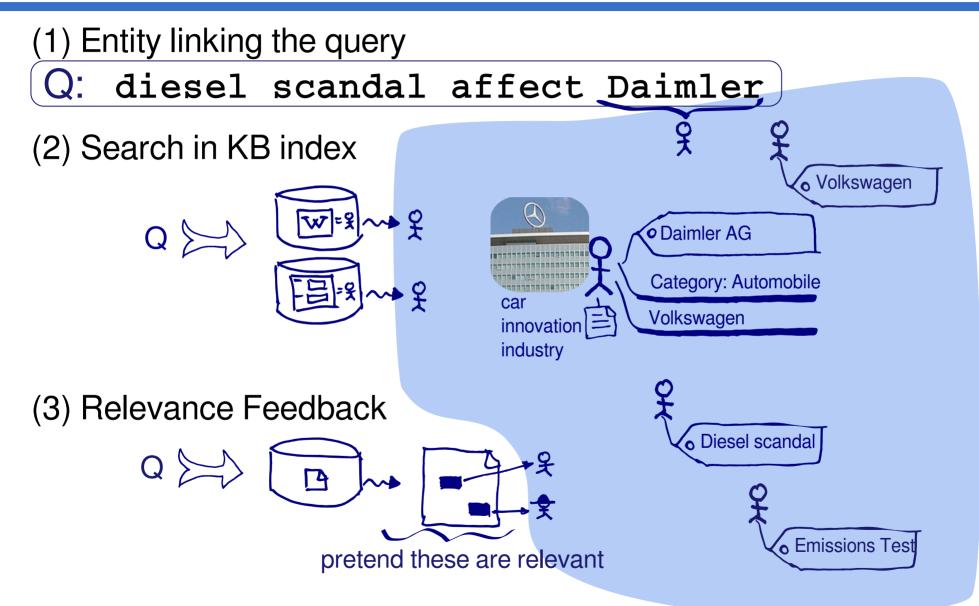


### **How to Find Relevant Entities?**

#### Q: diesel scandal affect Daimler

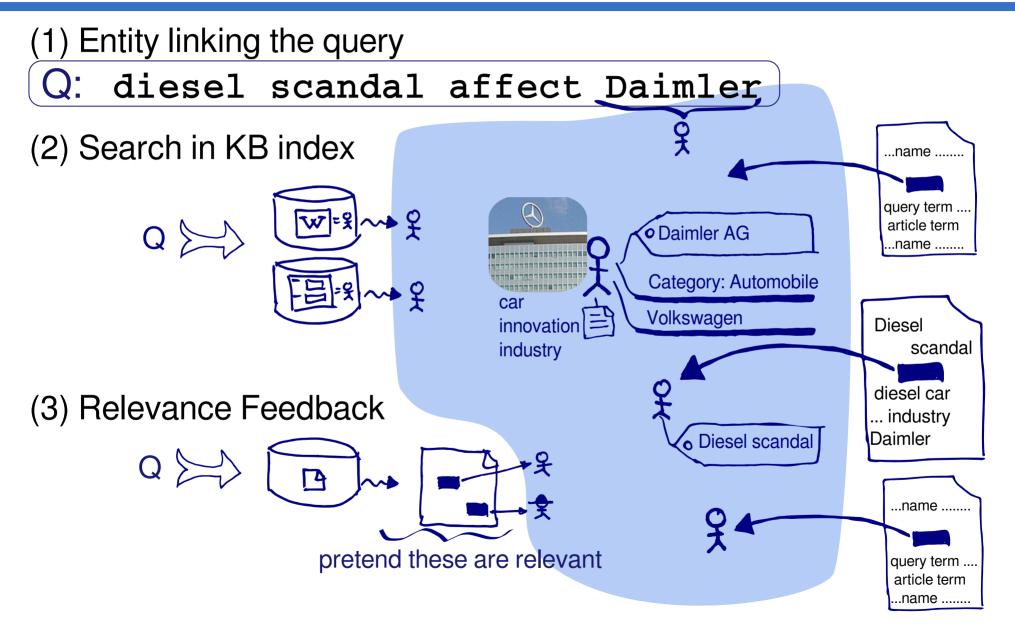


# **How to Find Relevant Entities?**



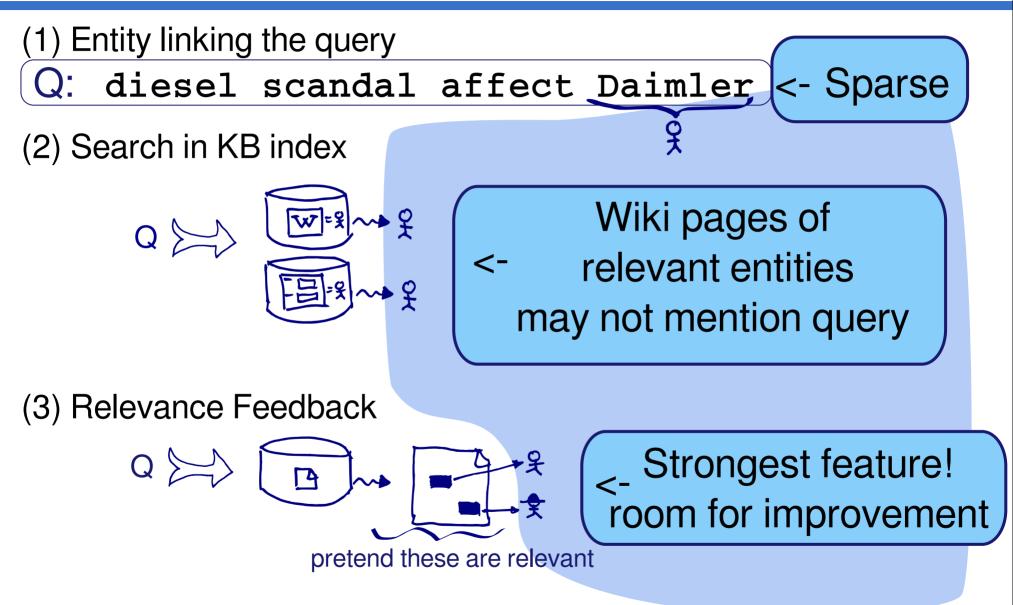
#### [Dalton, Dietz, Allan 14]

# How to Use Entities for Text Ranking?





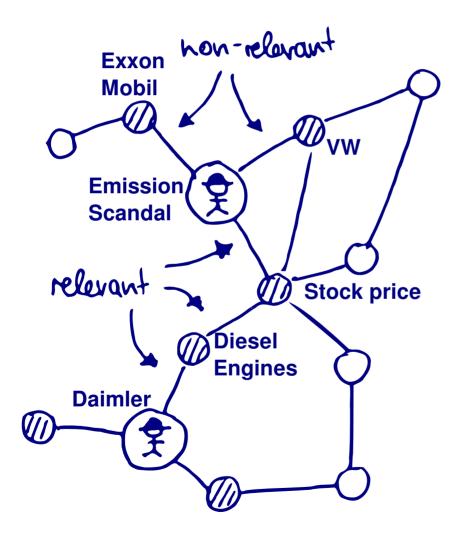
# **Finding Relevant Entities: What Works?**





# **Identifying Relevant Relations in a KG**

Q: diesel scandal affect Daimler



Naive approach: Select sub-KG of relevant entities.

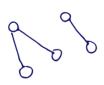
So many connections in a knowledge graph

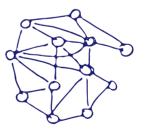
- Some are relevant!
- But many are only relevant in a certain (other?) context.

KGs started with the "most popular" facts then it grew in number of nodes and number of connections, aiming for better coverage.

KGs in 2013

KGs in 2019





# Hub nodes: New York City, California, United States

KGs started with the "most popular" facts then it grew in number of nodes and number of connections, aiming for better coverage.

KGs in 2013

KGs in 2019



# Hub nodes: New York City, California, United States

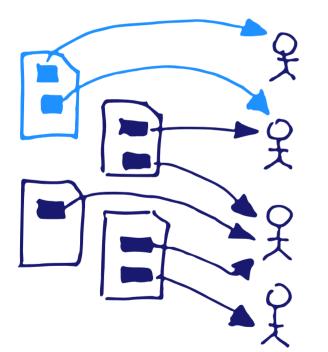
KGs started with the "most popular" facts then it grew in number of nodes and number of connections, aiming for better coverage.



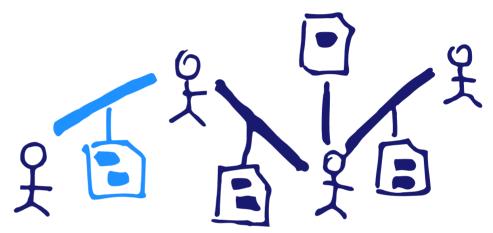
### Hub nodes: New York City, California, United States



(1) Retrievetext + entity linksand entities





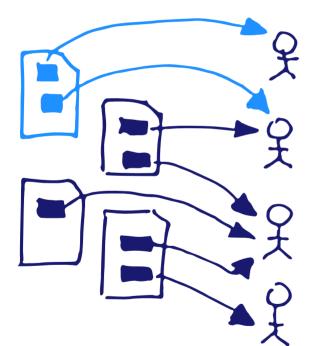


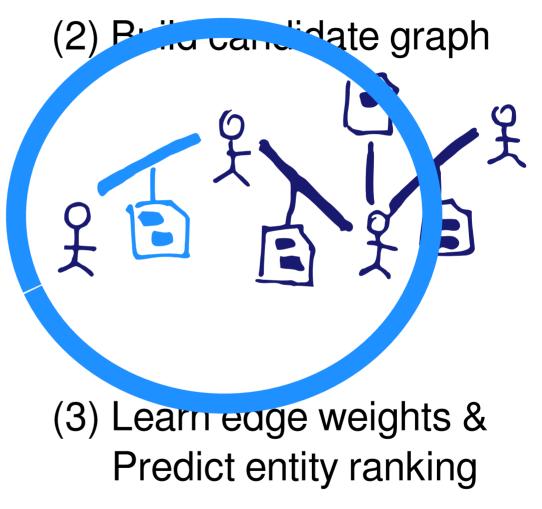
(3) Learn edge weights & Predict entity ranking

# **ENT Rank for Entity Ranking**



(1) Retrievetext + entity linksand entities



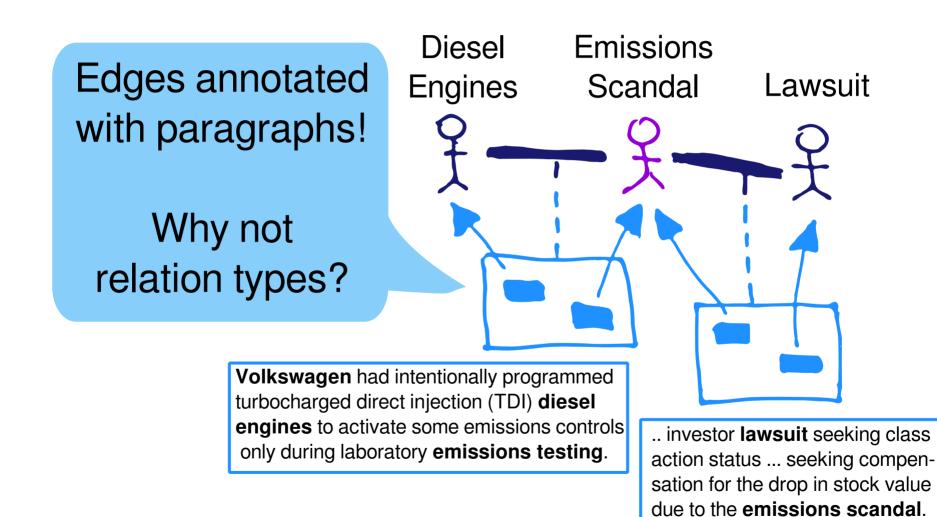




due to the emissions scandal.

Features: Emissions Diesel Engines Lawsuit Scandal Entity Neighbor 2 - 2Text Volkswagen had intentionally programmed turbocharged direct injection (TDI) diesel engines to activate some emissions controls .. investor lawsuit seeking class only during laboratory emissions testing. action status ... seeking compensation for the drop in stock value

# ENT Rank for Entity Ranking [Dietz, SIGIR 19]

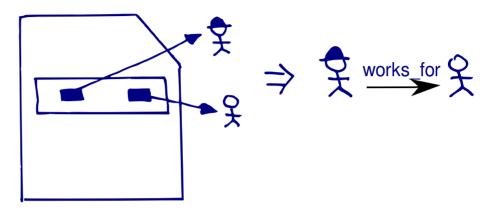


[Schuhmacher, Roth, Ponzetto, Dietz 16



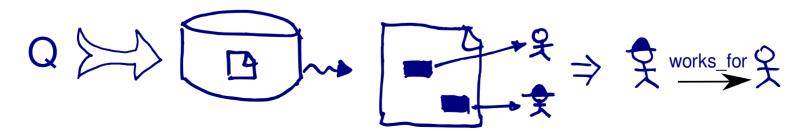
# **Extracting Relevant Relations**

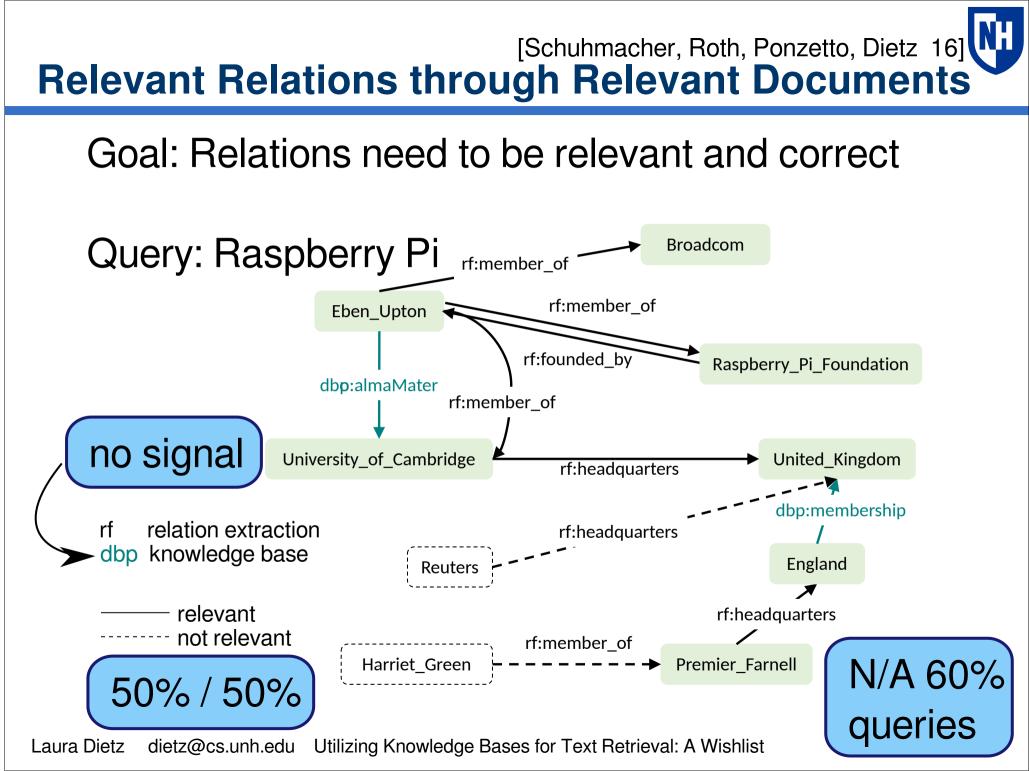
Relation Extraction: [Roth et al 14] (best at TAC KBP 13)



#### Research question:

relevant documents + extraction = relevant relations? [Schuhmacher, Roth, Ponzetto, Dietz 16]



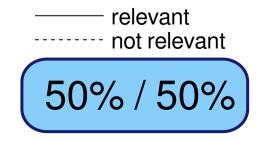




# **Issue 1: Correct Vs. Relevant Extractions**

Goal: Relations need to be relevant and correct only considering correct extractions....

Schema-based:50% relevant[Schuhmacher 16]OpenIE-based:50% relevant[Kadry & Dietz 17]Human-based:50% relevant(sentence-level)

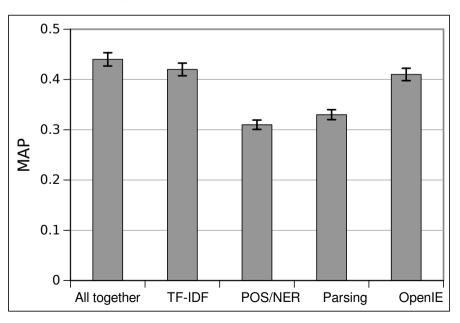




# **Issue 2: Coverage of Relation Extractions**

Schema-based: N/A for 60% of queries (TAC KBP 13)

Open IE: 5% sentences with correct annotations (no coref) Leads to only marginal improvements for IR, e.g. Ranking entity-query support sentences for relevance.



# Interesting relations are a bit more complicated.

Volkswagen had intentionally programmed turbocharged direct injection (TDI) diesel engines to activate some emissions controls only during laboratory emissions testing. .. investor **lawsuit** seeking class action status ... seeking compensation for the drop in stock value due to the **emissions scandal**.

Beyond more than one sentence. Include multiple intermediate entities. ...also not just triples + coref...

# **Data: Effects of Water Pollution/Eutrophication**

#### 6

The **pollution** often comes from non point sources such as agricultural runoff, wind-blown debris and dust. Nutrient **pollution**, a form of **water pollution**, refers to contamination by excessive inputs of nutrients. It is a primary cause of **eutrophication** of surface **waters**, in which excess nutrients, usually nitrogen or phosphorus, stimulate algae growth.

#### 7

Nutrient **pollution**, a form of **water pollution**, refers to contamination by excessive inputs of nutrients. It is a primary cause of eutrophication of surface waters, in which excess nutrients, usually nitrogen or phosphorus, stimulate algal growth. Sources of nutrient **pollution** include surface runoff from farm fields and pastures, discharges from septic tanks and feedlots, and emissions from combustion. Excess nutrients have been summarized as potentially leading to:

#### 8

Human interference in the phosphorus cycle occurs by overuse or careless use of phosphorus fertilizers. This results in increased amounts of phosphorus as **pollutants** in bodies of **water** resulting in **eutrophication**. Eutrophication devastates **water** ecosystems by inducing anoxic conditions.

#### Ask me for the data ...

#### http://trec-car.cs.unh.edu/

Up next:

Multi-paragraph

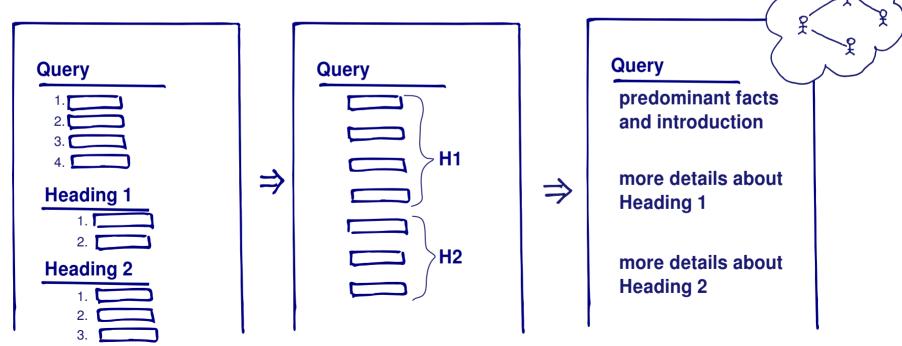
summarization

+ query-KG



## Given: query Q and outline of Headings

CAR Y1, Y2: Paragraph ranking per heading. Optimize relevance CAR Y3: Paragraph ordering. Maximize coverage, topical coherence







General purpose schema with many types

My Wishlist

6

High coverage/recall (40%?)

Extraction of complex relations (not just triples + coref)

Bridging existing KGs with text

Relevant information extraction

Query-specific knowledge graphs

TREC CAR Dataset http://trec-car.cs.unh.edu/ Ask me for a data set to play around with... Laura Dietz dietz@cs.unh.edu Utilizing Knowledge Bases for Text Retrieval: A Wishlist