49th International Conference on Very Large Data Bases

# **Effective Entity Augmentation By Querying External Data Sources**

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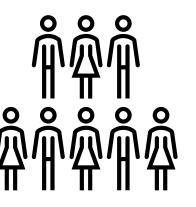


Portland State



# **Drug Repositioning Can Save Lives**





Patients with Castleman's disease

• Rare disease

• Potentially fatal: causes <u>severe inflammation</u>

No effective treatments currently exist



Unfortunate reality:

**Too rare**: no financial incentive for companies to develop treatments

#### Alternative:



Find an existing drug to treat Castleman's disease



# **Identify a Candidate Drug**

#### Find a candidate drug

| FDA-Appro    | ved Drugs     |                         |
|--------------|---------------|-------------------------|
| brand_name   | class         | uses                    |
| Humira       | TNF inhibitor | rheumatoid<br>arthritis |
| Enbrel       | TNF inhibitor | plaque Biomedical       |
| Local Data S | ource         | Researcher              |

Castleman's causes severe inflammation...

**Humira** is used to treat conditions involving <u>severe inflammation</u>

Candidate drug: Humira

Next step: gather more information about Humuria:Will it help or hurt?



#### Local entity:

|   | brand_name             | class         | uses                            |
|---|------------------------|---------------|---------------------------------|
| > | Humira                 | TNF inhibitor | rheumatoid<br>arthritis         |
|   |                        |               |                                 |
|   |                        |               |                                 |
|   |                        |               |                                 |
|   |                        |               |                                 |
|   | FDA-Appr               | roved Drugs   |                                 |
|   | FDA-Appr<br>brand name |               | uses                            |
|   |                        |               | uses                            |
|   | brand name             | class         | uses<br>rheumatoid<br>arthritis |

#### Generic Drugs

| generic_name | adverse_effects                        |
|--------------|--|
| Adalimumab   | After treatment with adalimumab        |
| Etanercept   | Etanercept binds specifically to tumor |

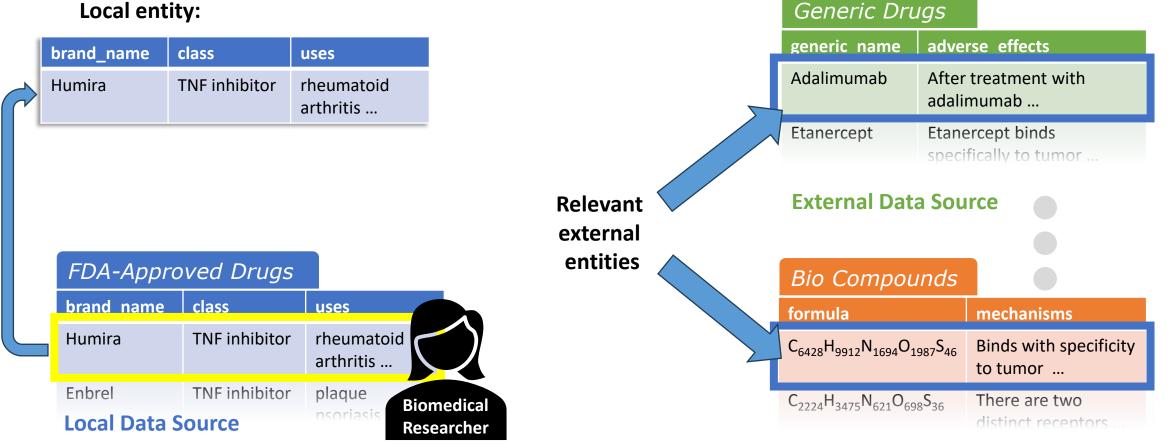
| External Data Source                     |                                     |  |  |
|--|-------------------------------------|--|--|
|  |                                     |  |  |
| Bio Compounds                            |                                     |  |  |
| formula                                  | mechanisms                          |  |  |
| $C_{6428}H_{9912}N_{1694}O_{1987}S_{46}$ | Binds with specificity to tumor     |  |  |
| $C_{2224}H_{3475}N_{621}O_{698}S_{36}$   | There are two<br>distinct receptors |  |  |

#### **External Data Source**



## What we Want: Info Relevant to Humira

#### Local entity:



**External Data Source** 



## **Augment Humira With that Relevant Info**

|                     |                  |                            |                                 | - |
|---------------------|------------------|----------------------------|---------------------------------|---|
| brand_name          | class            | uses                       | adverse_effects                 | G |
| Humira              | TNF inhibitor    | rheumatoid<br>arthritis    | After treatment with adalimumab |   |
|                     | <b>Geo</b> mecha | inisms                     |                                 |   |
| $\bigcirc \bigcirc$ | Binds to tur     | with specificity           |                                 |   |
| FDA-Appr            | roved Drug       | 'S                         |                                 |   |
| brand_name          | class            | uses                       |                                 |   |
| Humira              | TNF inhibito     | or rheumatoid<br>arthritis | $\bigcirc$                      |   |
| Enbrel              | TNF inhibito     | or plaque                  | Biomedical                      |   |

Researcher

**Local Data Source** 

#### Generic Drugs

| generic name                     | adverse effects  |  |
|----------------------------------|--|--|
| Adalimumab                       | Adalimumab After treatment with adalimumab                     |  |
| Etanercept                       | Etanercept binds specifically to tumor                         |  |
| External Data Source             |  |  |
| Bio Compou                       | inds   |  |
|                                  |  |  |
| formula                          | mechanisms   |  |
| $C_{6428}H_{9912}N_{1694}O_{19}$ | <sub>987</sub> S <sub>46</sub> Binds with specificity to tumor |  |

 $C_{2224}H_{3475}N_{621}O_{698}S_{36}$ 

There are two

#### **External Data Source**



## **Manually Querying for Relevant External Entities**

## Challenges:

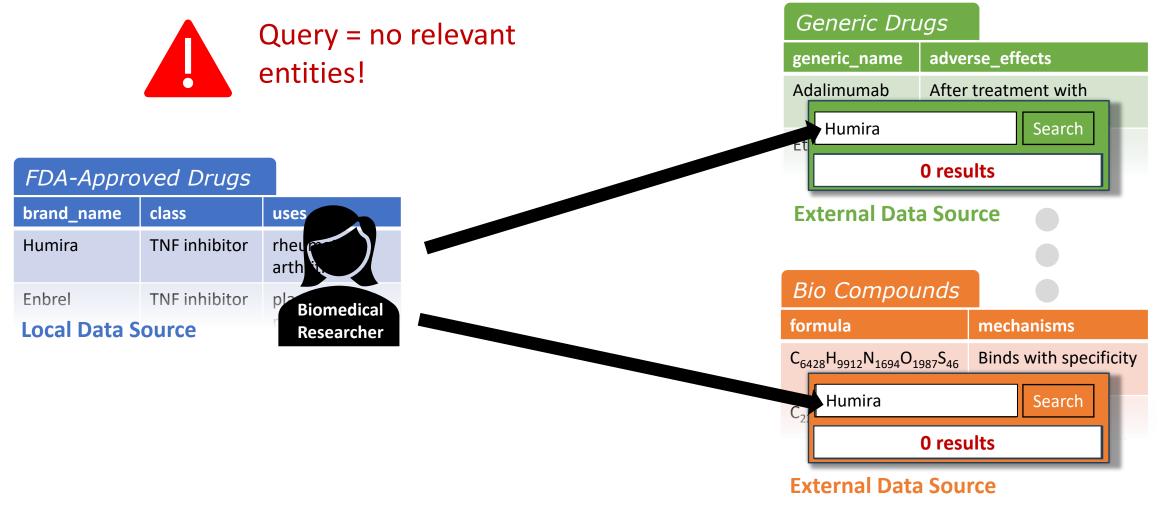
- Many external data sources
- Data heterogeneity: different representations
  - Humira = Adalimumab
    - $= C_{6428} H_{9912} N_{1694} O_{1987} S_{46} = ???$

| FDA-Appro    | ved Drugs     |                         |
|--------------|---------------|-------------------------|
| brand_name   | class         | uses                    |
| Humira       | TNF inhibitor | rheumatoid<br>arthritis |
| Enbrel       | TNF inhibitor | plaque Biomedical       |
| Local Data S | ource         | Researcher              |

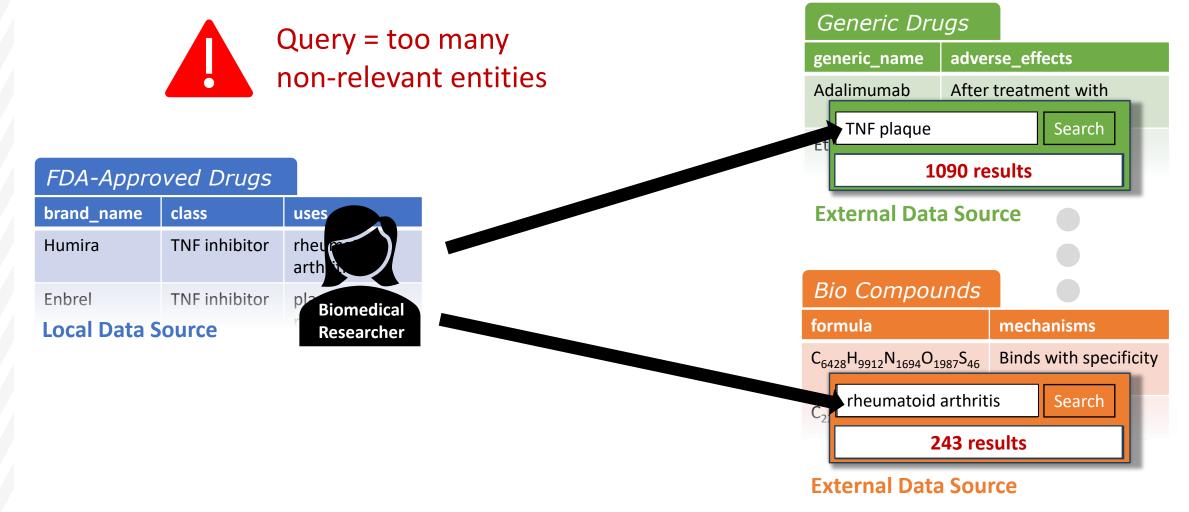
| Generic Dru  | ıgs                              |                        |
|--|----------------------------------|------------------------|
| generic_name   | advers                           | e_effects              |
| Adalimumab   | After t                          | reatment with          |
| Et   |                                  | Search                 |
|  |                                  |                        |
| External Data  | a Sour                           | ce                     |
|  |                                  |                        |
| Віо Сотрои   | Inds                             |                        |
| formula  |                                  | mechanisms             |
| C <sub>6428</sub> H <sub>9912</sub> N <sub>1694</sub> O <sub>1</sub> | <sub>987</sub> S <sub>46</sub> I | Binds with specificity |
| C <sub>21</sub>  |                                  | Search                 |
|  |                                  |                        |
| <b>External Data</b>   | Sourc                            | ce                     |

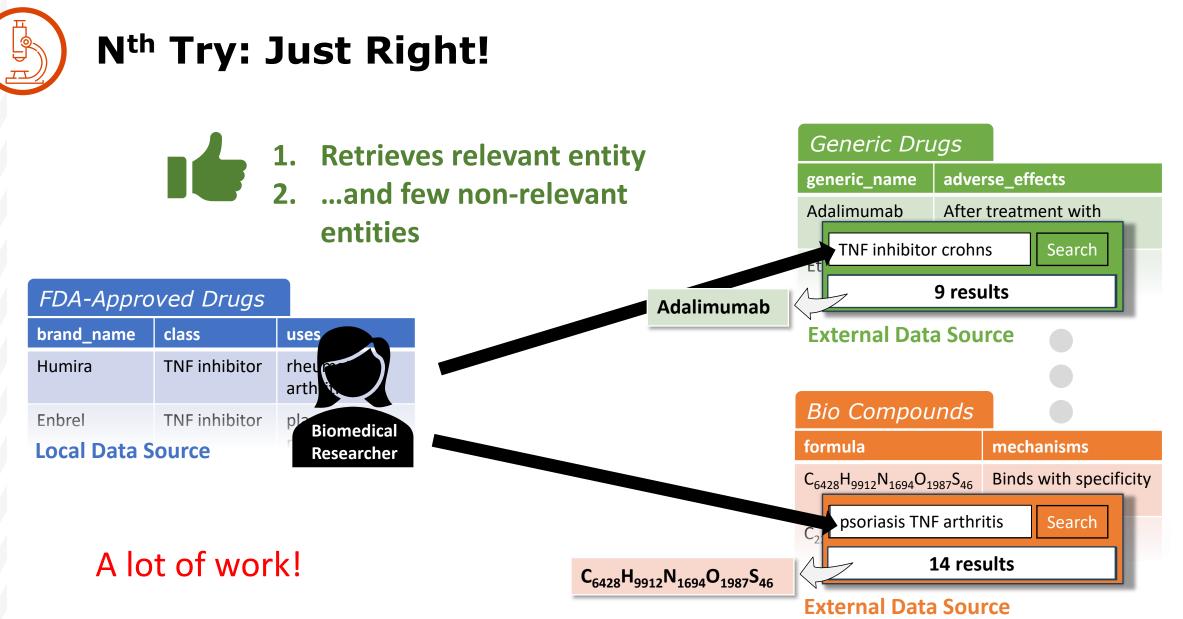


## 1<sup>st</sup> Try: Query = Too Specific to Local Source











# **Alternative: Use a Mediator**

Query on behalf of the user:

FDA-Approved Drugs

class

**TNF** inhibitor

**TNF** inhibitor

brand name

Local Data Source

Humira

Enbrel

1. User specifies *local* entity for augmentation

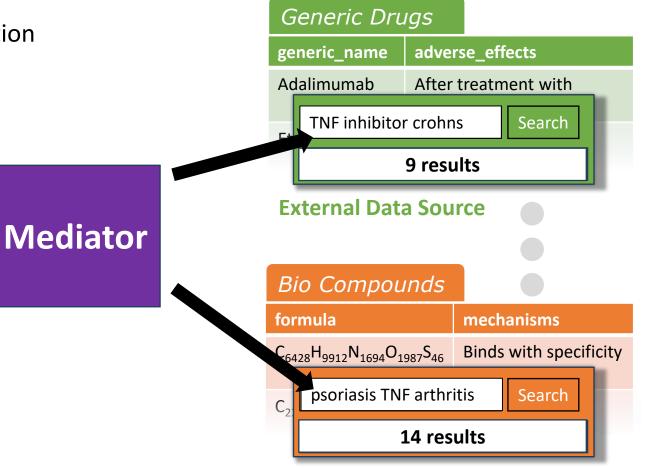
uses

rheu arth

pla

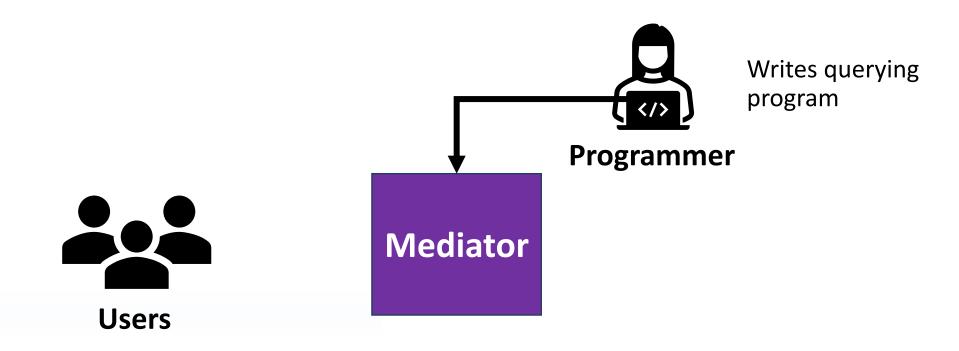
User

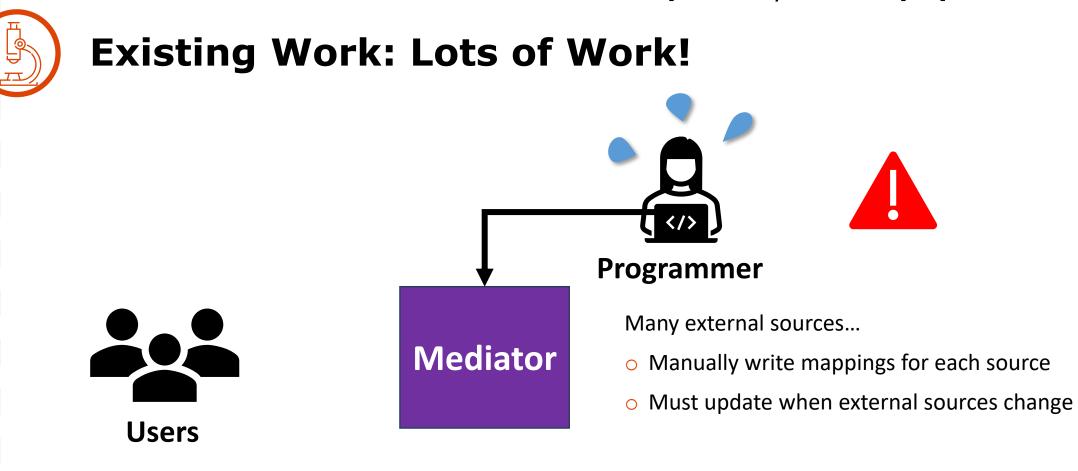
2. Mediator retrieves relevant information from external sources

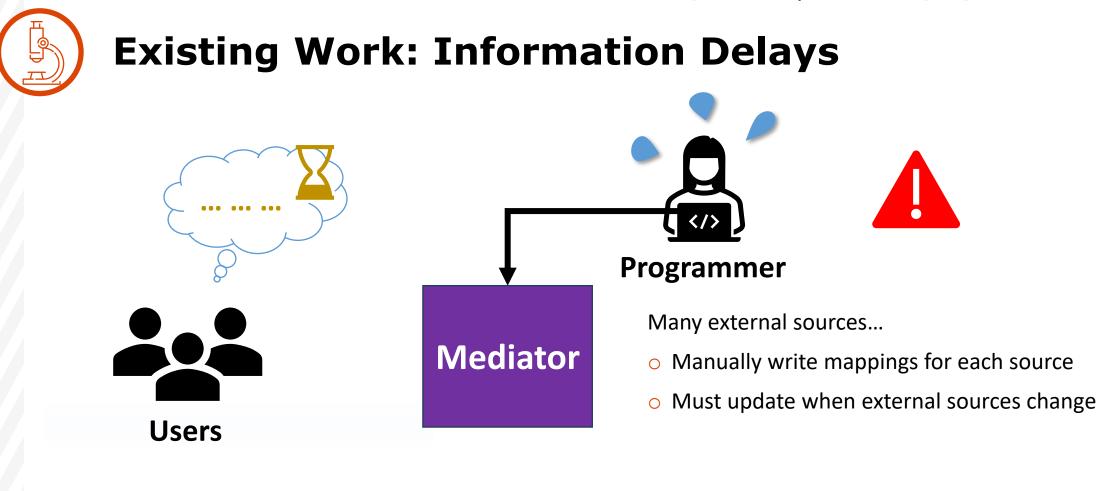


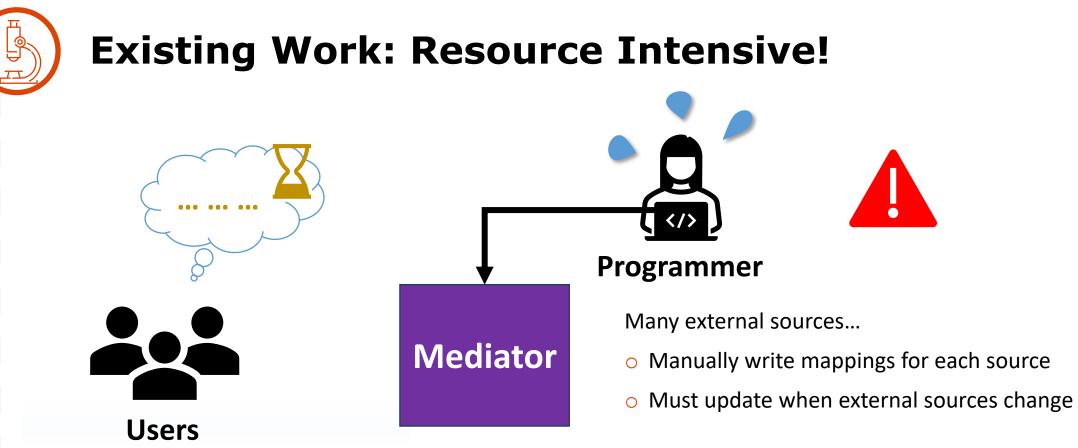


## **Existing Work: Mediator Written By Hand**







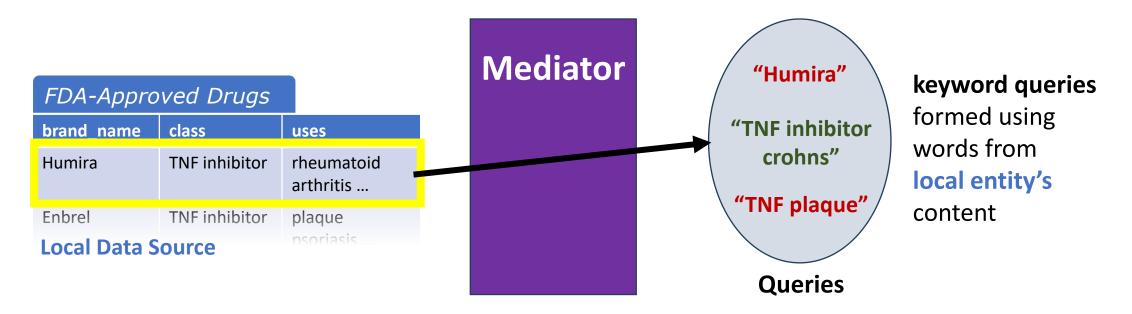


For example: the NIH funds a consortium of such systems (~14 systems)

- Just one system has 73 external datasources and millions of entities
- o Costs NIH US\$923 million per year!



## Learn Mediator that maps local entity $\rightarrow$ "Just right" query





# **How Do We Learn the Mediator?**

### **Offline Learning:**

- 1. Gather training data
- 2. Train mediator
- 3. Users query mediator



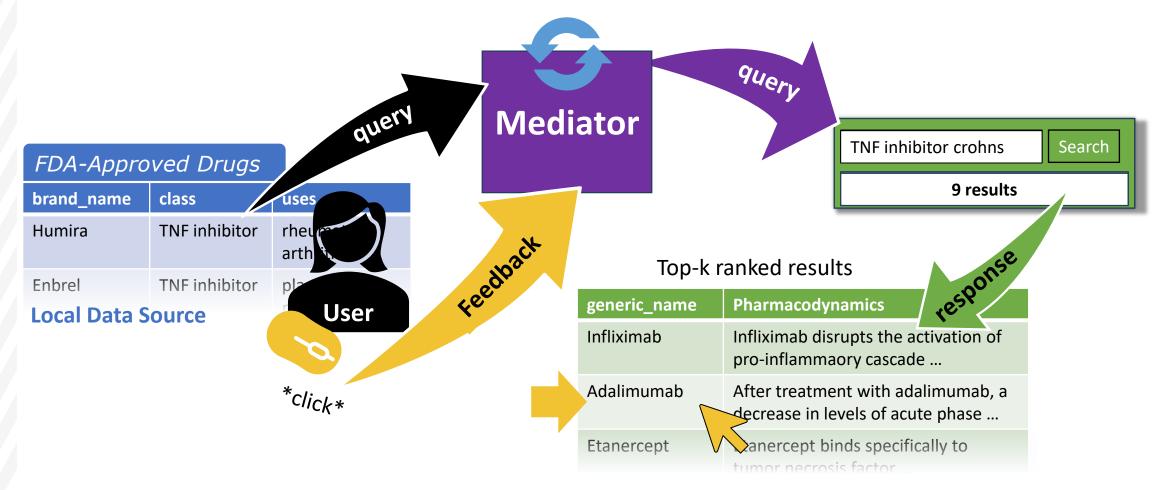
- Hire domain experts to label data
- External source updates  $\rightarrow$  must repeat!
- Still delays...

### **Online Learning:**

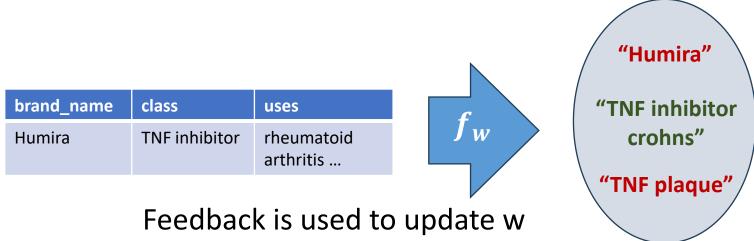
o Train mediator while users query it



Refine understanding of what makes a query good







#### **Design Challenges:**

**Online setting:** only know the quality of queries tried *Exploration*: try new queries that *may* be better *Exploitation*: use queries known to be good



Short-Run Success: find sufficiently good queries quickly
O Users must remain engaged with the system



# **Dataset-Level: Fast and Lean**

Idea: learn a simple predictor

**f**<sub>w</sub> = linear function of **local entity's** features (*lexical, distributional,* and *schematic*)



Encourage exploration in *feature-space*  Pro: will converge fast (simple function)Con: not expressive; model may not work for every local entity

### **Design Challenge:**



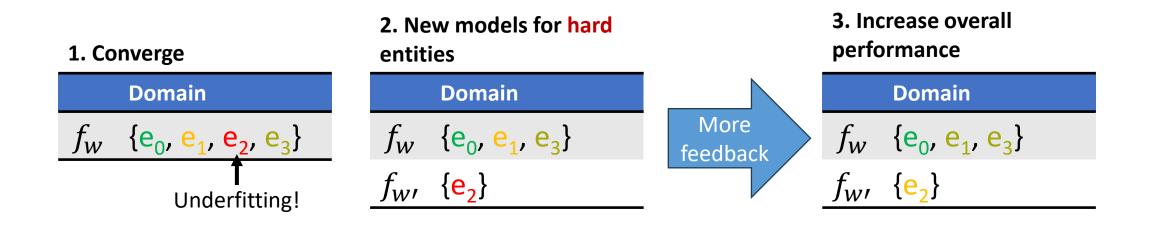
Long-Run Success: should continue to improve over time

- Methods should not waste feedback
- Avoid underfitting



Idea: learn a set of simple models to combat underfitting

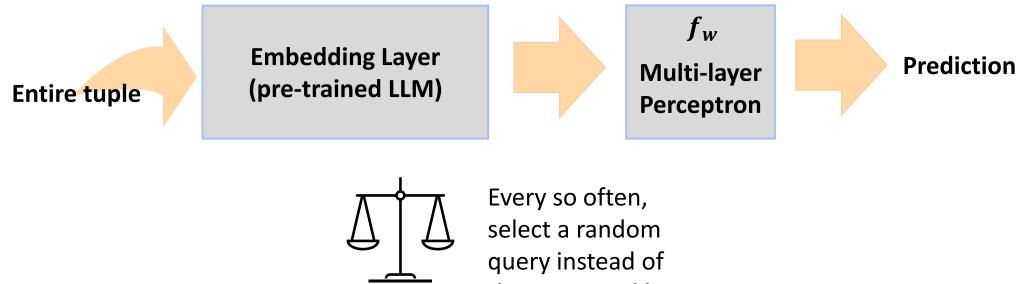
 $\mathbf{F}$  = a set of linear models starting with  $\{\mathbf{f}_w\}$  (Dataset-Level)





# LMM: Using LLM Embeddings

Idea: leverage prior knowledge of a pre-trained large language model (LLaMA)Help in short-run and long-run





## **Empirical Study Setup**

| Dataset     | Source   | Desc.   | #entities |
|-------------|----------|---|-----------|
| DrugCentral | Local    | Molecular information specific to drugs         | 3,475     |
|             | External | Regulatory information about drugs              | 4,927     |
| Drugs       | Local    | Drug reviews                                    | 13,725    |
|             | External | Wikipedia summaries of drugs                    | 46,976    |
| News        | Local    | Article titles and summaries                    | 30,000    |
|             | External | Article content                                 | 30,000    |
| WDC         | Local    | Products  | 57,109    |
|             | External | Products  | 55,247    |
| ChEBI       | Local    | Molecular information specific to drugs         | 5,483     |
|             | External | Molecules and their effects on living organisms | 189,467   |
| CORD-19     | Local    | Abstract  | 250,575   |
|             | External | Title, authors, etc,.                           | 340,826   |

Run simulations over a wide variety of datasets

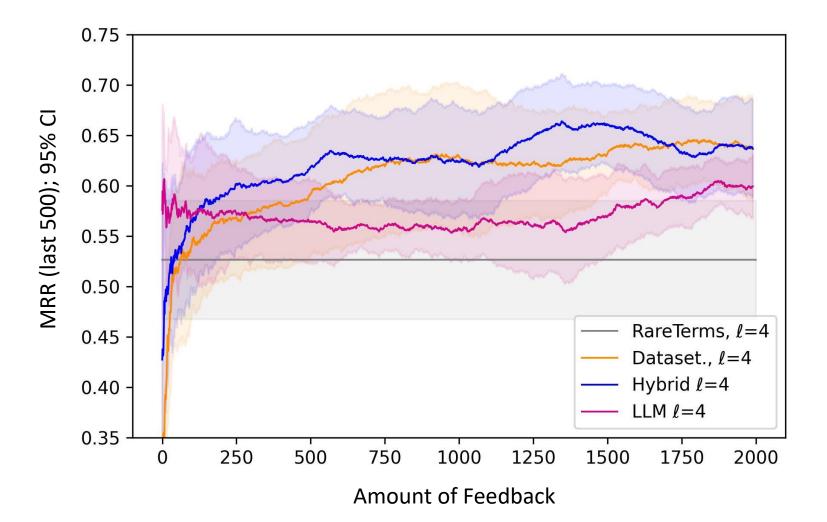
QUESTION: can our models...

- o learn quickly?
- o and keep learning?

<sup>•</sup> Ground truth = feedback

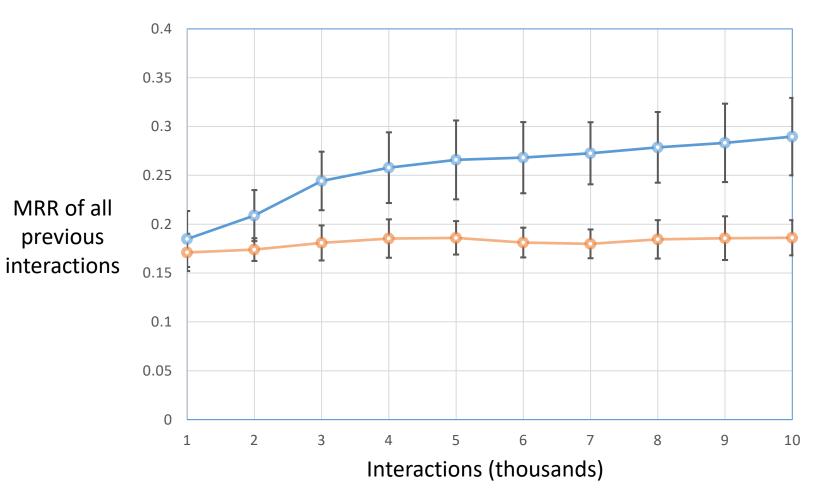


## Learning Results: ChEBI





## Hybrid vs. Dataset-Level



#### Hybrid vs. Dataset-Level

o CORD-19

• Same stream of local entities

## **Takeaways**



## **Motivation**

 Mediators require a lot of resources to build/maintain by hand





## **Approach/Problem**

- Learn the mediator online using user feedback!
- Methods to balance short-run and long-run success

### **Experiments**

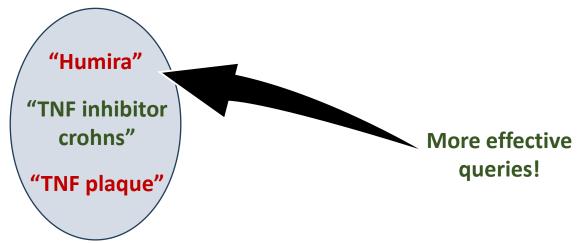
- Experiments over a variety of data sets
- They do well!



# **Other Techniques (See our Paper!)**

Term borrowing:

• Expand co-domain over time



Dynamic query length:

Adjust number of terms in query automatically
 Experiments using Longformer (another LLM)

...and more results!

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# Plug

## Generating Data Augmentation Queries Using Large Language Models @ LLMDB 2023 (Friday)



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# Thank you!

## **Please share your questions!**



