49th International Conference on Very Large Data Bases

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

**Christopher Buss**, Jasmin Mousavi, Mikhail Tokarev, Arash Termehchy, David Maier, Stefan Lee

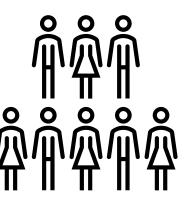


Portland State



# **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 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 arthritis
	FDA-Appr	roved Drugs	
	FDA-Appr brand name		uses
			uses
	brand name	class	uses rheumatoid arthritis

#### Generic Drugs

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

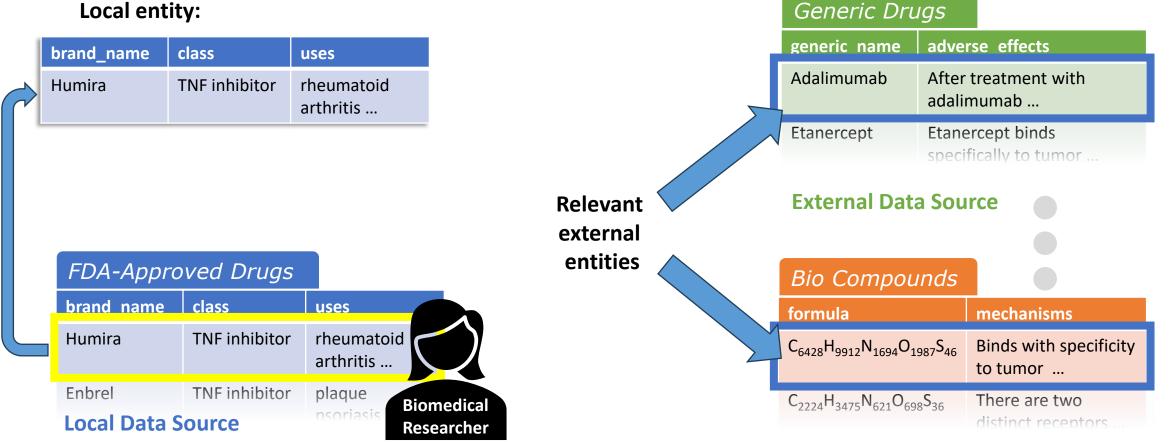
External Data Source			
Bio Compounds			
formula	mechanisms		
$C_{6428}H_{9912}N_{1694}O_{1987}S_{46}$	Binds with specificity to tumor		
$C_{2224}H_{3475}N_{621}O_{698}S_{36}$	There are two distinct receptors		

#### **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 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 arthritis	$\bigcirc$	
Enbrel	TNF inhibito	or plaque	Biomedical	

Researcher

**Local Data Source** 

#### Generic Drugs

generic name	adverse effects	
Adalimumab	Adalimumab After treatment with adalimumab	
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External Data Source		
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 $C_{2224}H_{3475}N_{621}O_{698}S_{36}$ 

There are two

#### **External Data Source**



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

## Challenges:

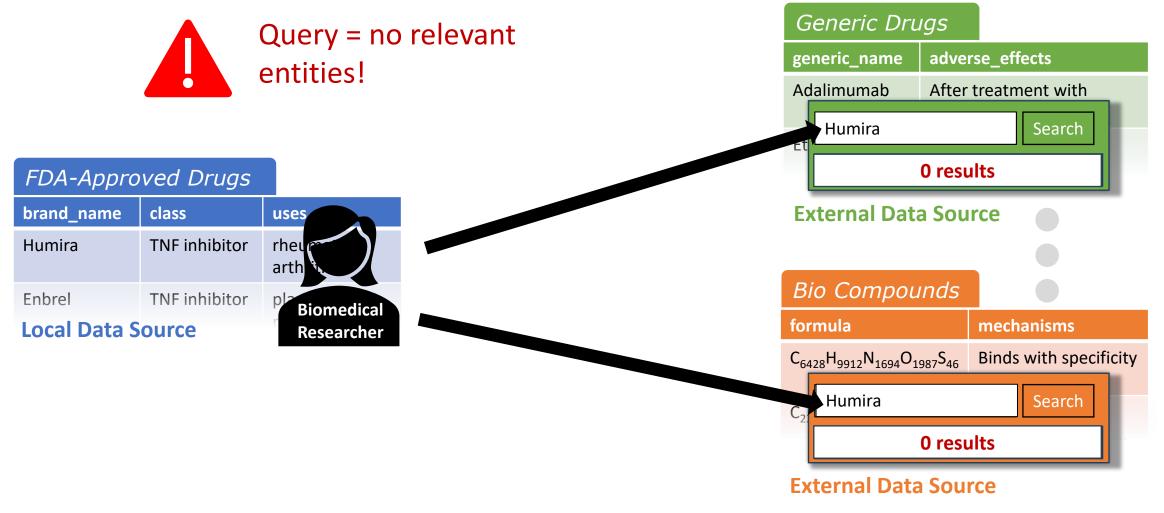
- Many external data sources
- Data heterogeneity: different representations
  - Humira = Adalimumab
    - $= C_{6428} H_{9912} N_{1694} O_{1987} S_{46} = ???$

FDA-Appro	ved Drugs	
brand_name	class	uses
Humira	TNF inhibitor	rheumatoid arthritis
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Local Data S	ource	Researcher

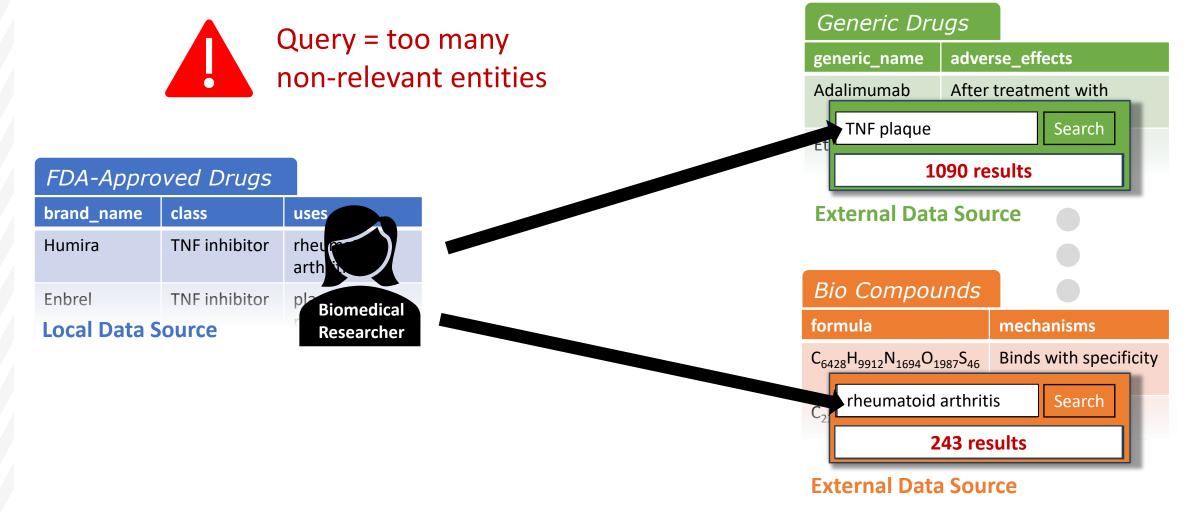
Generic Dru	ıgs	
generic_name	advers	e_effects
Adalimumab	After t	reatment with
Et		Search
External Data	a Sour	ce
Віо Сотрои	Inds	
formula		mechanisms
C <sub>6428</sub> H <sub>9912</sub> N <sub>1694</sub> O <sub>1</sub>	<sub>987</sub> S <sub>46</sub> I	Binds with specificity
C <sub>21</sub>		Search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

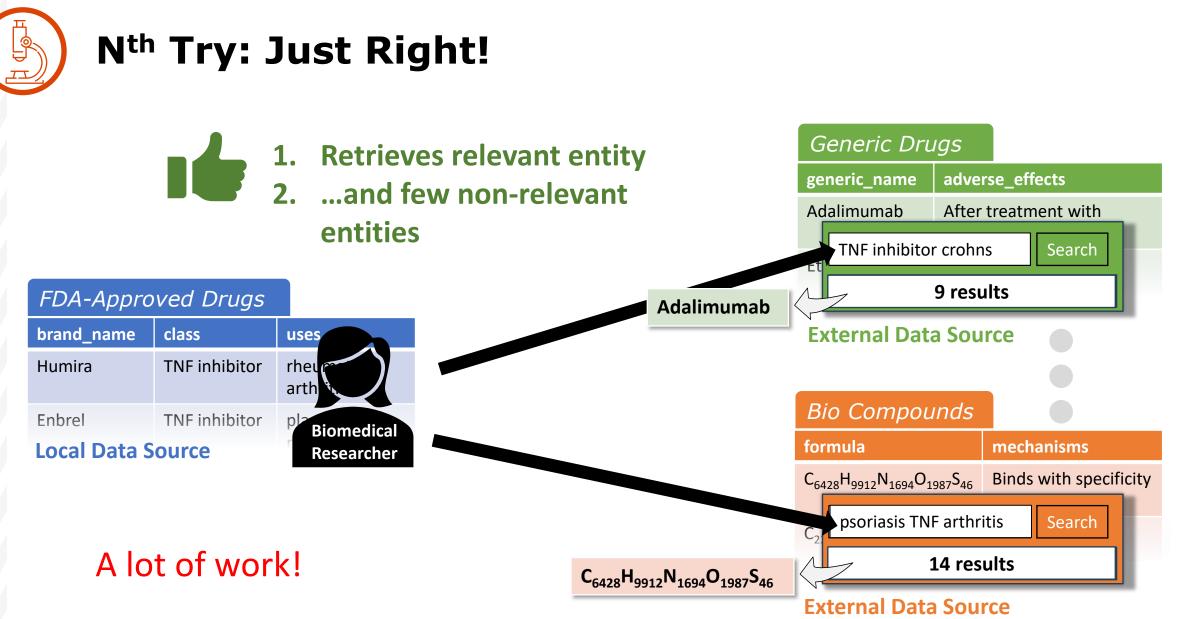


## 1<sup>st</sup> Try: Query = Too Specific to Local Source











# **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

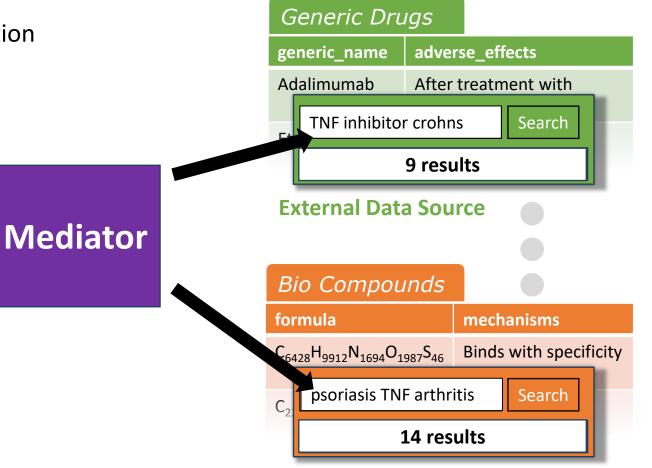
uses

rheu arth

pla

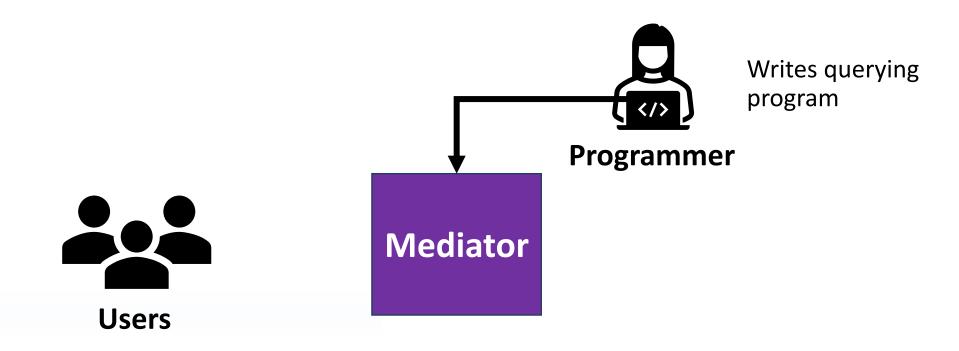
User

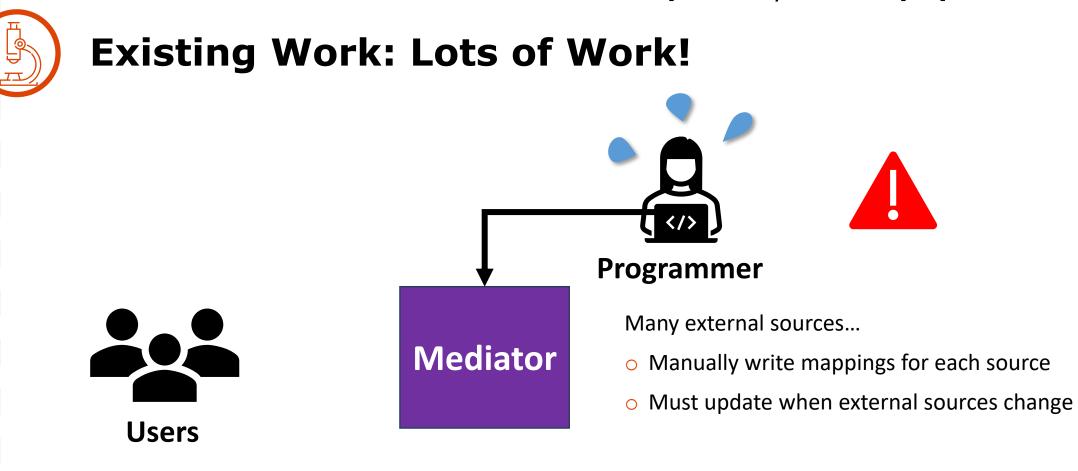
2. Mediator retrieves relevant information from external sources

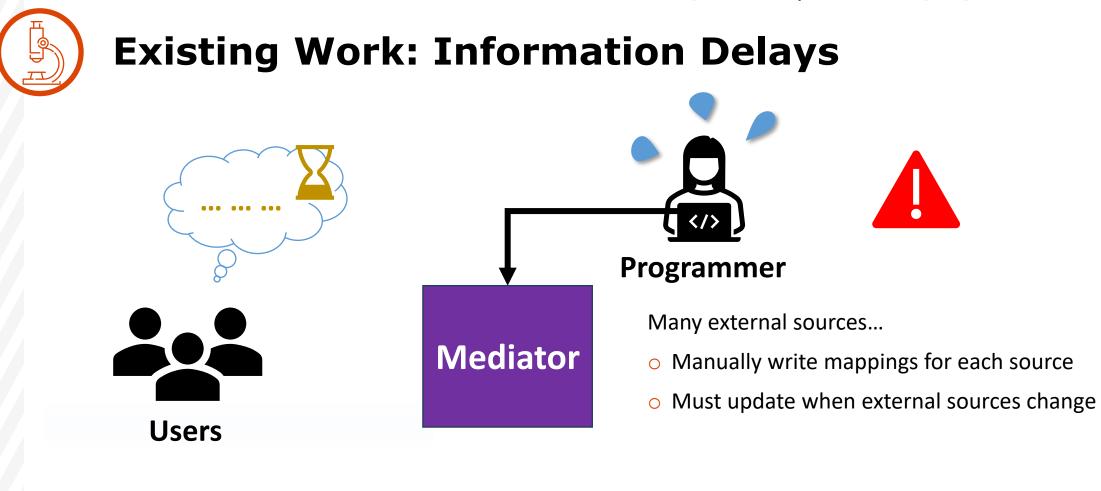


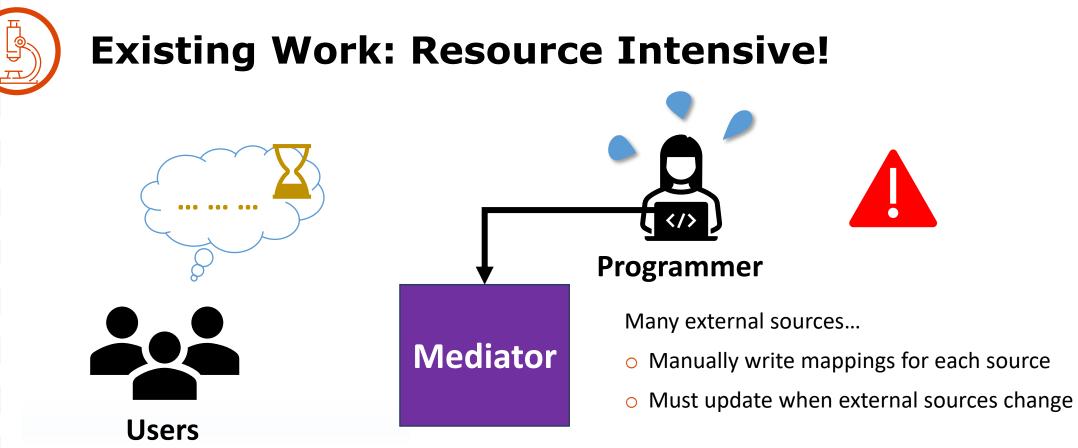


## **Existing Work: Mediator Written By Hand**







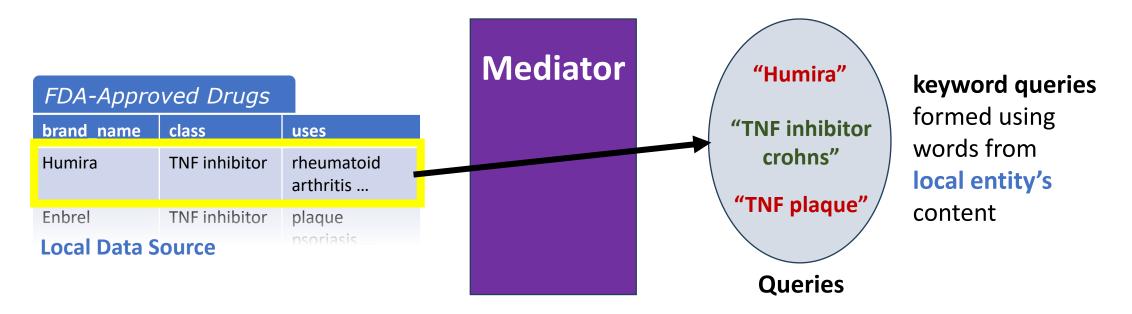


For example: the NIH funds a consortium of such systems (~14 systems)

- 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



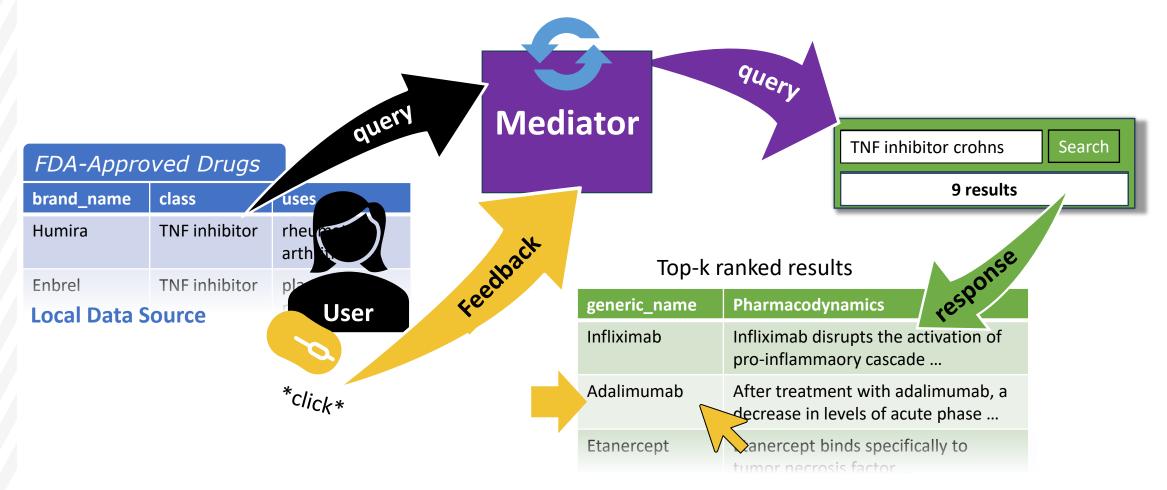
- Hire domain experts to label data
- External source updates  $\rightarrow$  must repeat!
- Still delays...

### **Online Learning:**

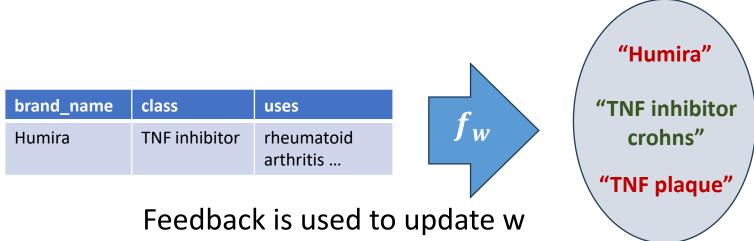
o Train mediator while users query it



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:**



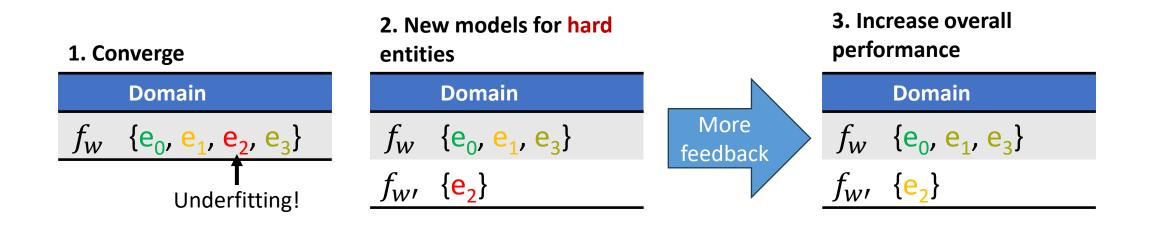
Long-Run Success: should continue to improve over time

- Methods should not waste feedback
- Avoid underfitting



Idea: learn a set of simple models to combat underfitting

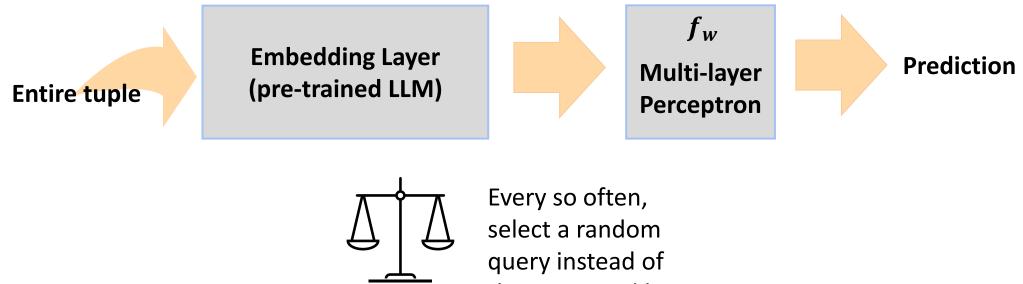
 $\mathbf{F}$  = a set of linear models starting with  $\{\mathbf{f}_w\}$  (Dataset-Level)





# 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

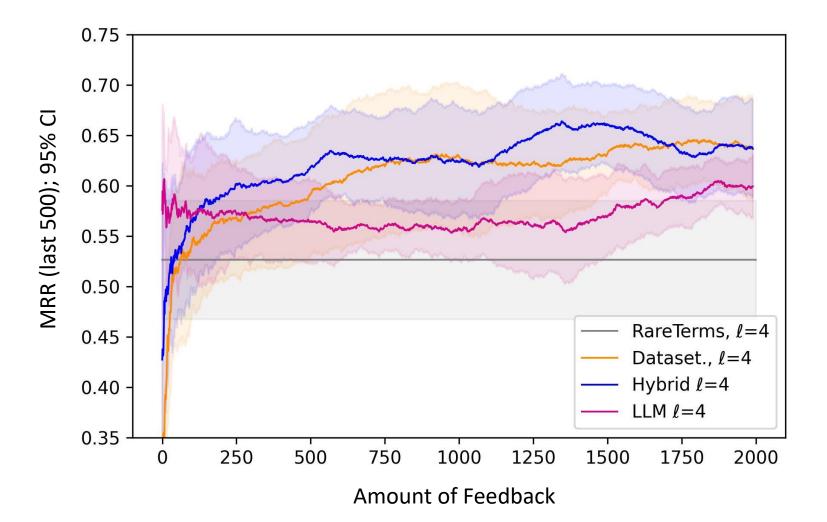
QUESTION: can our models...

- o learn quickly?
- o and keep learning?

<sup>•</sup> Ground truth = feedback

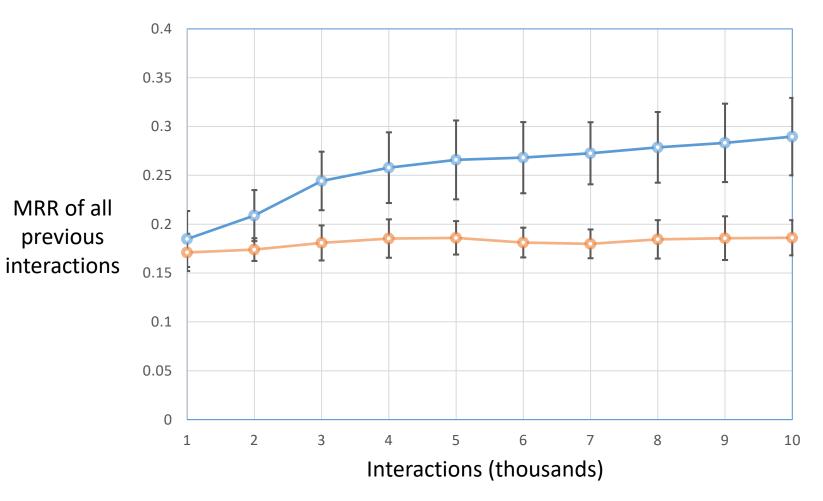


## Learning Results: ChEBI





## 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**

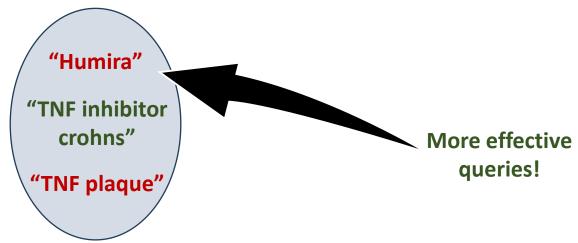
- Experiments over a variety of data sets
- They do well!



# **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!**



