

# Actualizing research into practical tools: A case study of GenRA, a new read-across tool



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# Outline

- Definitions
- Introduction to the Generalized Read-Across (GenRA) approach
- Current progress on web tool development
- Future work on web tool development
- Conclusions

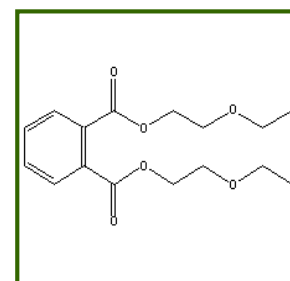
# Definitions: Read-across

- Read-across describes the method of filling a data gap whereby a chemical with existing data values is used to make a prediction for a 'similar' chemical.
- A target chemical is a chemical which has a data gap that needs to be filled i.e. the subject of the read-across.
- A source analogue is a chemical that has been identified as an appropriate chemical for use in a read-across based on similarity to the target chemical and existence of relevant data.

	Source chemical	Target chemical
Property		

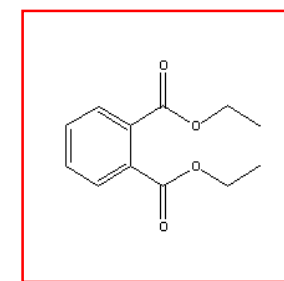
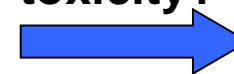
● Reliable data

○ Missing data



**Known to be  
harmful**

**Acute  
toxicity?**



**Predicted to be  
harmful**

# GenRA - Introduction

- GenRA (Generalized Read-Across) is a “local validity” approach predicting toxicity as a similarity-weighted activity of source analogues based on chemistry and/or bioactivity descriptors. (Shah et al, 2016)
- Generalized version of Chemical-Biological Read-Across (CBRA) developed by Low et al (2013)
- **Goal:** to establish an objective performance baseline for read-across and quantify the uncertainty in the predictions made.
- Tested and compared chemical, bioactivity (hitcalls from ToxCast) descriptors as well as a hybrid of the two.
- Underlying data used was taken from ToxRefDB, a collection of repeated dose toxicity study types e.g. chronic, multigeneration, developmental, subchronic etc
- Toxicity effects within those study types were recorded as binary outcomes

# Algorithm

1. Select chemical/biological fingerprint to characterize target and source analogues. The Jaccard similarity metric is used to quantify the pairwise similarity
2. Choose the number of source analogues, k (where k = the number of nearest neighbors)
3. Filter on the basis of *in vivo* toxicity data for source analogues
4. Read-across prediction for target chemical is the outcome of a similarity-weighted activity calculation from the nearest neighbors
5. Perform Receiver Operating Characteristic (ROC) to identify the threshold where the best performance is produced for a given s and k.

Jaccard similarity:

$$s_{ij} = \frac{\sum_l (x_{il} \wedge x_{jl})}{\sum_l (x_{il} \vee x_{jl})}$$

$$y_i = \frac{\sum_j^k s_{ij} x_j}{\sum_j^k s_{ij}}$$

# Results of GenRA

Tabulation of high-performing scores (p<0.01)

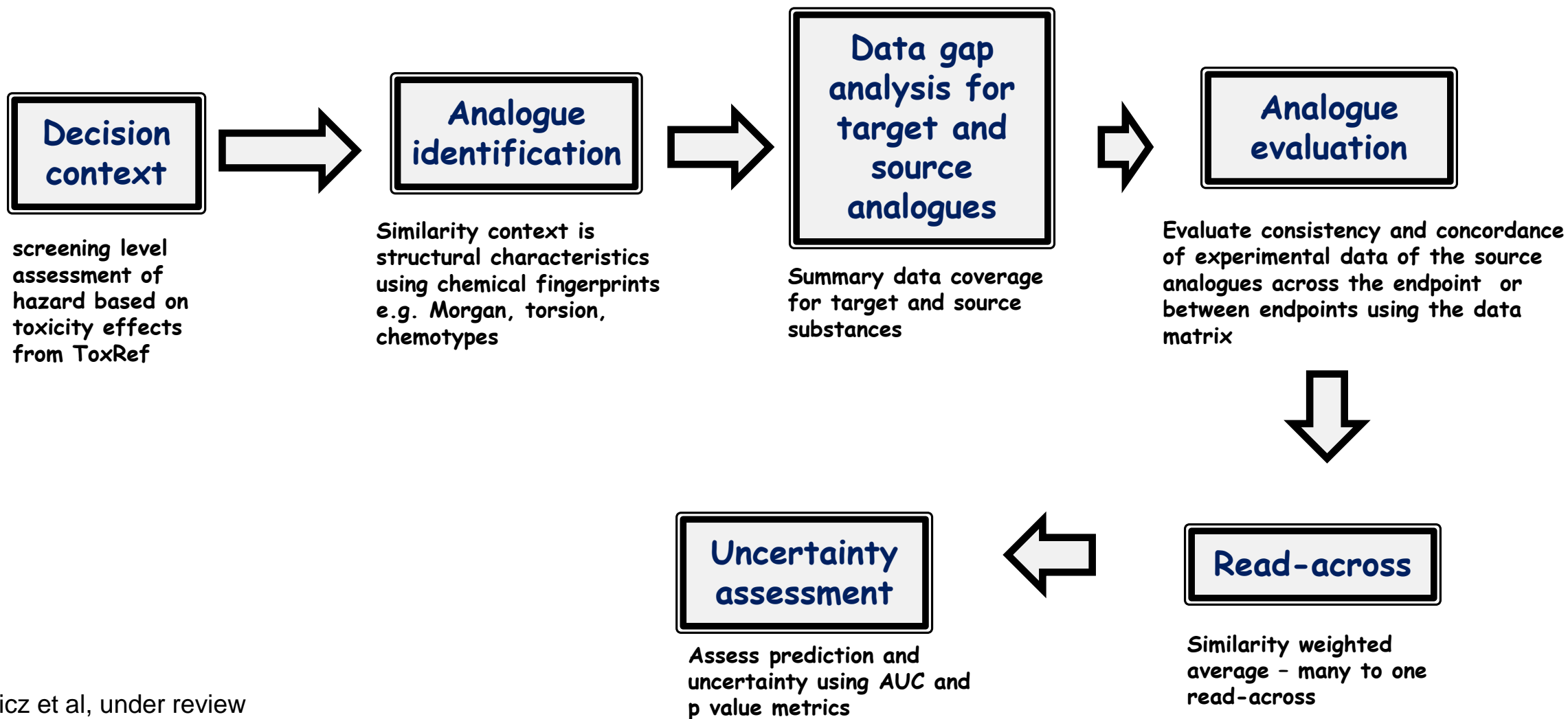
Study	Cases*	chm	bio	bc	chm > bio  bc	bio > chm bc	bc > chm bio	Any
chr	1115	97 (9%)	133 (12%)	67 (6%)	75 (7%)	107 (10%)	21 (2%)	203 (18%)
dev	712	63 (9%)	138 (19%)	50 (7%)	43 (6%)	117 (16%)	7 (1%)	167 (23%)
dnt	84	16 (19%)	26 (31%)	15 (18%)	5 (6%)	20 (24%)	2 (2%)	27 (32%)
mgr	668	78 (12%)	126 (19%)	56 (8%)	46 (7%)	106 (16%)	4 (1%)	156 (23%)
rep	34	0 (0%)	0 (0%)	0 (0%)	0 (0%)	0 (0%)	0 (0%)	0 (0%)
sac	128	2 (2%)	7 (5%)	6 (5%)	1 (1%)	6 (5%)	2 (2%)	9 (7%)
sub	922	107 (12%)	140 (15%)	78 (8%)	64 (7%)	122 (13%)	21 (2%)	212 (23%)
All	3663	363 (10%)	570 (16%)	272 (7%)	234 (6%)	478 (13%)	57 (2%)	774 (21%)

\*cluster-study-effect combination

- <https://cc>

e

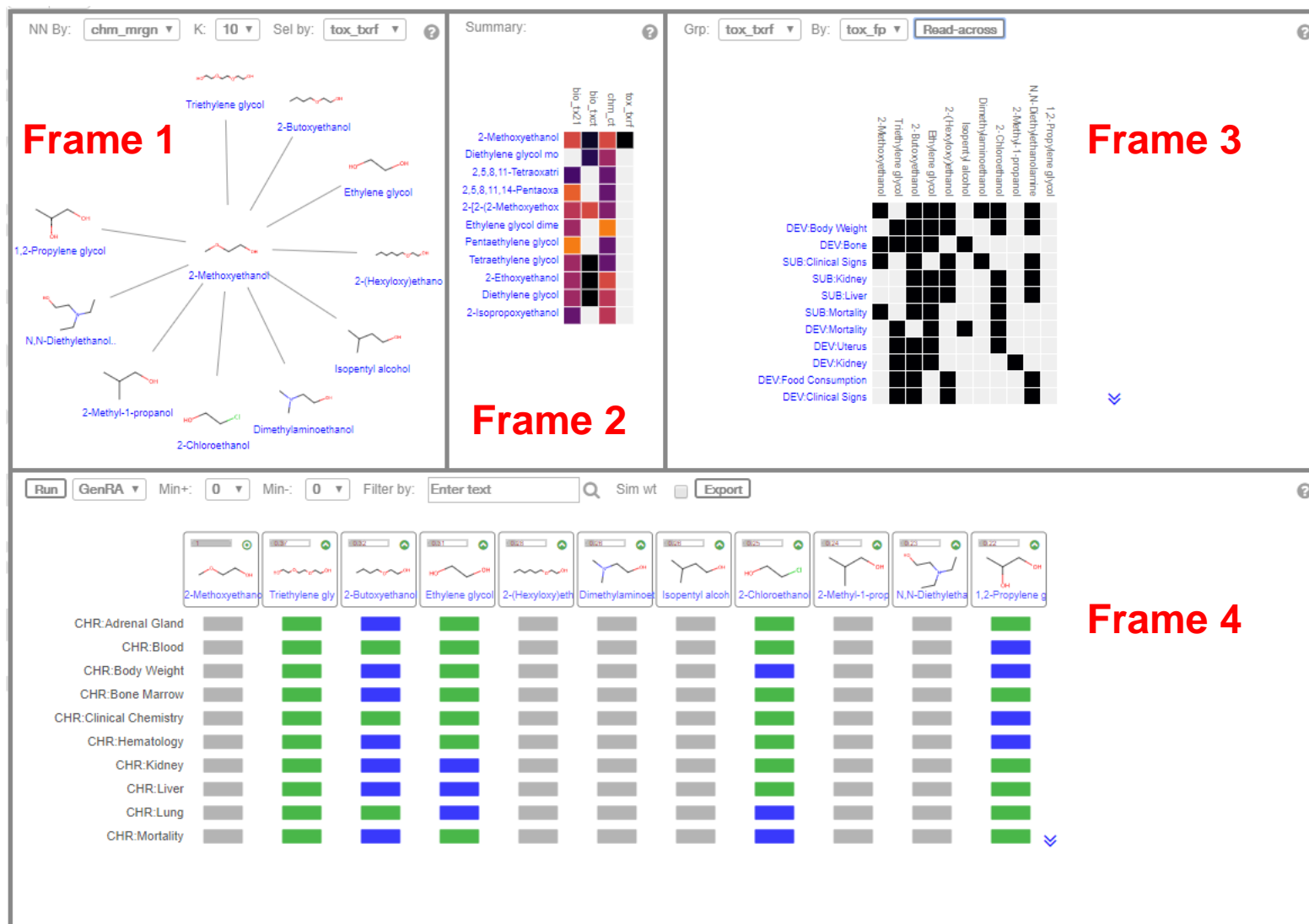
# Category Workflow and its alignment to GenRA



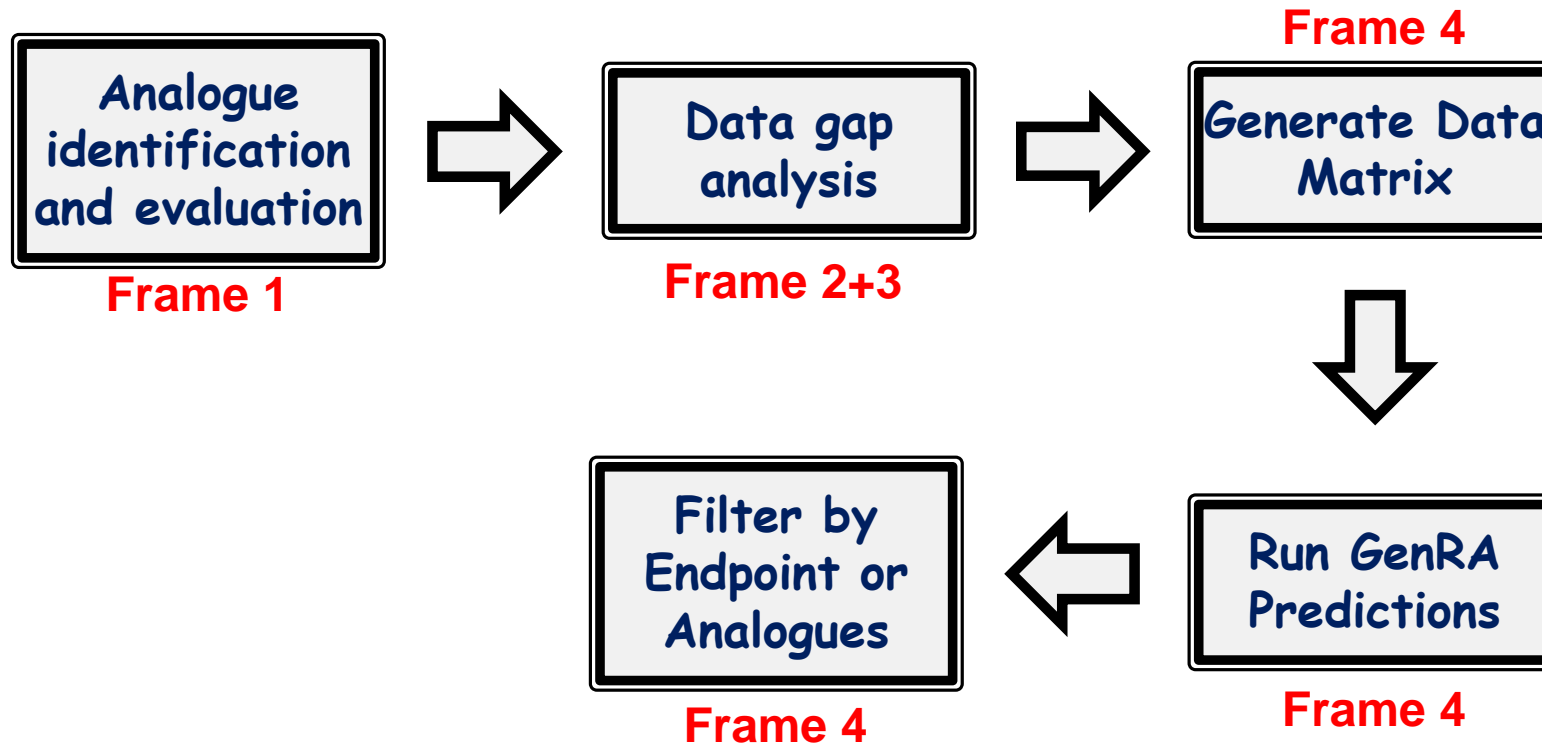
Patlewicz et al, under review  
Helman et al, in prep



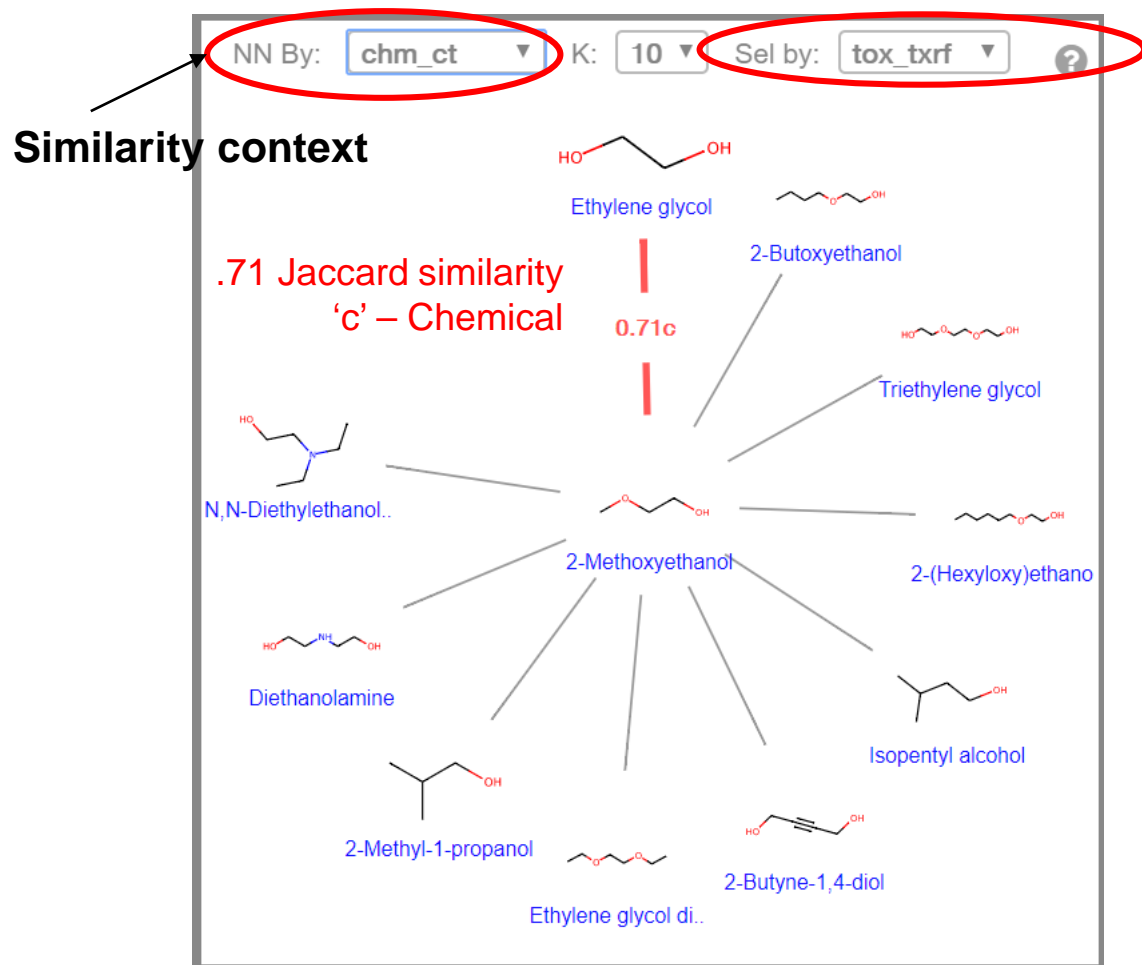
# GenRA: Web Tool Development



# Web Tool Workflow

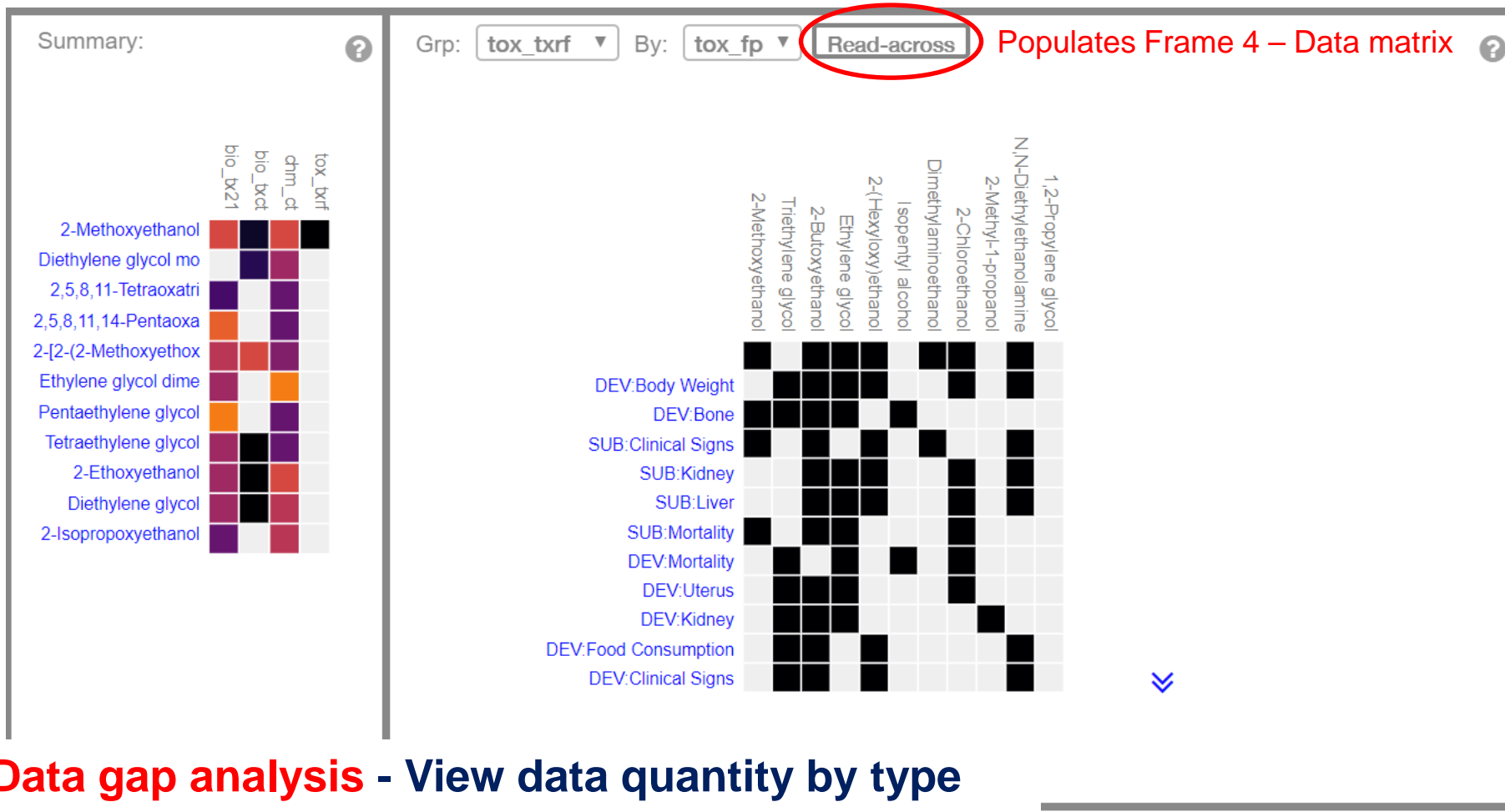


# Analogue Identification and Evaluation



- 'NN By' option determines fingerprint of choice i.e. search for source analogues on the basis of different chemical fingerprints.
- 'K' option determines the number of source analogues to search for
- 'Sel by' option subsets chemical space to search for source analogues that have certain types of data, by default *in vivo* data.

# Data Gap Analysis



- ‘Grp’ option controls the data that is being shown in the heatmap.
- ‘By’ option allows you to pivot the heatmap to look at different views of the same data.

# Generate Data Matrix



- 'Min+' and 'Min-' options control minimum number of pos and neg values in displayed endpoints.
- 'Sim wt' option changes width of cells to scale of the similarity of the neighbor.
- 'Export' will export data matrix to CSV

# Run GenRA Predictions



# Filter by Endpoint or Analogues

Run GenRA Min+: 0 Min-: 0 Filter by: liver Sim wt ☐ Export

	2-Methoxyethanol	Triethylene gly	2-Butoxyethanol	Ethylene glycol	2-(Hexyloxy)eth	Dimethylaminoet	Isopentyl alcoh	2-Chloroethanol	2-Methyl-1-prop	N,N-Diethyletha	1,2-Propylene g
CHR:Liver											
DEV:Liver											
SUB:Liver											
MGR:Liver											
DNT:Liver											
SAC:Liver											

Provide user with flexibility to  
deselect analogues or focus in on  
specific toxicity effects or study  
types of interest

# Future Work: Physicochemical Similarity Context

- Searching on the basis of physicochemical information in addition to structure can improve GenRA performance (Helman et al, 2018)
- Physicochemical information will be incorporated into the analogue identification and evaluation steps of the workflow.
- The ‘Lipinski Rule of 5’ properties (LogP, MW, #HB donors, #HB acceptors) were used in the analysis undertaken.



# Future Work: Point of Departure Prediction

- Algorithm readily extendible to continuous data.

$$y_i = \frac{\sum_j^k s_{ij} x_j}{\sum_j^k s_{ij}}$$

- Preliminary results appear promising for predicting acute oral rat LD50 data.

# Conclusions

- GenRA is an objective algorithmic approach for generating read-across predictions.
- A web tool has been developed to front-end the GenRA workflow
- The web tool will be made available on NCCT's CompTox dashboard soon.

# Acknowledgements

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