



# **GenRA Virtual Training**

## **Breakout Group Worksheet**

This worksheet was developed for the Breakout Group session of the GenRA Virtual Training, hosted by the U.S. Environmental Protection Agency's Center for Computational Toxicology and Exposure on May 23, 2023.

For more information about the GenRA tool:

- Visit the GenRA at <u>https://comptox.epa.gov/genra/</u>
- Contact GenRA support at <u>genra.support@epa.gov</u>
- Read more in
  - o Publication 2016: <u>https://doi.org/10.1016/j.yrtph.2016.05.008</u>
  - o Publication 2022: <u>https://doi.org/10.1016/j.comtox.2022.100258</u>
  - GenRA Manual: <u>https://www.epa.gov/chemical-research/generalized-read-across-genra-manual</u>

#### Goals

- **Hazard identification:** Use GenRA to evaluate the potential hazards of a substance found in a contaminated site. What can be determined based on the read-across workflow as related to a screening level human health hazard assessment?
- **Assessing toxicity:** Use of GenRA to infer the potential toxicity effect for a target chemical based on study-type and effect information. Based on the read-across predictions and their associated uncertainties, what might be the next steps?
  - E.g., Consumer product safety: Toxicity assessment of a given chemical used as an ingredient in a product. Based on the read-across predictions, what inferences could be made regarding product safety and what additional information might be needed?
- Gaining an understanding of:
  - How a read-across should be considered relative to the decision context.
  - What available Fingerprint types are in GenRA and what they represent.
  - How to navigate the Neighborhood Explorer to inform selection of analogues based on different fingerprint types. Additionally,
    - How to use the Physchem Data viewer, and
    - How to make binary toxicity effect predictions for a data-poor target chemical of interest.

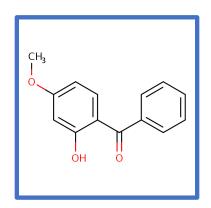




## **Target Chemical**

#### 1. 2-HYDROXY-4-METHOXYBENZOPHENONE (HMB)

Oxybenzone or benzophenone-3 or BP-3 (trade names Milestab 9, Eusolex 4360, Escalol 567, KAHSCREEN BZ-3) is an organic compound. It is a pale-yellow solid that is readily soluble in most organic solvents. Oxybenzone belongs to the class of aromatic ketones known as benzophenones. It is widely used in plastics, toys, furniture finishes, and other products to limit UV degradation.



IUPAC Name: (2-Hydroxy-4-methoxyphenyl)(phenyl)methanone CASRN: 131-57-7 DSSTox substance identifier (DTXSID): DTXSID3022405 Molecular Formula: C<sub>14</sub>H<sub>12</sub>O<sub>3</sub> SMILES: COC1=CC=C(C(=O)C2=CC=CC=C2)C(O)=C1 InChIKey: DXGLGDHPHMLXJC-UHFFFAOYSA-N

**Scenario:** Trace amounts of 2-hydroxy-4-methoxybenzophenone (HMB) were found in a drinking water source. The duration of exposure is unknown. After determining the concentration in drinking water, it was found to exceed the established Threshold of Toxicological Concern (TTC) values which necessitated a read-across approach. Based on the information provided you would like to:

- Identify data gaps, and
- Evaluate potential toxicity effects for this contaminant.





#### Directions

- Using the chemical information for HMB above, type the chemical name (or synonym), its Chemical Abstracts Service Registry (CASRN), and DSSTox Substance Identifier (DTXSID) below.
- 2. Access GenRA directly from <a href="https://comptox.epa.gov/genra/">https://comptox.epa.gov/genra/</a>
- 3. Either use the Ketcher drawing palette or search box to search for HMB using the SMILES or enter DTXSID in the search box.
- 4. Identify potential source analogues for HMB using the default settings in Panel 1 which depicts the radial plot.
  - a. How many source analogues are presented in the radial plot?
  - b. Based on the Jaccard similarity metric, what is the most similar source analogue to the target chemical?
  - c. Based on the Jaccard similarity metric, what is the least similar source analogue to the target chemical?
  - d. Based on the Jaccard similarity metric, what is the most similar source analogue to the target chemical if the panel is updated using the ToxCast data filter?
  - e. After this, change the Filter back to "ToxRef data", which is the default.
  - f. Evaluate the Physchem properties of each source analogue (based on the OPERA model estimates), access this information by clicking "Physchem Data".
    - i. Consider the likely physical state of the source analogues using the predicted melting and boiling points. How do they compare to the likely state of HMB? [Note: Plot window is resizable!]





- ii. What is the predicted Log Kow of HMB?
- iii. How many source analogues have Log Kow values above 2 and below 5?
- g. Close the "Physchem Data" window. Now, explore the analogues by using
  "Neighborhood Exploration". Click on the target chemical (HMB) to zoom in.
  [Reminder: The target chemical is indicated by a red circle.]
  - i. Identify the top 3 source analogues based on Morgan Fingerprints.
  - ii. Identify the 3 source analogues based on ToxCast Fingerprints.
- Identify the analogues on the basis of the hybrid option 50:50 Morgan to ToxCast.
- 5. Is the target chemical data poor? [Hint: evaluate the Panel 2 on Step 3 of GenRA tool!]
  - a. What data exists for the source analogs identified?
  - b. Do the data for the source analogues address the data gaps of interest for the target chemical? (Or, how feasible might a read-across be based on the quantity of data for the source analogues?)





- 6. Let's move on to Panel 3. Which types of toxicity effect information are available for the target chemical and across the analogues?
  - Evaluate the binary predictions of toxicity data arising from ToxRefDB (default).
     What data gaps exist for the source analogues relative to the target substance, and which effects might be reasonably predicted by those analogues?
  - b. Evaluate Chronic bone marrow assay end point for Benzophenone. Are there any experimental data?
  - c. Are there any experimental data for source analogues for Chronic bone marrow assay end point?
- Would Panels 2 and 3 provide any information on their potency (e.g., in mg/kg-day etc.) or hazard profile? [Note: Click "Generate Data Matrix".]
  - a. Evaluate the suitability of the source analogues by looking at the data availability of the source analogues. Are there enough data for chronic and sub-chronic assay endpoints?
    - i. If not, what are the next steps?





- b. Find the dose for the analogue (Benzophenone) at which a toxicity effect for chronic body weight (CHR:liver) was observed. Note below if there are any observations exist for HMB for CHR:liver end point.
- c. Consider the Physchem properties of each analogue and identify those with estimated Log Kow values within 1.5 of the estimate for the target chemical Log Kow (e.g., +1.5 and -1.5 of 3.785).
  - i. De-select any analogues that are below 2 and above 5. How many analogues are now selected for Read-Across?
  - ii. Once the desired source analogues are selected, click the "Run Read-Across" button to re-derive the GenRA predictions using the GenraPred engine. What is the prediction value?
- d. Export and save the results as an Excel file.
  - i. Does the file export the target predictions as well as all 10 analogues' data?
- e. Is there any subsequent analysis after this supported by GenRA?
- f. Filter column D, "pred\_class", to show only positive effects.
- g. Sort the pval from smallest to largest value.





### Reflection

1. In what case example from your work environment would GenRA be useful?

2. What have you learned about the process and workflow used to find information in GenRA?

3. What challenges did you encounter, and how did you solve them?

**WORKSHEET ANSWERS** 





#### Directions

 Using the chemical information for HMB above, type the chemical name (or synonym), its Chemical Abstracts Service Registry (CASRN) and DSSTox Substance Identifier (DTXSID) in the space below.

2-HYDROXY-4-METHOXYBENZOPHENONE, 131-57-7, DTXSID3022405

- 2. Access GenRA directly from <a href="https://comptox.epa.gov/genra/">https://comptox.epa.gov/genra/</a>
- Either use the Ketcher drawing palette or search box to search for HMB using the SMILES or enter DTXSID in the search box.

#### DTXSID3022405

4. Identify potential source analogues for HMB using the default settings in panel 1 which depicts the radial plot.



- a. How many source analogues are presented in the radial plot?
   (10, default)
- b. Based on the Jaccard similarity metric, what is the most similar source analogue to the target chemical?





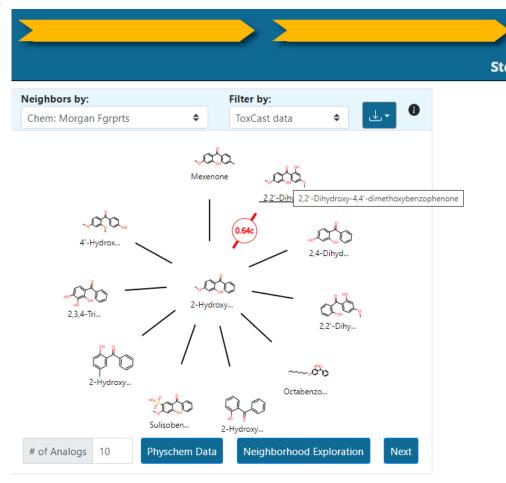
#### (Benzophenone, 0.38)

c. Based on the Jaccard similarity metric, what is the least similar source analogue to the target chemical?

(Methyl paraben, 0.23)

d. Based on the Jaccard similarity metric, what is the most similar source analogue to the target chemical if the panel is updated using the ToxCast data filter?





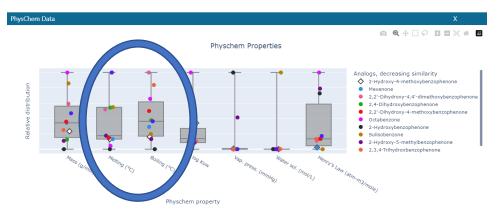
e. After this, change the Filter back to the "ToxRef data" default.







- f. Evaluate the Physchem properties of each source analogue (based on the OPERA model estimates), access this information by clicking "Physchem Data".
  - i. Consider the likely physical state of the source analogues using the predicted melting and boiling points. [Note: Plot window is resizable!]

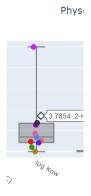


HMB is known to be a solid – based on the description. This is consistent with its predicted MP (a value exceeding 25 deg C is indicative of a solid). Note that the example source analogues Benzyl acetate and Methyl salicylate have much lower MP values than our target substance. See <u>https://www.epa.gov/sites/default/files/2015-05/documents/05-</u> <u>iad\_discretes\_june2013.pdf</u>

ii. What is the predicted Log Kow of HMB?(3.7854)



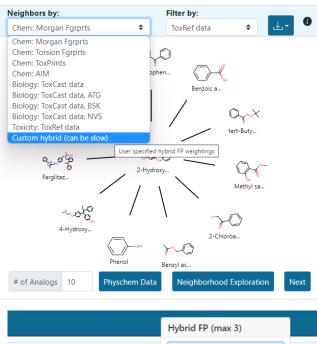


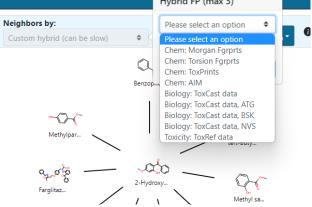


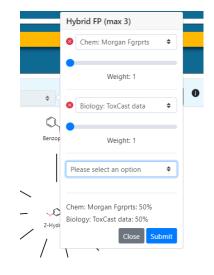
- iii. How many source analogues have Log Kow values above 2 and below 5?
   (Four (4): benzophenone, t-butyl perbenzoate, methyl salicylate, 2chloroacetophenone)
- g. Close the Physchem Data window. Now, explore the analogues by using
  "Neighborhood Exploration". Click on the target chemical (HMB) to zoom in.
  [Reminder: The target chemical is indicated by a red circle.]
  - i. Identify the top 3 source analogues based on Morgan Fingerprints.
     (Benzophenone, benzoic acid, and t-butyl perbenzoate)
  - ii. Identify the 3 source analogues based on ToxCast Fingerprints.(Phenothrin, tris(methylphenyl)phosphate, 4,4'-dichlorophenylsulfone)
- Identify the analogs on the basis of the hybrid option 50:50 Morgan to ToxCast.
   Close the neighborhood explorer and go back to panel one and select
   custom hybrid





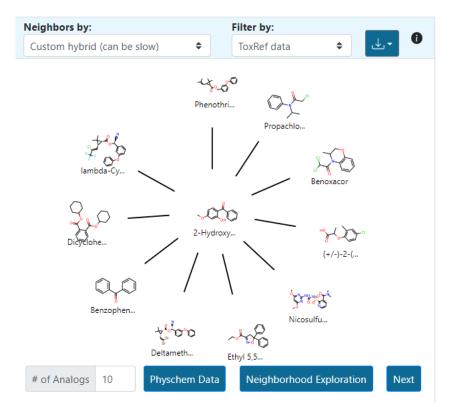












### Panel one will updated to reflect the below screen

Is the target chemical data poor? [Hint: evaluate the Panel 2 on Step 3 of GenRA tool!]
 The answer is Yes. Let's evaluate panels 2 and 3

PANEL 2





0

#### **Summary Data Gap Analysis**

Chemica	bio_txct	chm_ct	chm_httr	chm_mrgr	tox_txrf
2-Hydroxy-4-methoxybenzophenone	296	13	27	39	68
Phenothrin	270	15	44	63	54
Propachlor	266	12	21	33	180
Benoxacor	271	15	32	46	40
(+/-)-2-(4-Chloro-2-methylphenoxy)propio	261	11	21	37	84
Nicosulfuron	257	16	50	63	75
Ethyl 5,5-diphenyl-2-isoxazoline-3-carboxy	266	6	39	47	30
Deltamethrin	264	19	49	69	169
Benzophenone	350	9	18	19	198
Dicyclohexyl phthalate	279	6	35	34	55
lambda-Cyhalothrin	268	24	53	74	89
Rows: 11 Total Roy	ws: <b>11</b>				

Each column in Panel 2 corresponds to the fingerprint options from Panel 1. Each fingerprint is a binary bit vector reflecting the presence/absence of features (e.g., ToxPrints comprise 729 features, whereas Morgan fingerprints comprise 2048 features). The color density is scaled by fingerprint type from light to dark and reflects a measure of 'data availability'. The number of data records is reflected in each cell.

a. What data exists for the source analogs identified?





- (The data availability is segmented by data type where bio\_txct represents bioactivity hitcall data from ToxCast, e.g., chm\_ct represents chemistry ToxPrint information and tox\_txrf represents *in vivo* toxicity effect information from ToxRefDB v2.
- The color density represents a measure of 'data availability' for the target – from light to dark. The number of data records is reflected in the box itself.
- In the case of target HMB, there are moderate amount (tested in 296 assays) available Toxcast bioactivity hitcall data outcomes (bios\_txct), but majority of the analogues have less data in comparison to HMB except for benzophenone (tested in 350 assays). There is a moderate amount of in vivo toxicity data available for HMB in ToxRefDB (tox\_txrf).)
- However, analogues such as Benzophenone and Propachlor have a great deal more toxicity data which facilitates a read-across.
- b. Do the data for the source analogues address the data gaps of interest for the target chemical? (Or, how feasible might a read-across be based on the quantity of data for the source analogues?)

This depends on the use case = it would be a question for discussion as say for a regulatory submission – it might be that you need to address a specific endpoint – so do the analogues identified possess experimental data for the endpoint you need... as well as whether you need to readacross for more than one endpoint to generate more of a predicted toxicity profile.





6. Let's move on to Panel 3. Which types of toxicity effect information are available for the target chemical and across the analogues? [Hint: Evaluate the Panel 3 on Step 3 of GenRA tool.]

Pagination Generate Data Matrix 0 Group: ToxRef 🗢 By: Tox Fingerprint ۵ Ethyl 5,5-dip. Dicyclohexyl ambda-+/-)-2-(4-Ch licosulfuron Deltamethrin Hydroxy enoxacol enzophenone tenothrin opachlor endpoint -Cyha. 7 CHR:adrenal .. CHR:alanine a... CHR:albumin CHR:alkaline ... CHR:appeara... CHR:aspartate... CHR:body wei... CHR:bone ounderer and Total Rows: 248 Rows: 248 I< < Page 1 of 28 > >I 1 to 9 of 248

#### PANEL 3

(This view provides a representation of the data quantity across different toxicity effects for the source analogues selected. This informs the data gap analysis step of the workflow. Panels 3 and 4 are duplicative in terms of the data presented... we can change the view in Panel 4 based on whether we choose the potency type or start Panel 1 with the use case of wanting to predict ToxCast assay hit calls and therefore filter for this accordingly.)

a. Evaluate the binary predictions of toxicity data arising from ToxRefDB (default).
 (What data gaps exist for the source analogues relative to the target substance, and which effects might be reasonably predicted by those analogues?)

Use Panel 3 to help profile the data availability across study-toxicity effect combination exists noting that the representation can be changed





if potency predictions are needed. Panel 4 differentiates the view presented in Panel 3.

b. Evaluate Chronic bone marrow assay end point for Benzophenone. Are there any experimental data?

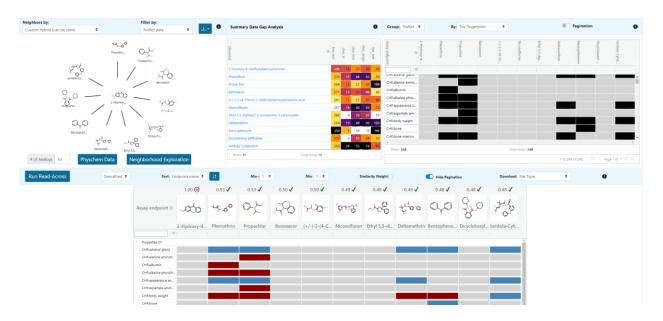
(No, no data)

c. Are there any experimental data for source analogues for Chronic bone marrow assay end point?

(Yes)

7. Would Panels 2 and 3 provide any information on their potency (e.g., in mg/kg-day etc.) or hazard profile? [Note: Click "Generate Data Matrix".]
(No. To evaluate this, click, the 'Generate Data Matrix' button and to move to the next

step of the workflow 'Run GenRA Prediction'. See below for the next view)



- Evaluate the suitability of the source analogues by looking at the data availability of the source analogues. Are there enough data for chronic and sub-chronic assay endpoints?
   (YES/depends on the end point of interest)
  - i. If not, what are the next steps?





(If little data are available for the source analogs or they fail to address the data gaps of interest for the target chemical then this might lead the user to change the number of source analogues or select a different similarity context.)

b. Find the dose for the analogue (Benzophenone) at which a toxicity effect for chronic body weight (CHR:liver) was observed. Note if there are any existing observations for HMB for the CHR:liver end point.

(Sort by 'positive obs', then toggle between ascending and descending obs.: 15 mg/kg/day) (filter by endpoint name e.g., Liver))



(Yes, and take a note of POS, ACT 0.613 generated for HMB after running Read-Across using the GenraPred engine which shows that the prediction is consistent with the empirical data available. The degree of cell opacity denotes the confidence associated with any prediction. The confidence is measured by 2 characteristics, the Area under the curve (AUC) and the p-value (see Shah et al., 2016 for further details). The higher the AUC and the lower the p-value then the more confident the prediction.)

- c. Consider the Physchem properties of each analogue and identify those with estimated Log Kow values within 1.5 of the estimate for the target chemical Log Kow (e.g., +1.5 and -1.5 of 3.785)
  - De-select any analogues that are below 2 and above 5. How many analogues are now selected for Read-Across?
     (5, Propachlor, benoxacor, 2,4-chloro-2-methylphenoxy) propionic acid, ethyl 5,5-diphenyl-2-isoxazoline-3-carboxylate, and benzophenone)



#### United States Environmental Protection Agency

GenraPred 🗢	Sort: Er	ndpoint name 🗢	↓å	Min+ 1 ♦	Min- 1 ◆	Similarit	y Weight	Show Page	gination	Download: File	Туре
	1.00 🔘	0.53 🗙	0.53 🗸	0.50 🗸	0.50 🗸	0.49 🗙	0.49 🗸	0.49 X	0.48 🗸	0.48 X	0.48 🗶
Assay endpoint $\equiv$		4.50	of	A	40.	,bust,	~6	***********	qo	30	70
	2-Hydroxy-4	Phenothrin	Propachlor	Benoxacor	(+/-)-2-(4-C	Nicosulfuron	Ethyl 5,5-di	Deltamethrin	Benzopheno.	. Dicyclohexyl.	. lambda-Cyh
▼											
<ul> <li>Properties (7)</li> </ul>											
Boiling (°C)	327.4	350.2	292.1	305.4	298.0	281.2	363.4	430.1	305.4	331.0	422.6
Henry's Law (at	2.48e-10	2.85e-08	1.24e-07	6.35e-09	1.65e-09	5.43e-11	4.16e-08	1.87e-09	7.13e-07	3.08e-08	1.12e-08
Mass (g/mol)	228.25	350.46	211.69	260.11	214.65	410.41	295.34	505.21	182.22	330.42	449.85
Melting (°C)	65.5	61.4	76.9	107.6	94.6	142.0	84.4	99.9	48.3	66.1	49.4
Vap. press. (m	2.31e-07	1.43e-07	0.000231	4.44e-06	1.9e-05	1.07e-10	8.56e-07	1.49e-08	0.00191	8.7e-07	7.5e-09
Water sol. (mol	0.000213	2.33e-07	0.0032	7.8e-05	0.00742	0.0229	2.48e-05	1.86e-09	0.000752	1.2e-05	2.2e-09
log Kow	3.785	6.012	2.182	2.700	3.129	0.0122	3.152	6.200	3.181	5.834	6.799
CHR:adrenal gland											
Rows: 255					Total Row	s: 255					Filtered: 255

ii. Once the desired source analogues are selected, click the 'Run Readacross" button to re-derive the GenRA predictions using the GenraPred engine.

0.48 ✓ 0.48 X 0.48 X	0.49 🗙	0.49 🗸							
ao 30 2%		0.45	0.49 X	0.50 🗸	0.50 🗸	0.53 🗸	0.53 🗙	1.00 🔘	
	مميد	~bob	\$ <del>777</del> 5	"}{}}~	Fb	of.	4,00		Assay endpoint $\equiv$
in Benzopheno Dicyclohexyl lambda-Cyh	Deltamethrin	Ethyl 5,5-di	Nicosulfuron	(+/-)-2-(4-C	Benoxacor	Propachlor	. Phenothrin	2-Hydroxy-4	
								<b>'</b>	V
									> Properties (7)
									CHR:adrenal gland
									CHR:alanine aminotr
									CHR:albumin
								*	CHR:alkaline phosph.
									CHR:appearance an
									CHR:aspartate amin
									CHR:body weight
•									CHR:bone
Filtered: 255			: 255	Total Row					Rows: 255

(The prediction is 1, still consistent with the POS empirical data for CHR-liver.)

- d. Export and save the results as an Excel file.
  - i. Does the file export the target predictions as well as all 10 analogues' data?

(NO, just target chemical predictions and selected 5 analogues)

ii.





hem_id	DTXCID002405	DTXCID002405_ur	pred_class	ACT	AUC	pval	DTXCID504	274 DTXCID504	274_ur DTXCID809572	DTXCID809572_ur	DTXCID904194	DTXCID904194_ur	DTXCID5020360	DTXCID5020360_	UDTXCID101961	DTXCID101961_units
ple	target		_				analog		analog		analog		analog		analog	
referred name	2-Hydroxy-4-met	thoxybenzophenor	e				Propachlor		Benoxacor		(+/-)-2-(4-Chloro	-2-methylphenoxy	Ethyl 5,5-diphen	yl-2-isoxazoline-3	Benzophenone	
sstox_sid	DTXSID3022405						DTXSID4024	274	DTXSID3029572		DTX5ID9024194		DTXSID7040360		DTXSID0021961	
sstox cid	DTXCID002405						DTXCID504	74	DTXCID809572		DTXCID904194		DTXCID5020360		DTXCID101961	
imilarity	1.00001						0.5254	78938	0.502310231	L	0.497943006		0.493418743		0.482142857	7
Mass g/mol	228.247							11.69	260.11		214.65		295.338		182.222	
Melting *C	65,5044							5.9232	107.642		94,6045		84.3575		48,2661	
Boiling °C	327,437							2.105	305.371		297,982		363,43		305,417	
og Kow	3.7854							18155	2.69953		3.12858		3.15207		3.18141	
/ap. press. mmHg	2.30687E-07						0.0002		4.43675E-06		1.90404E-05		8.55809E-07		0.00191301	
Nater sol. mol/L	0.000212758							20445	7.79988E-05		0.00741697		2.48317E-05		0.000751735	
ienry's Law atm-m3/mole	2.48418E-10						1.238		6.35321E-05		1.65471E-09		4.16158E-08		7.12881E-07	
lydrogen Bond Donors	2.404105-10						1.230	AL-07	0.333216-03		2.004716-03		4.101306-00		,	
Hydrogen Bond Acceptors																
	DTYCID002405	DTXCID002405 -	prod class	- ACT	* AUC	* pval	- DTVCID504	74 TTXCID504	274 - DTXCID809572 -	DTVCID809572 -	DTXCID904194	DTVCID004104		DTYCID5030261 -	DTVCID101961	DTXCID101961 vits
HR:adrenal gland	no effect		Neg	ACI	0 AUC	• pvar	1 no_effect	- DTXCID504	no data	DTACID809572]*	no data		no data	DTACID302030(*	no_effect	D1XCID101901 * IIIS
HR:aurenai gianu HR:alanine aminotransferas			Pos		1	0	1	65.5 mg/kg/day			no data		no data		no data	
HR:alanine aminotransferas HR:albumin	no data		F 03		1	U	no data	oora mg/ kg/ day								
			Pos		1	0	1	125.2 ma/ka/d-	no_data		no_data no data		no_data		no_data	
CHR:alkaline phosphatase (a			POS Neg		1	0	1 1 no effect	125.3 mg/kg/day					no_data		no_data	
	no_effect					0		ACA melleday	no_data		no_data		no_data		no_data	
CHR:aspartate aminotransfer			Pos		1		1	16.1 mg/kg/day			no_data		no_data		no_data	and the false of
HR:body weight	pos_effect		Pos		1	0	1	6.25 mg/kg/day			no_data		no_data			0 mg/kg/day
HR:bone	no_effect		Neg		0	0	1 no_data		no_data		no_data		no_data		no_effect	
HR:bone marrow	no_effect		Neg		0	0	1 no_effect		no_data		no_data		no_data		no_data	
HR:brain	pos_effect		Pos		0.522	0	0.405	222.9 mg/kg/day			no_data		no_data		no_effect	
HR:cholesterol	pos_effect		Pos		1	0	1	19.3 mg/kg/day			no_data		no_data		no_data	
HR:clinical signs	pos_effect		Pos		1	0	1	25 mg/kg/day			no_data		no_data			0 mg/kg/day
HR:clitoral gland	no_effect		Neg		0	0	1 no_data		no_data		no_data		no_data		no_effect	
HR:creatine phosphokinase			Pos		1	0	1	125.3 mg/kg/day			no_data		no_data		no_data	
HR:ear	no_effect		Neg		0	0	1 no_effect		no_data		no_data		no_data		no_data	
HR:epididymis	no_effect		Neg		0	0	1 no_effect		no_data		no_data		no_data		no_effect	
HR:erythrocyte (rbc) count (			Neg		0	0	1 no_effect		no_data		no_data		no_data		no_data	
HR:esophagus	no_effect		Neg		0	0	1 no_effect		no_data		no_data		no_data		no_effect	
HR:estrous cycle	pos_effect		Pos		1	0	1	0.625 mg/kg/day	no_data		no_data		no_data		no_data	
HR:eye	no_effect		Neg		0	0	1 no_effect		no_data		no_data		no_data		no_effect	
HR:food consumption	pos_effect		Pos		1	0	1	25 mg/kg/day	no_data		no_data		no_data		60	0 mg/kg/day
HR:full gross necropsy	no_effect		Neg		0.478	0	0.51 no_effect		no_data		no_data		no_data		70	0 mg/kg/day
HR:gallbladder	no_effect		Neg		0	0	1 no_data		no_data		no_data		no_data		no_effect	
HR:gamma glutamyl transfe	pos_effect		Pos		1	0	1	53.6 mg/kg/day	no_data		no_data		no_data		no_data	
HR:glucose	no_effect		Neg		0	0	1 no_effect		no_data		no_data		no_data		no_data	
HR:harderian gland	no_effect		Neg		0	0	1 no_data		no_data		no_data		no_data		no_effect	
HR:heart	no effect		Neg		0	0	1 no effect		no data		no data		no data		no effect	
HR:hematocrit (hct)	pos effect		Pos		1	0	1	292.1 mg/kg/day			no data		no data		no data	
HR:hemoglobin (hgb)	pos effect		Pos		1	0	1	292.1 mg/kg/day			no data		no data		no data	
HR:intestine large	no effect		Neg		0	0	1 no effect		no data		no data		no data		no effect	
HR:intestine small	no_effect		Neg		0	0	1 no_effect		no data		no data		no_data		no_effect	
HR:kidney	pos effect		Pos		1	0	1	125.3 mg/kg/day			no data		no data			5 mg/kg/day
CHR:lactic acid dehydrogena:					-	~	no data		no data		no data		no data		no data	
HR:lacuc aciu denyurogena:	no effect		Ner		0	0	1 no effect		no_data		no_data		no_data		no effert	

- e. Are there any subsequent analysis after this supported by GenRA? (No, any subsequent analysis is contingent on the end user expertise. However, the excel file columns can be sorted and filter e.g., filtering by positive effect etc.)
- Filter the column D (pred\_class) to show only positive effects.
   Have the excel file ready to be shared with the breakout room and filter based on only positive effects





chem_id	DTXCID002405	DTXCID002405_ur pred_class	ACT	Α
role	target			
preferred name	2-Hydroxy-4-m	ethoxybenzophenone		
dsstox_sid	DTXSID3022405			
dsstox_cid	DTXCID002405			
similarity	1.000	01		
Mass g/mol	228.2	47		
Melting °C	65.50	14		
Boiling °C	327.4	37		
log Kow	3.78	54		
Vap. press. mmHg	2.30687E-	07		
Water sol. mol/L	0.0002127	58		
Henry's Law atm-m3/mole	2.48418E-	10		
Hydrogen Bond Donors				
Hydrogen Bond Acceptors				
chem_id 🔹	DTXCID002405	DTXCID002405    pred_class	ACT	▼ A
CHR:adrenal gland	no_effect 4	Sort A to Z		0
CHR:alanine aminotransferas	not offer	-		1
CHR:albumin	no data 🗛	S <u>o</u> rt Z to A		
CHR:alkaline phosphatase (al	pos_effec S	ort by Color	>	1
CHR:appearance and color	no effect		>	0
CHR:aspartate aminotransfer	pos effec	iheet <u>V</u> iew	7	1
CHR:body weight	pos_effec	Clear Filter From "pred_class"		1
CHR:bone	no effect	ilter by Color	>	0
CHR:bone marrow	no effect			0
CHR:brain	pos effec	Text <u>F</u> ilters	>	0.522
CHR:cholesterol	pos effec	Search	Q	1
CHR:clinical signs	pos effec		/	1
CHR:clitoral gland	no effect	(Select All)		0
CHR:creatine phosphokinase	pos effec			1
CHR:ear	no effect			0
CHR:epididymis	no effect	Pos		0
CHR:erythrocyte (rbc) count d	no effect	TN		0
CHR:esophagus	no_effect	TP		0
CHR:estrous cycle	pos_effec			1
CHR:eye	no_effect			0
CHR:food consumption	pos_effec			1
CHR:full gross necropsy	no_effect			0.478
CHR:gallbladder	no_effect			0
CHR:gamma glutamyl transfe	pos_effec			1
CHR:glucose	no_effect			0
CHR:harderian gland	no effect			0
CHR:heart	no effect	OK Canc	el	0
CHR:hematocrit (hct)	pos effect	FUS	.:	1
CHR:hemoglobin (hgb)	pos effect	Pos		1





nethoxybenzophenc 5 001 247 044 437 554 -07 758 -10 •••••••••••••••••••••••••••••••••••	pred_class Pos Pos Pos Pos Pos Pos Pos Pos	JT ACT	▼ AUC 1 1		)	
5 001 247 044 437 854 -07 758 -10	pred_class Pos Pos Pos Pos Pos Pos Pos Pos	J ACT	1	0	)	• DTX
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001 247 044 137 354 -07 758 -10	Pos Pos Pos Pos Pos Pos	ACT	1	0	)	
247 044 437 554 -07 758 -10	Pos Pos Pos Pos Pos Pos	JT ACT	1	0	)	
044 437 554 -07 758 -10	Pos Pos Pos Pos Pos Pos	JT ACT	1	0	)	
437 354 -07 758 -10	Pos Pos Pos Pos Pos Pos	J ACT	1	0	)	• DTX
354 -07 758 -10	Pos Pos Pos Pos Pos Pos	JT ACT	1	0	)	
-07 758 -10	Pos Pos Pos Pos Pos Pos	JT ACT	1	0	)	
-10	Pos Pos Pos Pos Pos Pos	JT ACT	1	0	)	
-10	Pos Pos Pos Pos Pos Pos	-▼ ACT	1	0	)	
	Pos Pos Pos Pos Pos Pos	-¥ ACT	1	0	)	_
TXCID002405	Pos Pos Pos Pos Pos Pos	J ACT	1	0	)	
▼ DTXCID002405	Pos Pos Pos Pos Pos Pos	JT ACT	1	0	)	
▼ DTXCID002405	Pos Pos Pos Pos Pos Pos	JT ACT	1	0	)	
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g. Sort the pval from smallest to largest value (The higher the AUC and the lower the p-value then the more confident the prediction.) Here again show the excel file and sort the pval.

ACT	Auc	TTXCID
<	A J Sort Smallest to Largest	
	A ↓ Sort Largest to Smallest	
	Sor <u>t</u> by Color	>
	Sheet <u>V</u> iew	>
	Sciear Filter From "pval"	
	Filter by Color	>
	Number <u>Filters</u>	>
	Search	Q
	- 20 (Select All) - 20 (A5 - 20 0.45 - 20 0.45 - 20 0.555 - 20 0.555 - 20 0.655 - 20 0.677 - 20 0.677 - 20 0.675 - 20 0.675 - 20 0.625 - 20 0.825 - 20 1	
	ОК	Cancel no_effe

		4110				
ACT		AUC		pval		X
	0.522		0		0.405	
	0.522		0			
	0.522		0		0.465	
	0.522		0		0.565	
	0.679		0		0.65	
	0.65		0		0.655 n	┝
	0.671		0		0.67	
	0.669		0		0.675	
	0.679		0		0.685	
	0.512		0		0.825	
	1		0		1	
	1		0		1	
	1		0		1	
	1		0		1	
	1		0		1	
	1		0		1	
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	1		0		1	
	1		0		1	
	-		-			





## Reflection

1. In what case example from your work environment would GenRA be useful?

[Open response]

2. What have you learned about the process and workflow used to find information in GenRA?

[Open response]

3. What challenges did you encounter, and how did you solve them?

[Open response]