The OECD QSAR Toolbox for Grouping Chemicals into Categories

OECD (Q)SAR Toolbox v.4.4.1

Example illustrating endpoint vs. endpoint correlation for apical endpoints

Outlook

• Background

- Objectives
- The exercise
- Workflow

Background

This presentation is designed to introduce the user to:

- Illustration of different types of endpoint vs. endpoint correlations using:
 - LLNA and GPMT skin sensitization data;
 - DPRA and LLNA skin sensitization data;
 - > Skin sensitization and Ames mutagenicity data.

Outlook

- Background
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- Workflow

Objectives

This presentation demonstrates a number of functionalities of the Toolbox:

• Illustration of endpoint vs. endpoint correlations using different types of endpoint data

Outlook

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- Workflow

The exercise

- Illustration of different endpoint data correlations:
 > LLNA vs. GPMT skin sensitization data
 - > DPRA (reactivity) vs. LLNA (skin sensitization) data
 - > LLNA (skin sensitization) vs. Ames mutagenicity data

Outlook

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- Workflow

Workflow

- The Toolbox has six modules which are typically used in a workflow:
 - Chemical Input
 - Profiling
 - Data
 - Category Definition
 - Filling Data Gaps
 - Report
- In this example we will use the modules in a different order, tailored to the aims of the example.

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 - Correlation of data background

Correlation of endpoint data Background

- This functionality introduces the user to the opportunity to analyze correlations between selected gap filling endpoint (endpoint used for prediction) and other endpoint data.
- It is applicable for correlation analysis of data presented in ordinary, interval or ratio scale.
- If correlated data are measured in interval or ratio scale they are transformed in ordinary scale and the strength of the correlation is estimated by Spearman correlation coefficient.
- Basically, this functionality provides a correlation between target endpoint (this is the initial endpoint selected by the user) displayed on ordinate axis (Y-axis) and other endpoint data displayed on abscissa (X-axis).

Correlation of endpoint data Spearman coefficient factor

- Spearman's rank correlation coefficient is a nonparametric rank statistic proposed by Charles Spearman as a measure of the strength of an association between two variables. It assesses how well the relationship between two variables can be described using a monotonic function.
- Spearman correlation coefficient could be used for exploring the correlation between:
 - two ranked variables
 - one measurement variable and one ranked variable (in this case, the measurement variable need to be to converted to ranks)
- Spearman correlation varies from -1 to +1 and the interpretation of the coefficient factor is provided below:
 - 0.00 0.19 very weak correlation
 - 0.20 0.39 weak correlation
 - 0.40 0.59 moderate correlation
 - 0.60 0.79 strong correlation
 - 0.80 1.0 very strong

Outlook

- Background
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- Workflow
 - Correlation of data background
 - Types endpoint correlations

Types of endpoint correlations are as follows:

- Continuous vs. continuous*
- Categorical vs. categorical:
 - ✓ Categorical vs. categorical
 - ✓ Categorized continuous vs. categorical
 - ✓ Categorized continuous vs. categorized continuous*

*Both type correlation is not illustrated in this presentations. They are presented in "Tutorial_4_TB 4.4_Illustrating endpoint vs. endpoint correlation using ToxCast data"

Outlook

- Background
- Objectives
- The exercise
- Workflow
 - Correlation of data background

• Types endpoint correlations

• Categorical vs. categorical

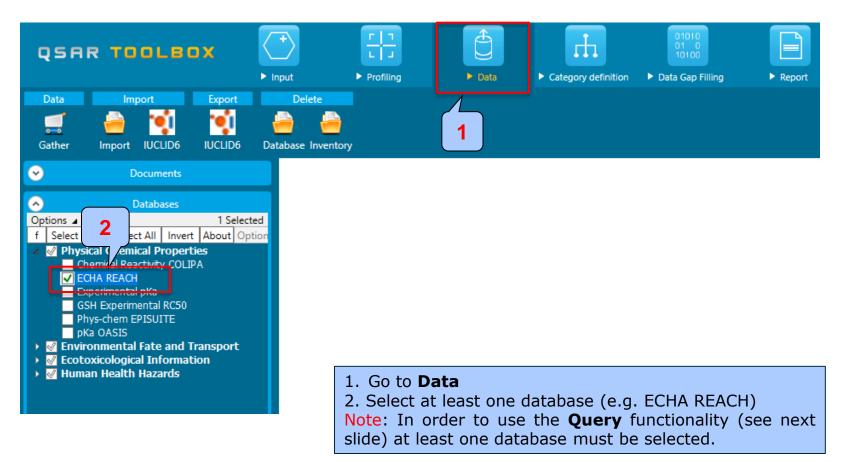
- The aim of this type of correlation is to illustrate how categorical types of data correlate with each other.
- Categorical type data is the statistical data type consisting of categorical variables or of data that has been converted into that form. Such data is binary Ames data (dichotomic type): positive, negative or polytomic type data such as GPMT data: strong, weak and negative.
- Two examples illustrating this type correlation will be demonstrated:
 - Example 1: Correlation of two types skin sensitization data
 - LLNA (Positive, Negative) vs. GPMT (Weakly positive, Strongly positive, Negative)
 - Example 2: Correlation of skin sensitization and Ames mutagenicity data
 - LLNA (Negative, Weakly positive, Strongly positive) vs. AMES (Positive, Equivocal, Negative)
- Step by step workflow is presented on next few slides. Summary of the workflow steps are provided below:
 - Query Tool and select FSQ file(step 1)
 - Gather experimental data (step 2)
 - Enter Gap filling (step 3)
 - Perform correlation between endpoints (step 4).

Example 1: Correlation between LLNA and GPMT data

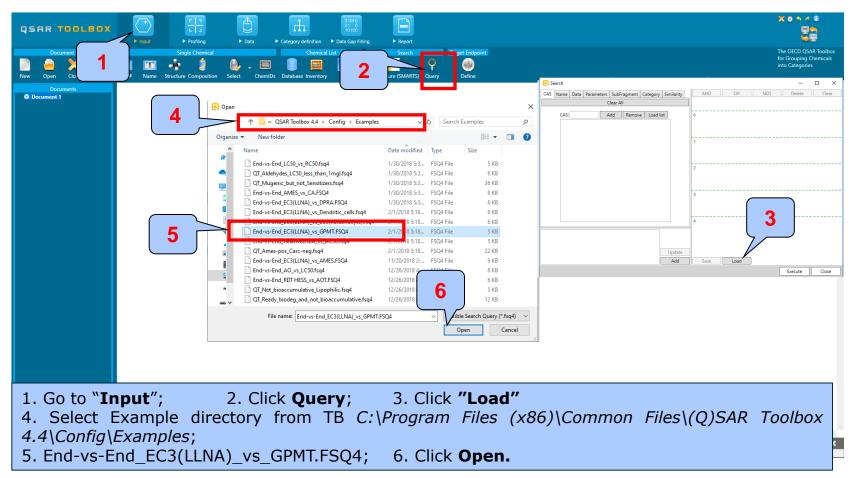
According to OECD Guideline 406 for testing of chemicals for skin sensitization, the LLNA test can be used as a first stage in the assessment of skin sensitization potential. If a positive result is seen, a test substance may be designated as a potential sensitizer, and it may not be necessary to conduct a further guinea pig test.¹

Based on that Guideline the aim of the illustrated correlation is to show how the capacity of LLNA test is compatible to that of the GPMT assay.

¹ <u>https://www.oecd-ilibrary.org/docserver/9789264070660-</u> en.pdf?expires=1573465805&id=id&accname=guest&checksum=D7DE1063A1EA331FCA23AC81BE1FDAC1



Example 1: Correlation between LLNA and GPMT data

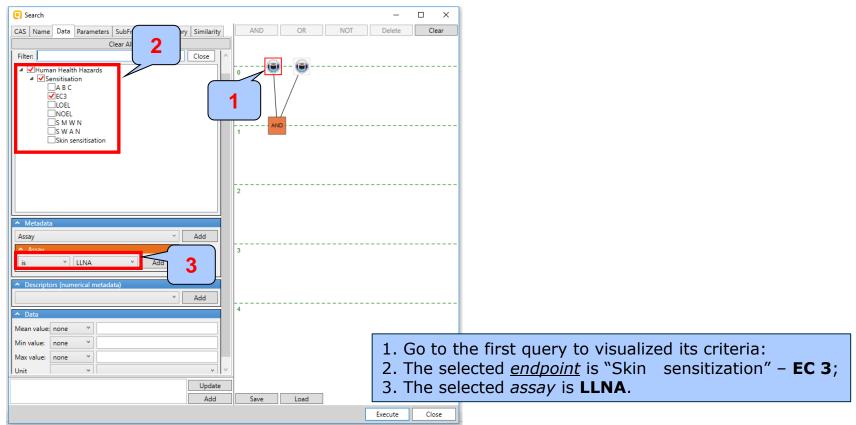


The OECD (Q)SAR Toolbox for Grouping Chemicals into Categories

QSAR	TOOLBO	× ► Input	► Profiling	► Data	Category definition	01010 01 0 10100 ► Data Gap Filling	► Report			X 0 5 6 0
	Document	# 1	Single Chemical	🗘 - 🎟	Chemical	List	Search	Target Endpoint		The OECD QSAR Toolbox for Grouping Chemicals into Categories
	en Close Sav Documents		Structure Composition	Select ChemIDs	Database Inventory	List Su	bstructure (SMARTS)			Developed by LMC, Bulgaria
😵 Docum				🕘 Query to	ol			×		
					is created with t	he following d	lata sources:			
					sensitisation da	tabase (norm	alised)			
				Skin Sensitiz	zation					
				Would you lik	ke to restore the	em?				
								Yes No		
					1. C	lick Ye	es.			
										×
0										

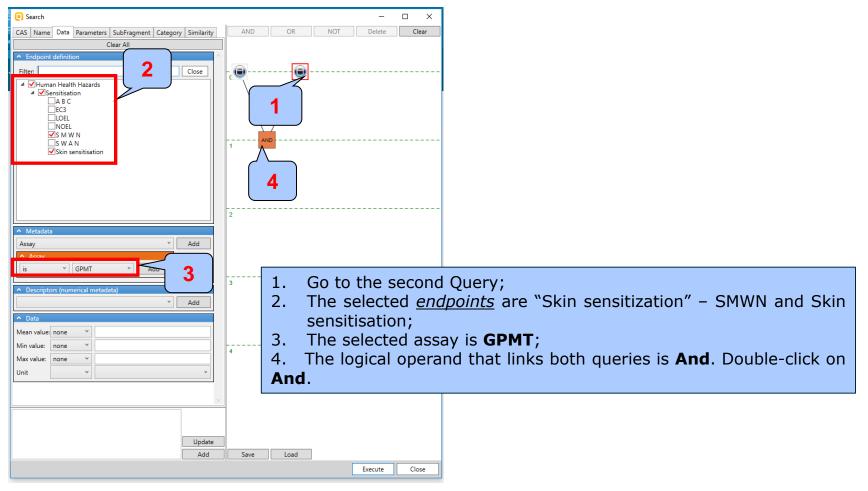
Types of endpoint correlations

Categorical vs. categorical Gather experimental data – step 2



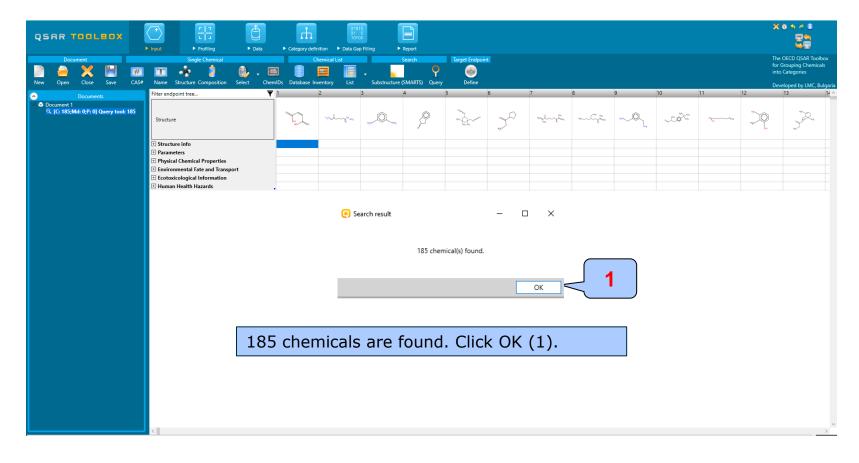
Types of endpoint correlations

Categorical vs. categorical Gather experimental data – step 2

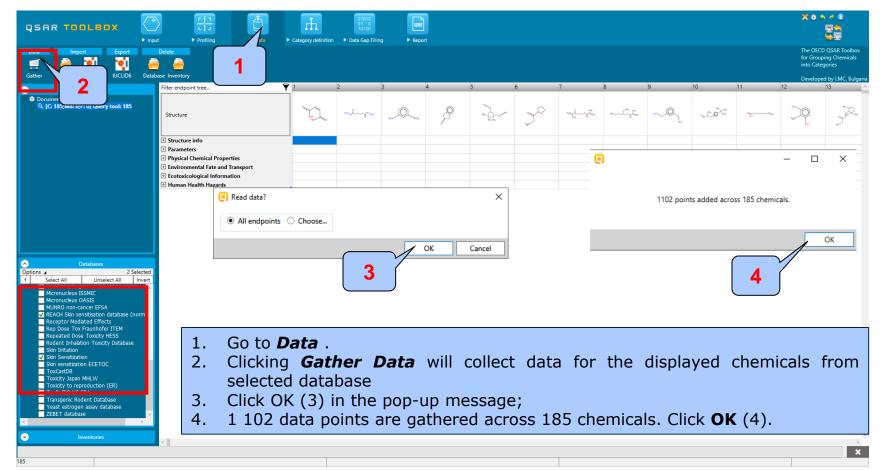


Categorical vs. categorical

Gather experimental data – step 2

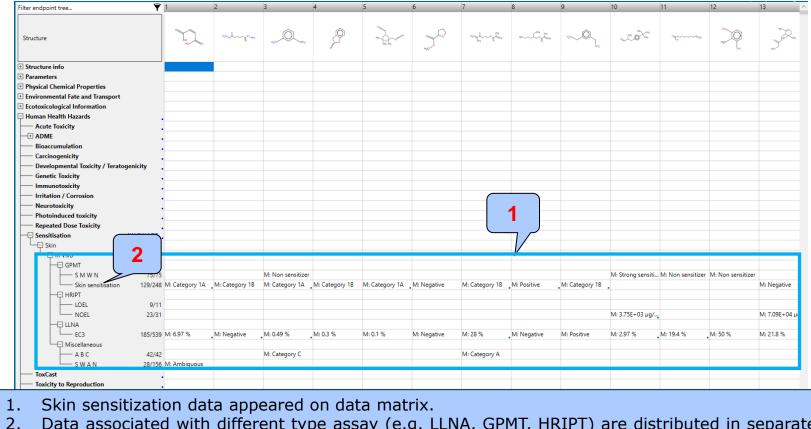


Categorical vs. categorical Gather experimental data – step 2



Categorical vs. categorical Gather experimental data – step 2

Example 1: Correlation between LLNA and GPMT data



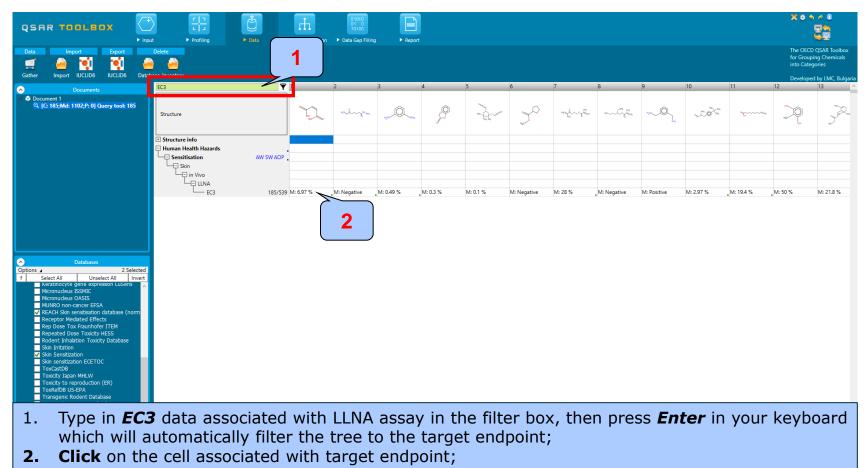
2. Data associated with different type assay (e.g. LLNA, GPMT, HRIPT) are distributed in separate nodes

What is "scale" and "scale conversion" ?

Reminder slide

- Skin sensitisation as an example is a "qualitative" endpoint for which the results are presented with categorical type of data (for example: positive; negative; weak sensitizer; strong sensitizer, etc).
- Skin sensitisation potential data of the chemicals comes from different databases coded with different names (for example: data from John Moores University of Liverpool are: *Strongly sensitizing, Moderately sensitizing etc.*; data from European centre for Ecotoxicology and Toxicology of chemicals are: *Positive, Negative, and Equivocal*).
- The main purpose of the scales is to unify all data available in the Toolbox databases for a certain endpoint.
- "Scale conversion" is the TB instrument to create conversions between scales. It is more reasonable to convert from a more informative to less informative scale.
- The default scale for Skin Sensitisation data is "Skin Sensitisation ECETOC". It converts all skin sensitization data into: Positive and Negative. This allows skin sensitization data to be used as much as possible for gap filling purposes.

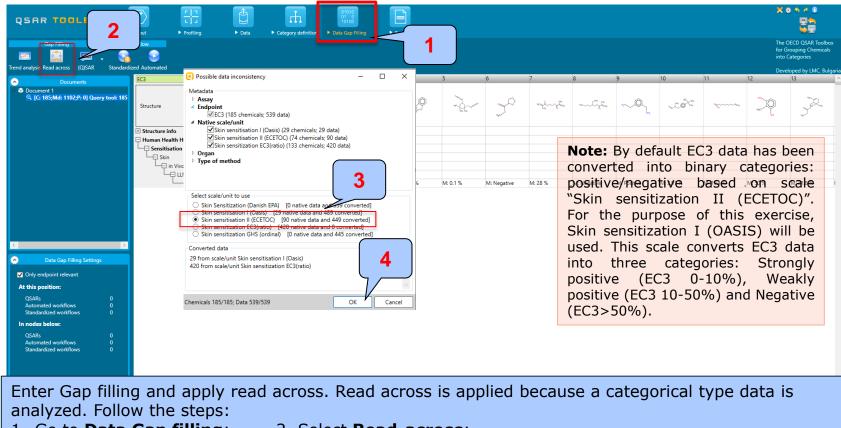
Categorical vs. categorical Define target endpoint – step 3



Types endpoint correlations

Categorical vs. categorical Enter Gap filling – step 4

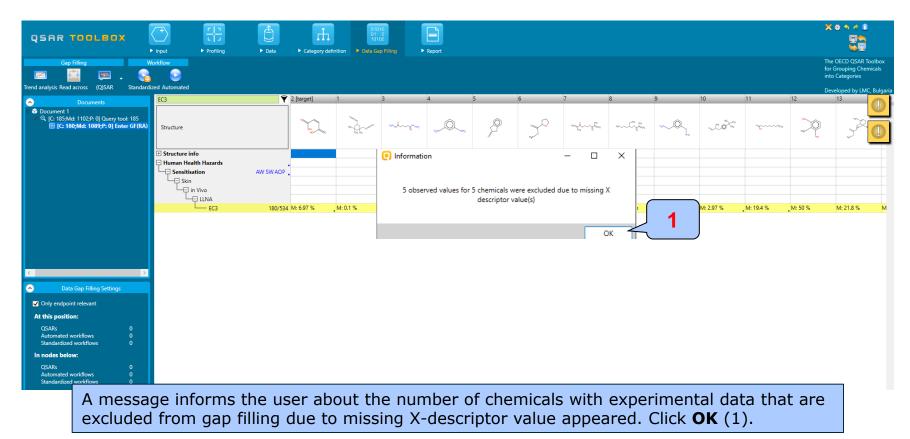
Example 1: Correlation between LLNA and GPMT data



- 1. Go to Data Gap filling; 2. Select Read-across;
- 3. Select **Skin sensitization II (ECETOC)** scale (see Note);

4. Click **OK**;

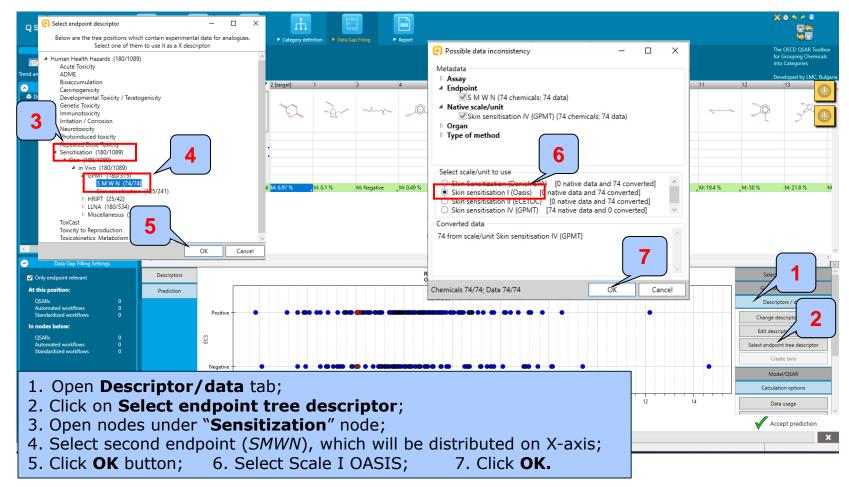
Categorical vs. categorical Enter Gap filling – step 4



Types endpoint correlations

Categorical vs. categorical

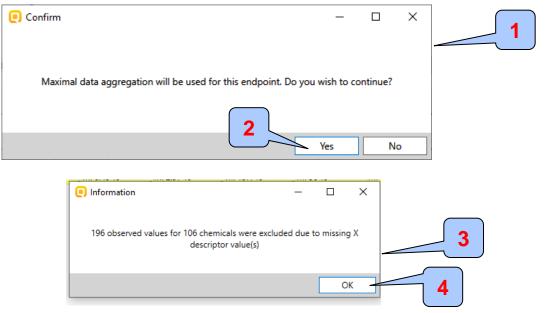
Perform correlation between LLNA and GPMT data- step 5



Categorical vs. categorical

Perform correlation between LLNA and GPMT data- step 5

Example 1: Correlation between LLNA and GPMT data



- 1. As only one data point per chemical is permitted in this type of correlation, the maximal data point will be considered for each chemical;
- 2. Click Yes;

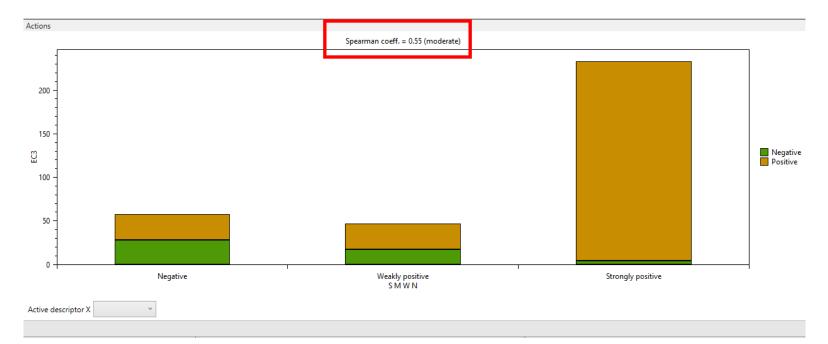
A message informs the user about the number of chemicals with experimental data that are excluded from gap filling due missing data for SMWN endpoint appears. This will not affect the value of correlation coefficient;
 Click OK.

Types endpoint correlations

Categorical vs. categorical

Interpretation of correlation results (LLNA vs. GPMT)

Example 1: Correlation between LLNA and GPMT data



• Correlation analysis between two categorical type skin sensitization data (LLNA and GPMT) shows moderate endpoint correlation (Spearman coefficient is 0.55).

- The second example illustrating categorical vs. categorical type correlation is:
 - Example 2: Correlation between Skin sensitization and Ames mutagenicity data
 - LLNA (Negative, Weakly positive, Strongly positive)
 - AMES (Positive, Equivocal, Negative)
- Step by step workflow is presented on next few slides. Summary of the workflow steps are provided below:
 - Query Tool and select FSQ file(step 1)
 - Gather experimental data (step 2)
 - Enter Gap filling (step 3)
 - Perform correlation between endpoints (step 4).

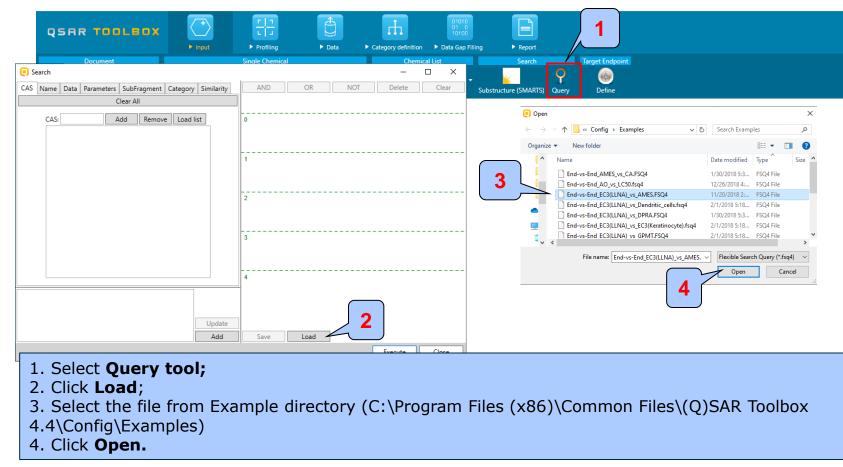
Categorical vs. categorical Gather experimental data – step 2

Example 2: Correlation between LLNA and AMES data

The correlation between LLNA and Ames data has been investigated in view of the proposition that mutagenicity data can be used as part of an integrated approach to testing and assessment (IATA) for skin sensitisation^{1,2}.

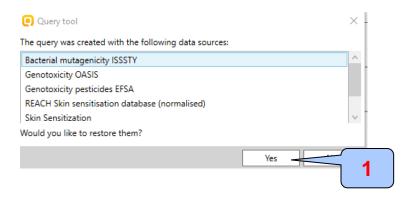
¹ Patlewicz G., C. Kuseva, A. Kesova, I. Popova, T. Zhechev, T. Pavlov, D. W. Roberts, O. Mekenyan, Towards AOP application – Implementation of an integrated approach to testing and assessment (IATA) into a pipeline tool for skin sensitization. *Regul. Toxicol. Pharmacol.* 69 (3) (2014), 529 - 545.
 ² Wolfreys, M,A, Basketter, A. D. Mutagens and Sensitizers—An Unequal Relationship?. Cutaneous and Ocular Toxicology. 2004

Categorical vs. categorical Gather experimental data – step 2



Categorical vs. categorical Gather experimental data – step 2

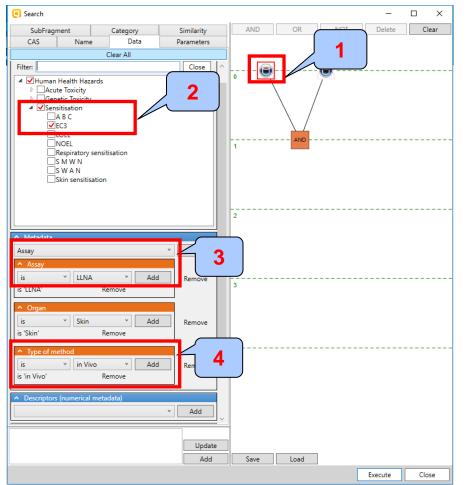
Example 2: Correlation between LLNA and AMES data



1. Click **Yes** to confirm that you want to restore the databases used during the creation of the . fsq file.

Types endpoint correlations

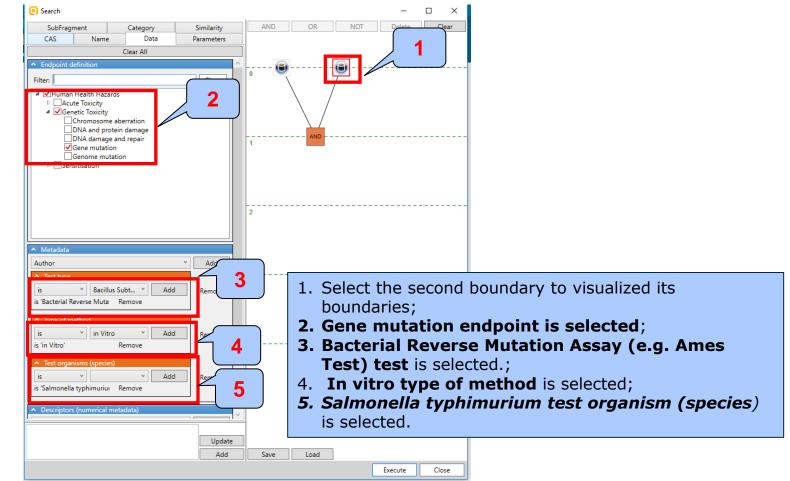
Categorical vs. categorical Gather experimental data – step 2



- 1. Select the first boundary to visualized its boundaries:
- 2. EC3 is selected;
- 3. LLNA assay is selected;
- **4. In vivo** type of method is selected.

Types endpoint correlations

Categorical vs. categorical Gather experimental data – step 2



Categorical vs. categorical

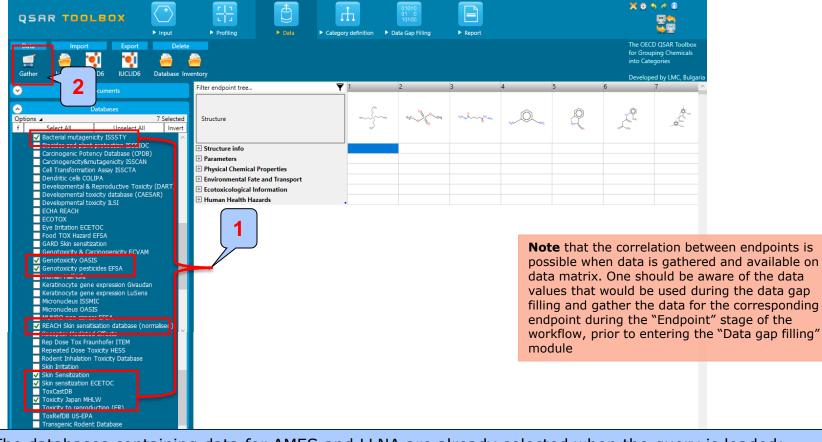
Gather experimental data – step 2

Earch result		_		×	
	425 chemical(s) found.				
			C	к –	1
425 chemicals	are found; Click	OK	(1).		

Types endpoint correlations

Categorical vs. categorical Gather experimental data – step 2

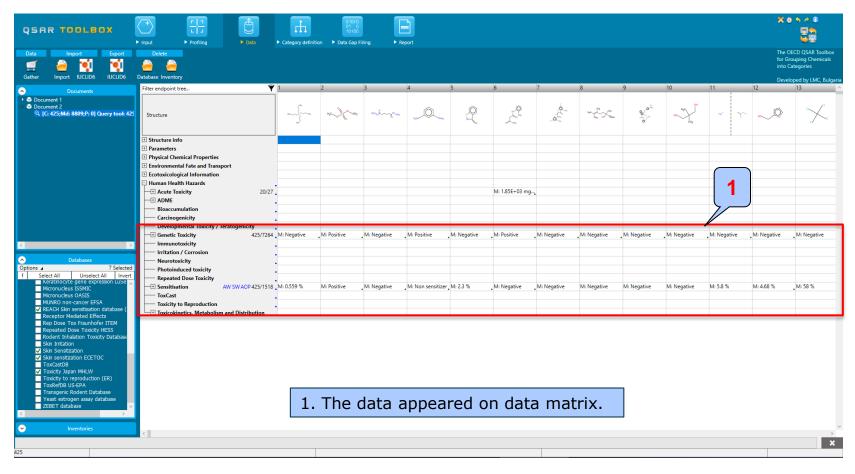
Example 2: Correlation between LLNA and AMES data



The databases containing data for AMES and LLNA are already selected when the query is loaded;
 Click "Gather"

Types endpoint correlations

Categorical vs. categorical Gather experimental data – step 2



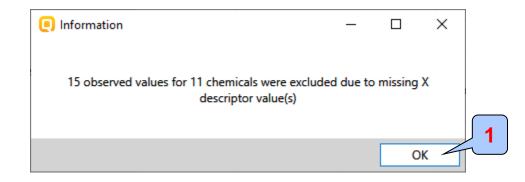
Types endpoint correlations

Categorical vs. categorical Define target endpoint – step 3

QSAR TOO 3	→ T T → Data → C	ategory definition	01010 01 0 10100 Data Gap Filling		ר	Possible data inconsistency - X	
Trend analy is Read across (Q)SAR Standardized	2			- 1	J	Assay for Gr Assay into C Endpoint Devel	CD QSAR Toolbox uping Chemicals tegories ped by LMC, Bulgaria
Occuments	Filter endpoint tree Filter endpoint tree Structure Structure info Structure info Parameters Physical Chemical Properties	1		3 **~~ ⁰ ~~~ ₆ ****	4	 ✓ Isc3 (425 chemicals; 983 data) ✓ Native scale/unit ✓ Skin sensitisation I (Oasis) (52 chemicals; 52 data) ✓ Skin sensitization EC3(ratio) (244 chemicals; 717 data) ✓ Organ ✓ Type of method 	
	Environmental Fate and Transport Ecotoxicological Information Human Health Hazards Actual Coxicity 20/27 ADME Bioaccumulation Carcinogenicity Developmental Toxicity / Teratogenicity		M: Positive	M: Negative	. M: Posit	Select scale/unit to use Skin Sensitization (Danish EPA) [0 native data and 983 converted] Skin sensitization I (Oasis) [52 native data and 894 converted] Skin sensitization I (CertorO) [214 native data and 769 converted] Skin sensitization GHS (ordinal) [0 native data and 760 converted] Converted data Converted data	. M: Negative
Data Gap Filling Settings Only endpoint relevant At this position: Select a cell with a rigid (bold) path	Immunotoxicity Irritation / Corrosion Neurotoxicity Photoinduced toxicity Repeated Dose Toxicity Sensitisation AW SW AOP	2				52 from scale/unit Skin sensitisation I (Oasis) 717 from scale/unit Skin sensitization EC3(ratio) 5 Chemicals 425/425; Data 983/983 OK Cancel	
Automated workflows 0 Standardized workflows 0	Skin G in Vivo G GPMT 112/21 HRIPT 34/5 ULNA TO TO T	, <i>V</i>		M: Category 1B		M: 45 µg/cm2	
	C3 425/9 C3 425/9 C3 425/9 C3 Miscellaneous 77/2 Undefined Assay 15/1 ToxCast Toxicity to Reproduction Toxicokinetics, Metabolism and Distribution		A: Positive	M: Negative	M: 0.49 S		. ^{M: 58 %}

Categorical vs. categorical Enter Gap filling – step 4

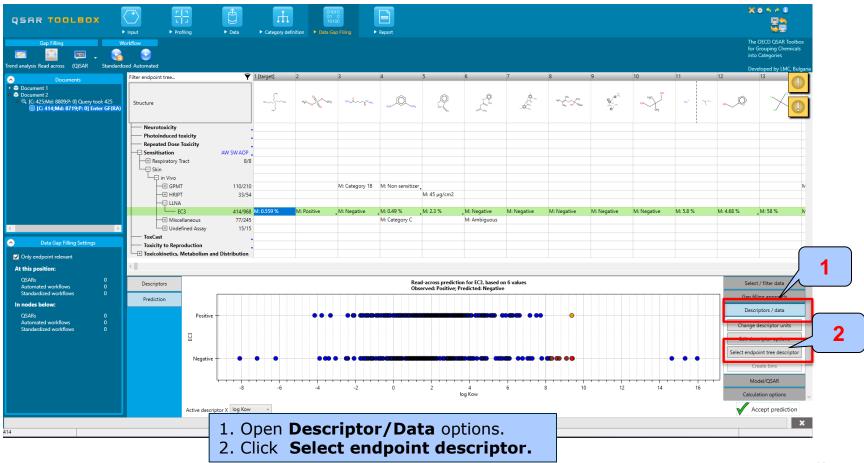
Example 2: Correlation of LLNA and AMES data



The message informs the user that some chemicals are excluded from gap filling; Click OK (1);

Categorical vs. categorical

Perform correlation between LLNA and AMES data – step 5



Categorical vs. categorical

Perform correlation between LLNA and AMES data – step 5

Example 2: Correlation between LLNA and AMES data

Select endpoint descriptor	-		×
Below are the tree positions which contain experimental data for ana Select one of them to use it as a X descriptor:	logues.		
▲ Human Health Hazards (414/8719)			,
Acute Toxicity (20/27)			
ADME			
Bioaccumulation			
Carcinogenicity Developmental Toxicity / Teratogenicity 4 Genetic Toxicity (414/7192) 4 in Vitro (414/7006)			
Bacillus Subtilis Recombination Assay (2/6)			
 Bacterial Reverse Mutation Assay (e.g. Ames Test) (414/6465) 			
Gene mutation (414/6465)			
Escherichia coli (47/100)			
 Salmonella typhimurium (414/6365) 			
▷ No Data (1/1)			
2 Vithout S9 (387/2975) Vithout S9 (396/3068)			
DNA Damage and Repair Assay, Unscheduled DNA Synthesis in Mammali.	an Cells ir	Vitro (4	/5)
in Vitro Mammalian Cell Micronucleus Test (19/19)			
in Vitro Mammalian Cell Transformation Assay (1/2)			
in Vitro Mammalian Chromosome Aberration Test (184/416)			
Mammalian Cell Gene Mutation Assay (46/71)			
Other (2/5)			
Single Cell Gel Electrophoresis (comet) Assay (1/2)			
Sister Chromatid Exchange Assay in Mammalian Cells (4/13)			
			_
	OK	Car	ncel

1. In the Select endpoint descriptor open the branches below Genetic Toxicity;

2. Select "**With S9**" under In Vitro|Bacterial Reverse Mutation Assay (e.g. Ames Test)|Gene Mutation| Salmonella typhimurium;

3. Click **OK**.

Types endpoint correlations

Categorical vs. categorical

Perform correlation between LLNA and AMES data – step 5 Example 2: Correlation between LLNA and AMES data

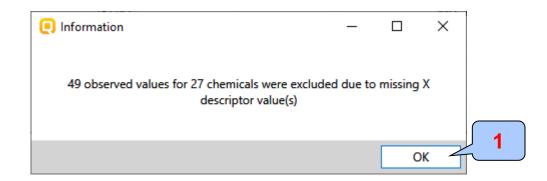
1etadata		🖸 Confirm — 🗆 🔿
Strain	^	
 ✓ No Information For Experimental Cells (13 chemicals; 13 data) ✓ Other (1 chemicals; 2 data) ✓ TA 100 (363 chemicals; 630 data) ✓ TA 102 (100 chemicals; 108 data) ✓ TA 104 (53 chemicals; 108 data) ✓ TA 104 (53 chemicals; 515 data) ✓ TA 1537 (301 chemicals; 467 data) ✓ TA 1538 (184 chemicals; 225 data) ✓ TA 1538 (184 chemicals; 26 data) ✓ TA 7001 (8 chemicals; 6 data) 	ł	Maximal data aggregation will be used for this endpoint. Do you wish to continue?
☑TA 7002 (8 chemicals; 8 data) ☑TA 7003 (7 chemicals: 7 data)	~	2 Yes No
Select scale/unit to use		
Gene mutation I [2846 native data and 129 converted] Gene mutation II [129 native data and 0 converted]		
onverted data		
29 from scale/unit Gene mutation II		
emicals 387/387; Data 2975/2975	Cancel	

Possible data inconsistency window appears. Click **OK** (1). As only one data point per chemical is permitted in this type of correlation, the maximal value will be considered for each chemical; Click **Yes** (2).

Categorical vs. categorical

Perform correlation between LLNA and AMES data – step 5

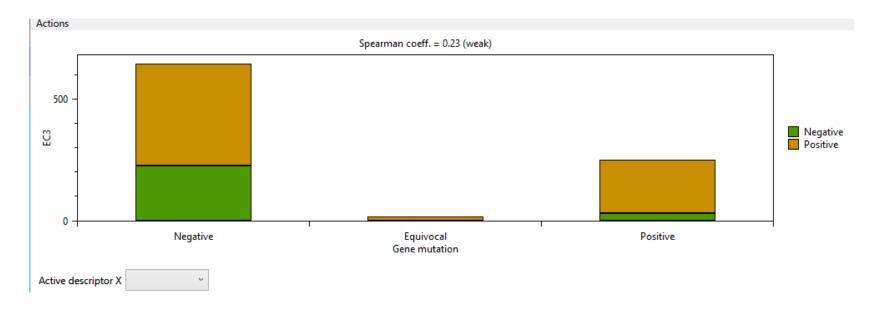
Example 2: Correlation between LLNA and AMES data



The pop-up message informs on the total number gathered data across the number chemicals that will be excluded in trend analysis due to missing X descriptor value(s). These are analogues with no AMES data. This will not affect the value of correlation coefficient; 1. Click **OK**;

Categorical vs. categorical

Interpretation of correlation results (LLNA vs. AMES)



Correlation analysis between two categorical type data: LLNA and AMES shows weak correlation between two endpoints (Spearman coefficient is 0.23).

Outlook

- Background
- Objectives
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- Workflow
 - Correlation of data background

Types endpoint correlations

- Categorical vs. categorical
- Categorized continuous vs. categorical

Types endpoint correlations Categorized continuous vs. categorical

- The aim of this type correlation is to illustrate how categorized continuous and categorical type of data correlate with each other.
- Categorized continuous data is the continuous type data (e.g LC50 or AC50, EC3, %) converted into categories.
- In this example we will illustrated how DPRA ratio data (%) correlates with LLNA data:
 - DPRA (ratio data expressed in % and converted in categories)
 - LLNA (categorical type: Strongly positive, Weakly positive, Negative)
- Step by step workflow is presented on next few slides. Summary of the workflow steps are provided below:
 - Query Tool and select FSQ file(step 1)
 - Gather experimental data (step 2)
 - Define target endpoint (step 3)
 - Enter Gap filling (step 4)
 - Perform correlation between endpoints (step 5).

Categorized continuous vs. categorical Query Tool and select FSQ file - step 1

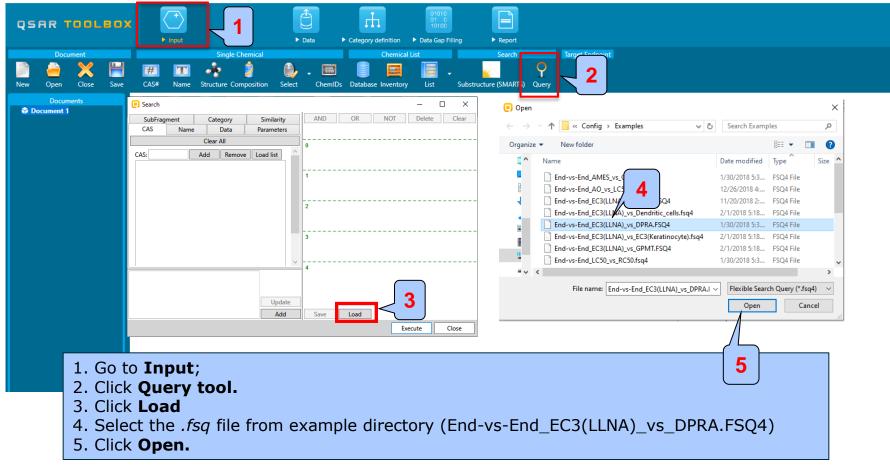
Example: Correlation between LLNA (Strongly positive, Weakly positive, Negative) and DPRA (Strongly positive, Weakly positive, Negative) data

The purpose of performing of this correlation is to establish whether information from non-testing methods (DPRA, *in chemico* assay) provides sufficient evidence about a substance's skin sensitization potential as compared to that which has been elicited in an in vivo assay (LLNA).

Types endpoint correlations

Categorized continuous vs. categorical Query Tool and select FSQ file - step 1

Example: Correlation between LLNA (Strongly positive, Weakly positive, Negative) and DPRA (Strongly positive, Weakly positive, Negative) data



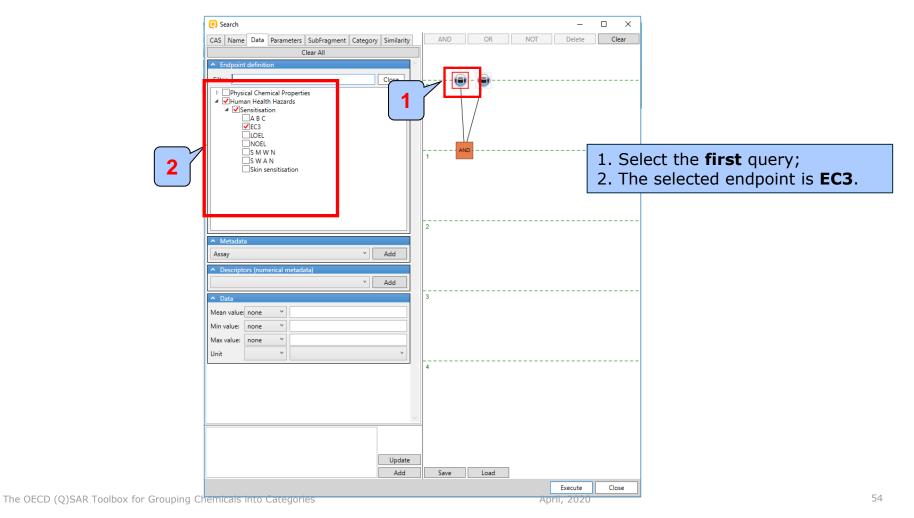
Types endpoint correlations Categorized continuous vs. categorical *Query Tool and select FSQ file - step 1*

Example: Correlation between LLNA (Strongly positive, Weakly positive, Negative) and DPRA (%) data

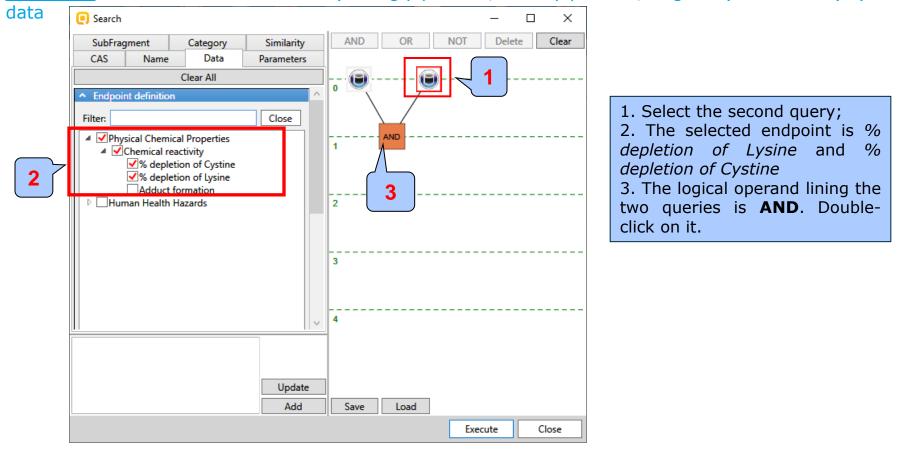
📒 Query tool	×
The query was created with the following data	i sources:
Chemical Reactivity COLIPA	
REACH Skin sensitisation database (normalis	ed)
Skin Sensitization	
Would you like to restore them?	
	1 Yes No

Click OK (1) in the message informing that the databases used to create the query will be restored.

Types endpoint correlations Categorized continuous vs. categorical *Query Tool and select FSQ file - step 1*

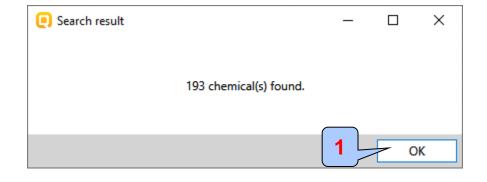


Types endpoint correlations Categorized continuous vs. categorical *Query Tool and select FSQ file - step 1*



Types endpoint correlations Categorized continuous vs. categorical *Query Tool and select FSQ file - step 1*

Example: Correlation between LLNA (Strongly positive, Weakly positive, Negative) and DPRA (%) data



193 chemicals are found. Click OK (1).

Types endpoint correlations

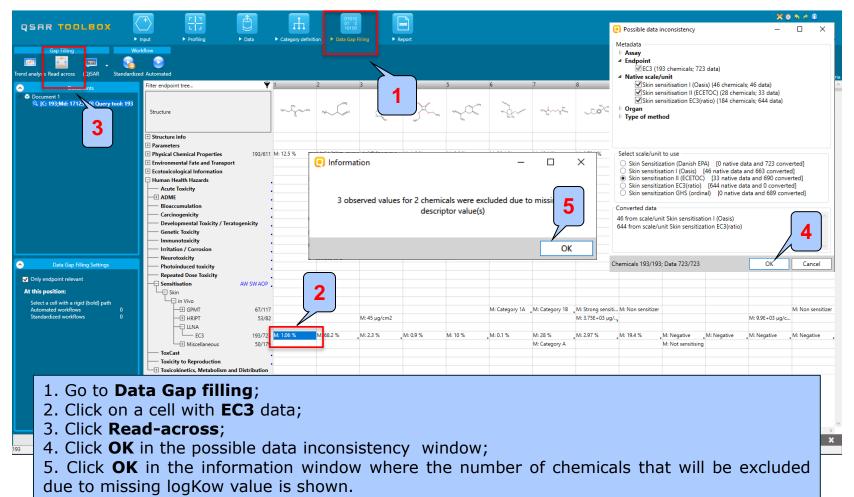
Categorized continuous vs. categorical

Gather experimental data – step 2

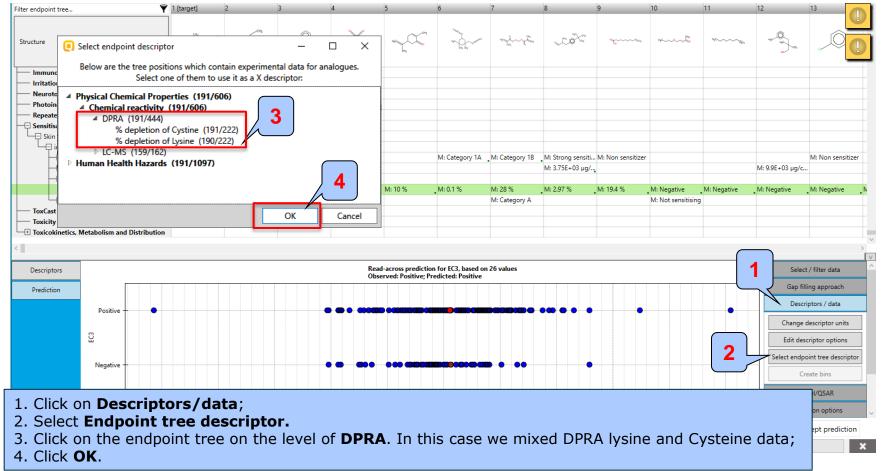
Gather Import	Here Control of the second sec	Category is the processing of	► Report	s	The OFCD QSAR Toolbox for Grouping Chemicals into Categories	×
Documents Databases Options Select All Unselect All Inv Physical Chemical Properties Chemical Reactivity COLIPA		44-57-540 44-57-540 44-560	Q X	~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~	All endpoints Choose	
Experimental pKa GSH Experimental RC50 Phys-chem PEJSUTTE pKa 0ASIS Environmental Fate and Transport Ectoxicological Information Commental Fate and Transport Ectoxicological Information Acute Oral toxicty D8 AOME Database Batterial mutagencity ISSSTY Biocdes and phatn protection ISSBIOC Carcinogenic Potency Database (OPD8) Carcinogenicky-Mutagencity ISSCN Carcinogenicky-Mutagencity ISSCN Carcinogenicky-Mutagencity ISSCN Carcinogenicky-Mutagencity ISSCN Carcinogenicky-Mutagencity ISSCN Carcinogenicky-Mutagencity ISSCN Carcinogenicky-Mutagencity ISSCN Developmental toxicty database (CAESAR) Developmental toxicty database (CAESAR) Developmental toxicty database (CAESAR) Developmental toxicty database (CAESAR) ECOTOX Evel Tration ECETOC Exact Trox User SESA Human HaFLife Keratinocyte gene expression Givaudan Keratinocyte gene expression Gi	Parameters Physical Chemical Properties Commental Fate and Transport Ecco szicological Information Hur an Health Hazards		nts added across 193	chemicals.	с х 5	OK Cancel
Rep Dose Tox Haumofer ITEM Repeated Dose ToxY HESS Rodent Inhaaton Toxicity Database Sin Intation Sin Sensitization ECETOC ToxTasthik Toxicity Japan MHLW Toxicity to reproduction (ER) ToxRef BU 3-59A Transgenic Rodent Database Yeast estrogen assay database ZEBET database	2. / 3. / 4. /	Click Gather l Click OK to co	button; Ilect all d	ata for	ady selected when the all endpoints; cals are collected, click	

Types endpoint correlations Categorized continuous vs. categorical

Enter Gap filling – step 4



Types endpoint correlations Categorized continuous vs. categorical *Enter Gap filling – step 4*



Types endpoint correlations

Categorized continuous vs. categorical Perform correlation between LLNA and DPRA data – step 5

Possible data inconsistency - ×	
Metadata Assay DPRA (191 chemicals; 444 data) Endpoint V % depletion of Cystine (191 chemicals; 222 data) V % depletion of Lysine (190 chemicals; 222 data) Native scale/unit	Maximal data aggregation will be used for this endpoint. Do you wish to continue?
Select scale/unit to use • • • • • • • • • • • • • • • • • • •	Yes No
Converted data 444 from scale/unit Chemical reactivity DPRA (ratio)	
Chemicals 191/191; Data 444/444 OK Cancel	

- 1. Select Chemical reactivity DPRA 13% (ordinal) scale;
- 2. Click **OK**;
- 3. Click **OK** in the pop-up message.

Types endpoint correlations

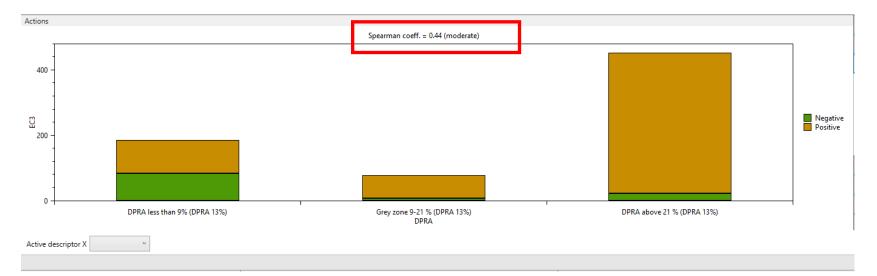
Categorized continuous vs. categorical Perform correlation between LLNA and DPRA data – step 5

Example: Correlation between LLNA (Strongly positive, Weakly positive, Negative) and DPRA (%) data

3 observed values for 1 chemical were excluded	l due to missir	ıg X desc	riptor
value(s)			1
		C	ж

1. Click OK.

Categorized continuous vs. categorical Interpretation of correlation results (LLNA vs. DPRA)



- In this example we have correlated continues DPRA (%) (plotted on the x axis) data distributed into 3 bins and categorical LLNA data (Strongly positive, Weakly positive, Negative)
 - Less than 9%
 - Grey zone 9 21%
 - Above 21%
- The high value of Spearman coefficient (0.44) shows moderate correlation between DPRA and LLNA data

Summary

- Different types of correlations have been illustrated in this tutorial based on the type of endpoint data:
 - Categorical vs. categorical:
 - Categorized continuous vs. categorical
- Correlation analysis has been evaluated by Spearman coefficient;
- Moderate endpoint correlations have been obtained for 2 out of 3 illustrated examples.