

# **How to judge whether QSAR/read-across predictions can be trusted? Novel approach for establishing model's applicability domain**

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## **Supplementary materials**

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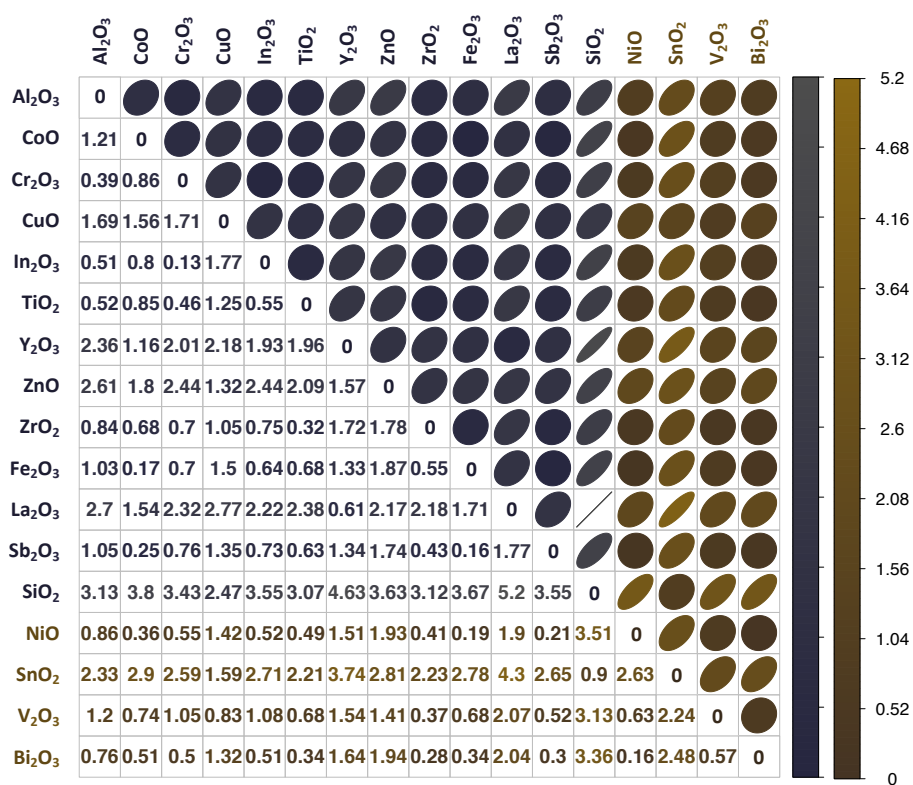
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## Case Study 1

**Table S1.** Descriptors used for Nano-QSAR model development along with experimentally observed and predicted values of nanotoxicity to *E. coli* bacteria for training and validation compounds (MeOx NPs) reported by Pathakoti *et al.*<sup>[1]</sup>

MeOx	QMELECT	LZELEHHO	Autoscaled QMELECT	Autoscaled LZELEHHO	Set*	Experimental $-\log(\text{LC}_{50})$	Predicted $-\log(\text{LC}_{50})$	Residuals	Standardized cross-validated residuals	Leverages
Al <sub>2</sub> O <sub>3</sub>	0.10	0.21	-0.36	0.95	T	2.42	1.78	0.64	1.26	0.22
CoO	0.10	0.17	-0.49	-0.25	T	3.13	3.41	-0.28	-0.72	0.10
Cr <sub>2</sub> O <sub>3</sub>	0.10	0.20	-0.55	0.61	T	2.06	2.15	-0.09	-0.36	0.18
CuO	0.13	0.18	1.05	0.02	T	4.24	4.06	0.18	0.22	0.19
In <sub>2</sub> O <sub>3</sub>	0.10	0.20	-0.65	0.53	T	2.83	2.18	0.65	1.22	0.19
TiO <sub>2</sub>	0.11	0.19	-0.10	0.50	T	2.14	2.6	-0.46	-1.07	0.11
Y <sub>2</sub> O <sub>3</sub>	0.10	0.13	-0.60	-1.40	T	5.79	4.98	0.81	1.67	0.24
ZnO	0.12	0.13	0.97	-1.30	T	5.8	5.9	-0.1	-0.50	0.49
ZrO <sub>2</sub>	0.11	0.18	0.02	0.20	T	2.58	3.12	-0.54	-1.18	0.08
Fe <sub>2</sub> O <sub>3</sub>	0.10	0.17	-0.45	-0.08	T	2.4	3.19	-0.79	-1.68	0.10
La <sub>2</sub> O <sub>3</sub>	0.09	0.12	-1.17	-1.63	T	4.96	4.93	0.03	-0.09	0.31
Sb <sub>2</sub> O <sub>3</sub>	0.10	0.17	-0.29	-0.10	T	3.12	3.33	-0.21	-0.56	0.08
SiO <sub>2</sub>	0.15	0.24	2.61	1.94	T	2.54	2.37	0.17	0.80	0.70
NiO	0.10	0.18	-0.37	0.09	V	3.79	3.01	0.78	1.19	0.10
SnO <sub>2</sub>	0.14	0.22	1.94	1.34	V	2.53	2.77	-0.24	-0.58	0.41
V <sub>2</sub> O <sub>3</sub>	0.11	0.17	0.23	-0.10	V	3.48	3.68	-0.2	-0.52	0.09
Bi <sub>2</sub> O <sub>3</sub>	0.10	0.18	-0.26	0.20	V	3.55	2.92	0.63	0.92	0.09

\* T – training set; V – validation set; QMELECT – the absolute electronegativity of the metal atom; LZELEHHO – the absolute electronegativity of the metal oxide



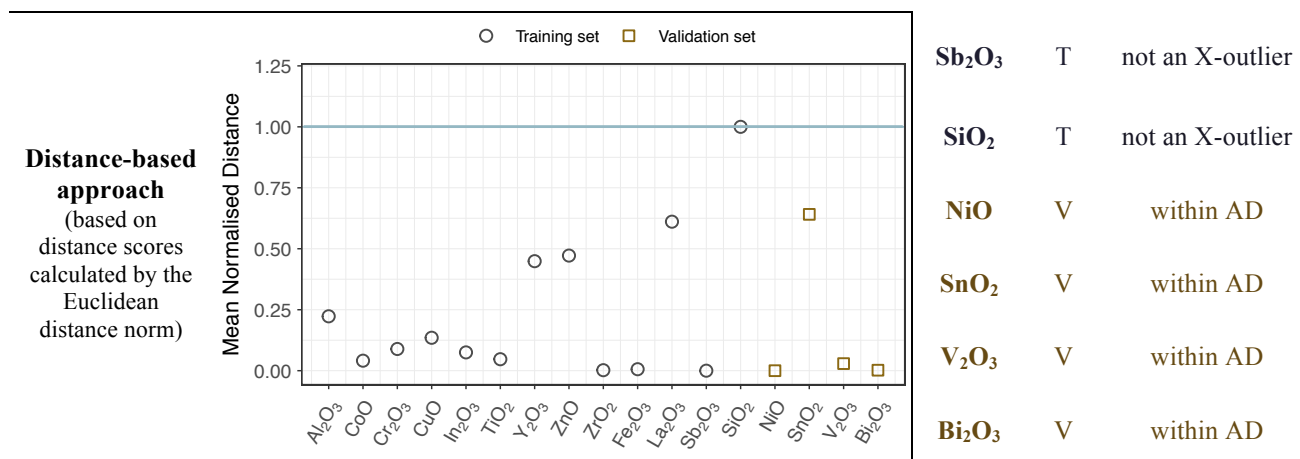
**Figure S1.** Euclidean distance matrix between the training (indicated in gray color) and validation (indicated in goldenrod color) compounds. The color intensity and shape of the ellipses displayed in the upper triangular part of a distance matrix reflect the values of the corresponding Euclidean distances expressed in numerical forms in the lower triangular part of this matrix. The smallest Euclidean distances (indicated as big circles) are considered as significant. In contrast, the highest distances that reflect the dissimilarity between compounds are considered as insignificant and shown as a thin ellipse.

**Table S2.** Input matrix data for 95% and 99% confidence interval boundaries estimation using the Student's *t*-distribution and developed in-house *R*-code

MeOx	Average Euclidean distance	Standardized cross-validated residuals	Set
Al <sub>2</sub> O <sub>3</sub>	1.50	1.26	1
CoO	1.22	-0.72	1
Cr <sub>2</sub> O <sub>3</sub>	1.33	-0.36	1
CuO	1.72	0.22	1
In <sub>2</sub> O <sub>3</sub>	1.34	1.22	1
TiO <sub>2</sub>	1.23	-1.07	1
Y <sub>2</sub> O <sub>3</sub>	1.90	1.67	1
ZnO	2.12	-0.50	1
ZrO <sub>2</sub>	1.18	-1.18	1
Fe <sub>2</sub> O <sub>3</sub>	1.17	-1.68	1
La <sub>2</sub> O <sub>3</sub>	2.30	-0.09	1
Sb <sub>2</sub> O <sub>3</sub>	1.14	-0.56	1
SiO <sub>2</sub>	3.60	0.80	1
NiO	1.07	1.19	2
SnO <sub>2</sub>	2.60	-0.58	2
V <sub>2</sub> O <sub>3</sub>	1.18	-0.52	2
Bi <sub>2</sub> O <sub>3</sub>	1.06	0.92	2

**Table S3.** Results on the applicability domain determination using: range-based method, geometrical-based method, distance-based method and standardization approach

	Basic theory of the standardization approach		
	MeOx	Set	AD Info.
Range-based approach	Al <sub>2</sub> O <sub>3</sub>	T	not an X-outlier
	CoO	T	not an X-outlier
	Cr <sub>2</sub> O <sub>3</sub>	T	not an X-outlier
	CuO	T	not an X-outlier
	In <sub>2</sub> O <sub>3</sub>	T	not an X-outlier
	TiO <sub>2</sub>	T	not an X-outlier
	Y <sub>2</sub> O <sub>3</sub>	T	not an X-outlier
	ZnO	T	not an X-outlier
	ZrO <sub>2</sub>	T	not an X-outlier
	Fe <sub>2</sub> O <sub>3</sub>	T	not an X-outlier
	La <sub>2</sub> O <sub>3</sub>	T	not an X-outlier

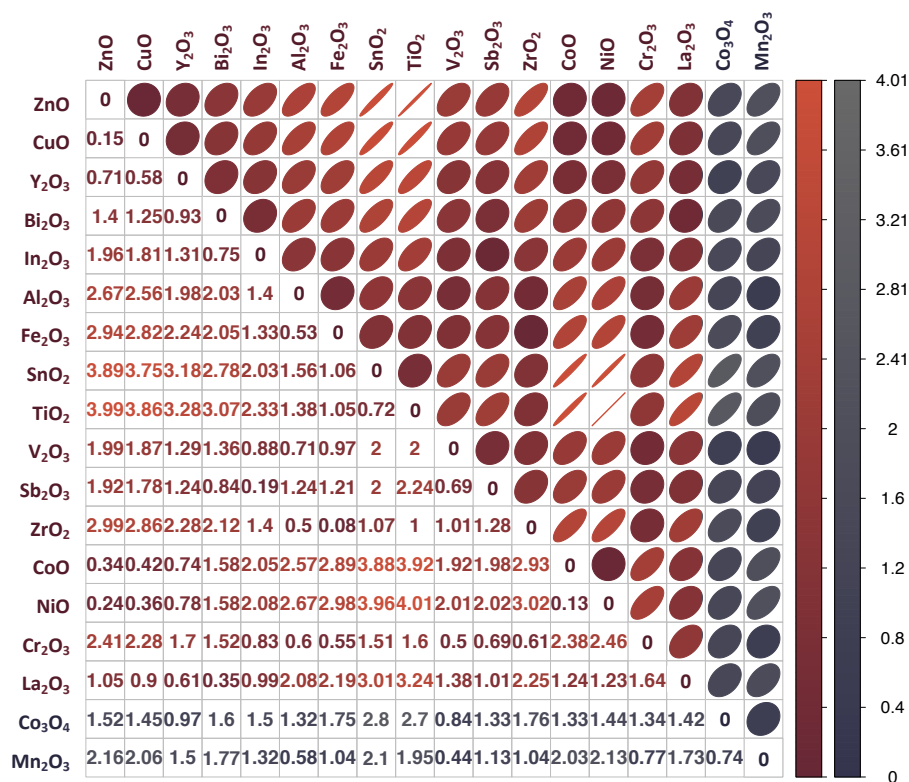


## Case Study 2

**Table S4.** Descriptors used for Nano-QSAR model development along with experimentally observed and predicted values of nanotoxicity to *E. coli* bacteria for training and validation compounds (MeOx NPs) reported by Mu *et al.*<sup>[2]</sup>

MeOx	$\Delta H_{me^+}$	Z/r	Autoscaled $\Delta H_{me^+}$	Autoscaled Z/r	Set*	Experimental $\log(1/EC_{50})$	Predicted $\log(1/EC_{50})$	Residuals	Standardized residuals	Leverages
<b>ZnO</b>	662.4	2.70	-1.29	-1.12	T	3.45	3.39	0.06	0.31	0.17
<b>CuO</b>	713.7	2.74	-1.14	-1.09	T	3.20	3.35	-0.15	-0.88	0.15
<b>Y<sub>2</sub>O<sub>3</sub></b>	837.2	3.33	-0.77	-0.64	T	2.87	3.13	-0.26	-1.51	0.10
<b>Bi<sub>2</sub>O<sub>3</sub></b>	1137.4	2.91	0.10	-0.96	T	2.82	2.87	-0.05	-0.32	0.31
<b>In<sub>2</sub>O<sub>3</sub></b>	1271.1	3.75	0.50	-0.32	T	2.81	2.62	0.19	1.05	0.20
<b>Al<sub>2</sub>O<sub>3</sub></b>	1187.8	5.56	0.25	1.06	T	2.49	2.46	0.03	0.14	0.23
<b>Fe<sub>2</sub>O<sub>3</sub></b>	1363.4	5.46	0.77	0.98	T	2.29	2.25	0.04	0.19	0.13
<b>SnO<sub>2</sub></b>	1717.3	5.80	1.80	1.24	T	2.01	1.89	0.12	0.65	0.29
<b>TiO<sub>2</sub></b>	1575.7	6.56	1.39	1.83	T	1.74	1.95	-0.21	-1.22	0.29
<b>V<sub>2</sub>O<sub>3</sub></b>	1097.7	4.69	-0.01	0.40	T	3.14	2.69	0.45	2.52	0.10
<b>Sb<sub>2</sub>O<sub>3</sub></b>	1233.1	3.95	0.38	-0.17	T	2.64	2.62	0.02	0.08	0.13
<b>ZrO<sub>2</sub></b>	1357.7	5.56	0.75	1.06	T	2.15	2.31	-0.16	-0.94	0.14
<b>CoO</b>	594.6	3.08	-1.48	-0.84	T	3.51	3.39	0.12	0.65	0.25
<b>NiO</b>	596.9	2.90	-1.48	-0.97	T	3.45	3.42	0.03	0.14	0.22
<b>Cr<sub>2</sub>O<sub>3</sub></b>	1266.6	4.84	0.48	0.51	T	2.51	2.48	0.03	0.14	0.08
<b>La<sub>2</sub>O<sub>3</sub></b>	1017.2	2.91	-0.25	-0.96	T	2.87	3.04	-0.17	-1.00	0.19
<b>Co<sub>3</sub>O<sub>4</sub></b>	811.1	4.60	-0.85	0.33	V	3.00	2.94	0.06	0.31	0.36
<b>Mn<sub>2</sub>O<sub>3</sub></b>	1018.0	5.17	-0.25	0.77	V	3.08	2.77	0.36	2.01	0.28

\* T – training set; V – validation set;  $\Delta H_{me^+}$  – the enthalpy of formation of a gaseous cation; Z/r – polarization force parameters.

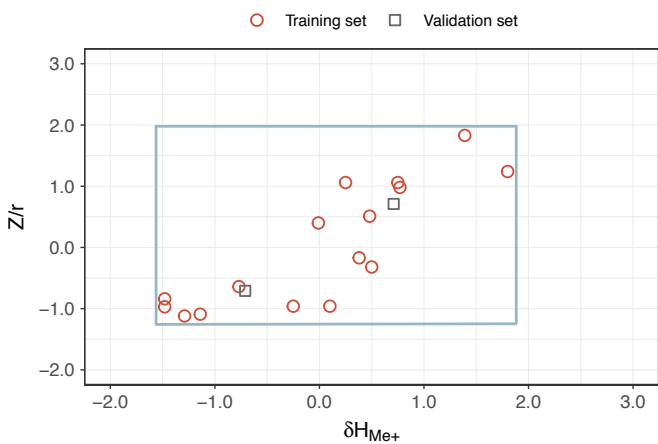
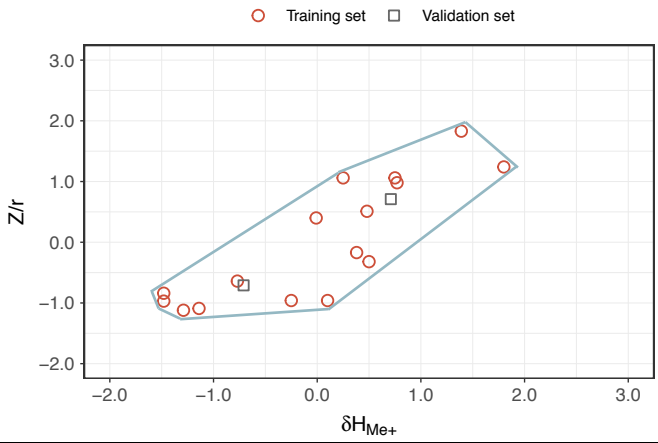
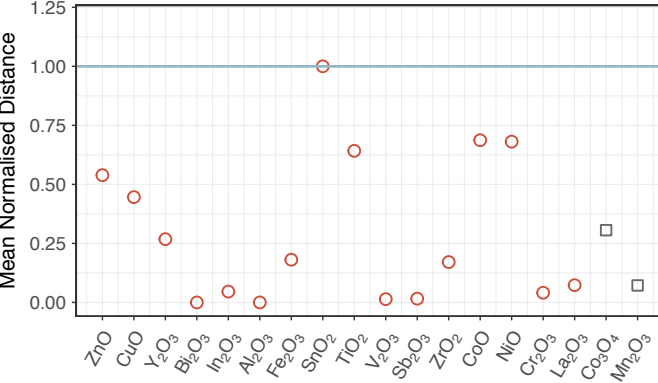


**Figure S2.** Euclidean distance matrix between the training (indicated in tomato color) and validation (indicated in gray color) compounds. The color intensity and shape of the ellipses displayed in the upper triangular part of a distance matrix reflect the values of the corresponding Euclidean distances expressed in numerical forms in the lower triangular part of this matrix. The smallest Euclidean distances (indicated as big circles) are considered as significant. In contrast, the highest distances that reflect the dissimilarity between compounds are considered as insignificant and shown as a thin ellipse.

**Table S5.** Input matrix data for 95% and 99% confidence interval boundaries estimation using the Student's *t*-distribution and developed in-house *R*-code

MeOx	Average Euclidean distance	Standardized residuals	Set
ZnO	1.91	0.31	1
CuO	1.82	-0.88	1
Y <sub>2</sub> O <sub>3</sub>	1.52	-1.51	1
Bi <sub>2</sub> O <sub>3</sub>	1.57	-0.32	1
In <sub>2</sub> O <sub>3</sub>	1.42	1.05	1
Al <sub>2</sub> O <sub>3</sub>	1.63	0.14	1
Fe <sub>2</sub> O <sub>3</sub>	1.66	0.19	1
SnO <sub>2</sub>	2.43	0.65	1
TiO <sub>2</sub>	2.51	-1.22	1
V <sub>2</sub> O <sub>3</sub>	1.37	2.52	1
Sb <sub>2</sub> O <sub>3</sub>	1.36	0.08	1
ZrO <sub>2</sub>	1.69	-0.94	1
CoO	1.93	0.65	1
NiO	1.97	0.14	1
Cr <sub>2</sub> O <sub>3</sub>	1.42	0.14	1
La <sub>2</sub> O <sub>3</sub>	1.54	-1.00	1
Co <sub>3</sub> O <sub>4</sub>	1.57	0.31	2
Mn <sub>2</sub> O <sub>3</sub>	1.49	2.01	2

**Table S6.** Results on the applicability domain determination using: range-based method, geometrical-based method, distance-based method and standardization approach

		<b>Basic theory of the standardization approach</b>		
		<b>MeOx</b>	<b>Set</b>	<b>AD Info.</b>
<b>Range-based approach</b>		<b>ZnO</b>	T	not an X-outlier
		<b>CuO</b>	T	not an X-outlier
		<b>Y<sub>2</sub>O<sub>3</sub></b>	T	not an X-outlier
<b>Geometric approach</b>		<b>Bi<sub>2</sub>O<sub>3</sub></b>	T	not an X-outlier
		<b>In<sub>2</sub>O<sub>3</sub></b>	T	not an X-outlier
		<b>Al<sub>2</sub>O<sub>3</sub></b>	T	not an X-outlier
		<b>Fe<sub>2</sub>O<sub>3</sub></b>	T	not an X-outlier
		<b>SnO<sub>2</sub></b>	T	not an X-outlier
		<b>TiO<sub>2</sub></b>	T	not an X-outlier
		<b>V<sub>2</sub>O<sub>3</sub></b>	T	not an X-outlier
		<b>Sb<sub>2</sub>O<sub>3</sub></b>	T	not an X-outlier
		<b>ZrO<sub>2</sub></b>	T	not an X-outlier
		<b>Distance-based approach</b> (based on distance scores calculated by the Euclidean distance norm)		<b>CoO</b>
<b>NiO</b>	T			not an X-outlier
<b>Cr<sub>2</sub>O<sub>3</sub></b>	T			not an X-outlier
<b>La<sub>2</sub>O<sub>3</sub></b>	T			not an X-outlier
<b>Co<sub>3</sub>O<sub>4</sub></b>	V			within AD
<b>Mn<sub>2</sub>O<sub>3</sub></b>	V			within AD

### Case Study 3

**Table S7.** Descriptors used for Nano-QSAR model development along with experimentally observed and predicted values of nanotoxicity to HaCaT cells for training and validation compounds (MeOx NPs) reported by Pan *et al.*<sup>[3]</sup>

MeOx	DCW (1,3)	Autoscaled DCW(1,3)	Set*	Experimental log(1/LC <sub>50</sub> )	Predicted log(1/LC <sub>50</sub> )	Residuals	Standardized cross-validated residuals	Leverages
Al <sub>2</sub> O <sub>3</sub>	20.32	-1.82	T	1.85	1.82	0.03	-0.01	0.35
Bi <sub>2</sub> O <sub>3</sub>	26.58	-0.12	T	2.50	2.47	0.03	-0.10	0.08
Cr <sub>2</sub> O <sub>3</sub>	24.77	-0.61	T	2.30	2.28	0.02	-0.18	0.11
In <sub>2</sub> O <sub>3</sub>	31.20	1.13	T	2.92	2.95	-0.03	-0.54	0.18
La <sub>2</sub> O <sub>3</sub>	30.57	0.96	T	2.87	2.88	-0.01	-0.42	0.15
NiO	28.85	0.49	T	2.49	2.70	-0.21	-1.86	0.10
Sb <sub>2</sub> O <sub>3</sub>	25.06	-0.54	T	2.31	2.31	0.00	-0.32	0.10
SnO <sub>2</sub>	28.62	0.43	T	2.67	2.68	-0.01	-0.40	0.09
V <sub>2</sub> O <sub>3</sub>	24.20	-0.77	T	2.24	2.22	0.02	-0.19	0.13
WO <sub>3</sub>	27.41	0.10	T	2.56	2.55	0.01	-0.28	0.08
ZnO	33.34	1.71	T	3.32	3.17	0.15	1.11	0.32
ZrO <sub>2</sub>	22.16	-1.32	T	2.02	2.01	0.01	-0.24	0.22
Mn <sub>2</sub> O <sub>3</sub>	28.31	0.35	T	2.64	2.65	-0.01	-0.38	0.09
CoO	25.85	-0.32	V	2.83	2.39	0.44	2.51	0.09
Fe <sub>2</sub> O <sub>3</sub>	21.79	-1.42	V	2.05	1.97	0.08	0.18	0.24
SiO <sub>2</sub>	21.06	-1.62	V	2.12	1.90	0.23	1.13	0.30
TiO <sub>2</sub>	21.28	-1.56	V	1.76	1.92	-0.16	-1.35	0.28
Y <sub>2</sub> O <sub>3</sub>	21.64	-1.46	V	2.21	1.96	0.26	1.33	0.26

\* T – training set; V – validation set; DCW(1,3) – SMILES-based optimal molecular descriptor which includes: aggregation size; individual size; mass percentage of metal elements; cationic charge and molecular weight.



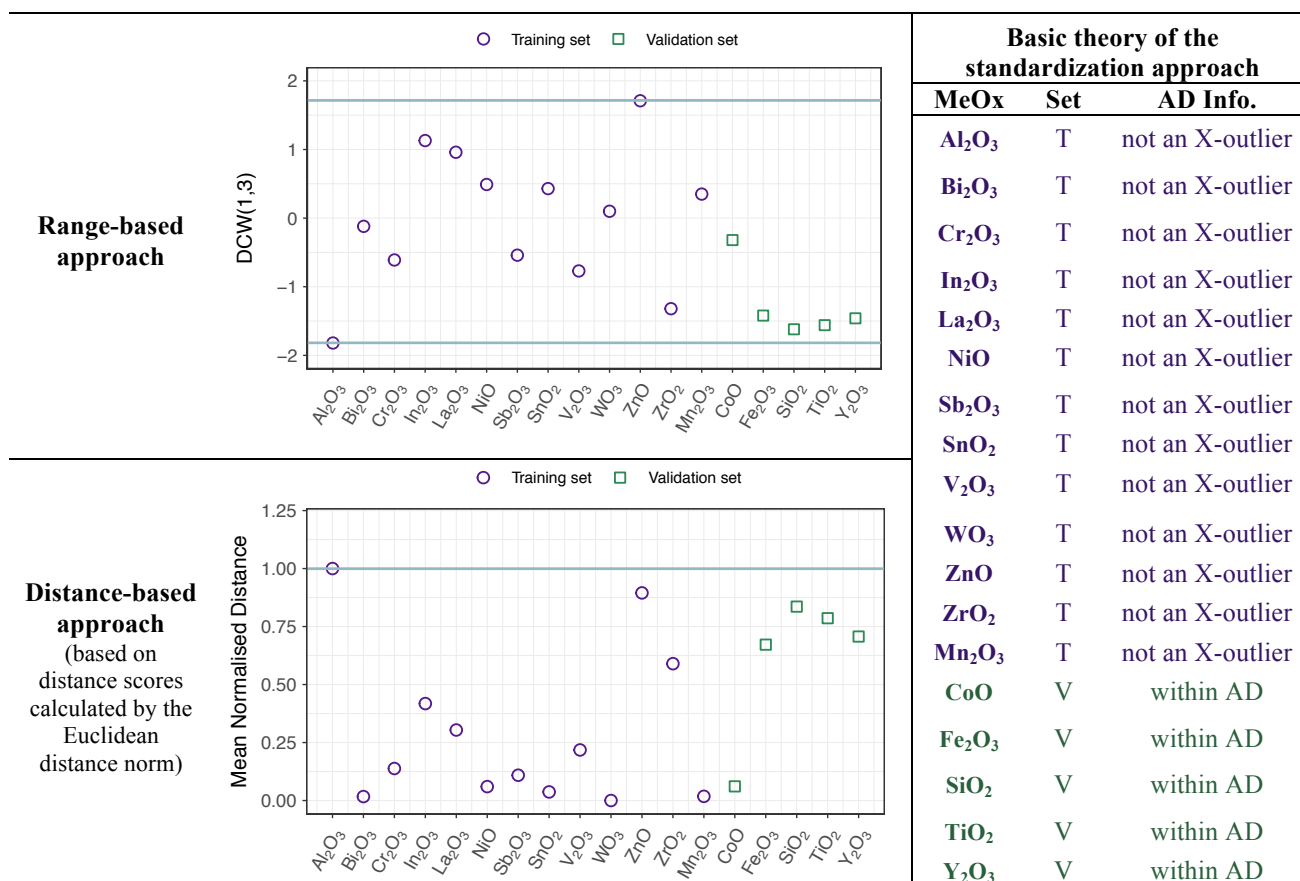
**Figure S3.** Euclidean distance matrix between the training (indicated in purple color) and validation (indicated in seagreen color) compounds. The color intensity and shape of the ellipses displayed in the upper triangular part of a distance matrix reflect the values of the corresponding Euclidean distances expressed in numerical forms in the lower triangular part of this matrix. The smallest Euclidean distances (indicated as big circles) are considered as significant. In contrast, the highest distances that reflect the dissimilarity between compounds are considered as insignificant and shown as a thin ellipse.

**Table S8.** Input matrix data for 95% and 99% confidence interval boundaries estimation using the Student's *t*-distribution and developed in-house *R*-code

<b>MeOx</b>	<b>Average Euclidean distance</b>	<b>Standardized cross-validated residuals</b>	<b>Set</b>
<b>Al<sub>2</sub>O<sub>3</sub></b>	1.97	-0.01	1
<b>Bi<sub>2</sub>O<sub>3</sub></b>	0.87	-0.10	1
<b>Cr<sub>2</sub>O<sub>3</sub></b>	1.01	-0.18	1
<b>In<sub>2</sub>O<sub>3</sub></b>	1.32	-0.54	1
<b>La<sub>2</sub>O<sub>3</sub></b>	1.19	-0.42	1
<b>NiO</b>	0.92	-1.86	1
<b>Sb<sub>2</sub>O<sub>3</sub></b>	0.98	-0.32	1
<b>SnO<sub>2</sub></b>	0.90	-0.40	1
<b>V<sub>2</sub>O<sub>3</sub></b>	1.10	-0.19	1
<b>WO<sub>3</sub></b>	0.85	-0.28	1
<b>ZnO</b>	1.85	1.11	1
<b>ZrO<sub>2</sub></b>	1.51	-0.24	1
<b>Mn<sub>2</sub>O<sub>3</sub></b>	0.88	-0.38	1
<b>CoO</b>	0.85	2.51	2
<b>Fe<sub>2</sub>O<sub>3</sub></b>	1.48	0.18	2
<b>SiO<sub>2</sub></b>	1.65	1.13	2
<b>TiO<sub>2</sub></b>	1.60	-1.35	2
<b>Y<sub>2</sub>O<sub>3</sub></b>	1.51	1.33	2



**Table S9.** Results on the applicability domain determination using: range-based method, geometrical-based method, distance-based method and standardization approach

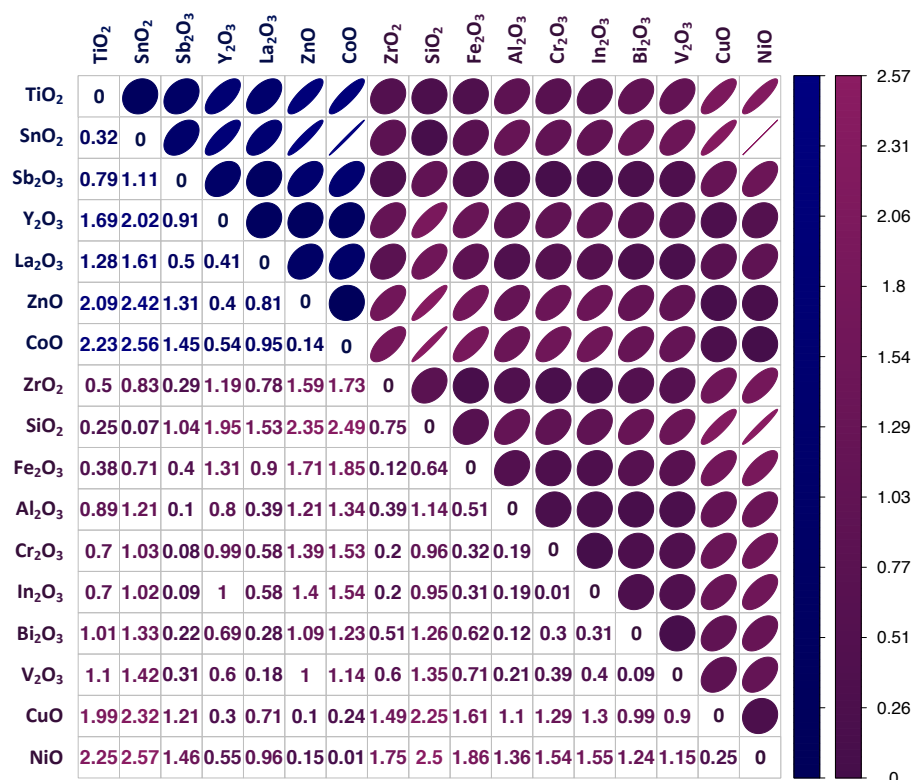


### Case Study 4

**Table S10.** Descriptors used for read-across model development along with experimentally observed and predicted values of nanotoxicity to *E. coli* bacteria for training and validation compounds (MeOx NPs) reported by Gajewicz<sup>[4]</sup>

MeOx	$\Delta H_{Me^+}$ [kcal/mol]	Autoscaled $\Delta H_{Me^+}$	Set*	Experimental $\log(EC_{50})^{-1}$	Predicted $\log(EC_{50})^{-1}$	Residuals	Standardized residuals	Leverages
TiO <sub>2</sub>	1575.73	1.11	T	1.74	2.10	-0.36	-1.56	0.35
SnO <sub>2</sub>	1717.32	1.43	T	2.01	1.81	0.20	1.06	0.49
Sb <sub>2</sub> O <sub>3</sub>	1233.06	0.32	T	2.64	2.56	0.08	0.51	0.16
Y <sub>2</sub> O <sub>3</sub>	837.15	-0.58	T	2.87	3.16	-0.29	-1.24	0.20
La <sub>2</sub> O <sub>3</sub>	1017.22	-0.17	T	2.87	2.78	0.09	0.56	0.15
ZnO	662.44	-0.99	T	3.45	3.45	0.00	0.11	0.30
CoO	601.80	-1.12	T	3.51	3.42	0.09	0.55	0.35
ZrO <sub>2</sub>	1357.66	0.61	V	2.15	2.41	-0.26	-1.12	0.20
SiO <sub>2</sub>	1686.38	1.36	V	2.20	1.99	0.21	1.10	0.45
Fe <sub>2</sub> O <sub>3</sub>	1408.29	0.73	V	2.29	2.16	0.13	0.75	0.23
Al <sub>2</sub> O <sub>3</sub>	1187.83	0.22	V	2.49	2.65	-0.16	-0.65	0.15
Cr <sub>2</sub> O <sub>3</sub>	1268.70	0.41	V	2.51	2.65	-0.14	-0.51	0.17
In <sub>2</sub> O <sub>3</sub>	1271.13	0.41	V	2.81	2.65	0.16	0.89	0.17
Bi <sub>2</sub> O <sub>3</sub>	1137.40	0.10	V	2.82	2.73	0.09	0.53	0.14
V <sub>2</sub> O <sub>3</sub>	1097.73	0.01	V	3.14	2.81	0.33	1.66	0.14
CuO	706.25	-0.89	V	3.20	3.46	-0.26	-1.10	0.27
NiO	596.70	-1.14	V	3.45	3.51	-0.06	-0.16	0.36

\* T – training set; V – validation set;  $\Delta H_{Me^+}$  – the enthalpy of the formation of gaseous cations having the same oxidation state as that in the metal oxide structure.

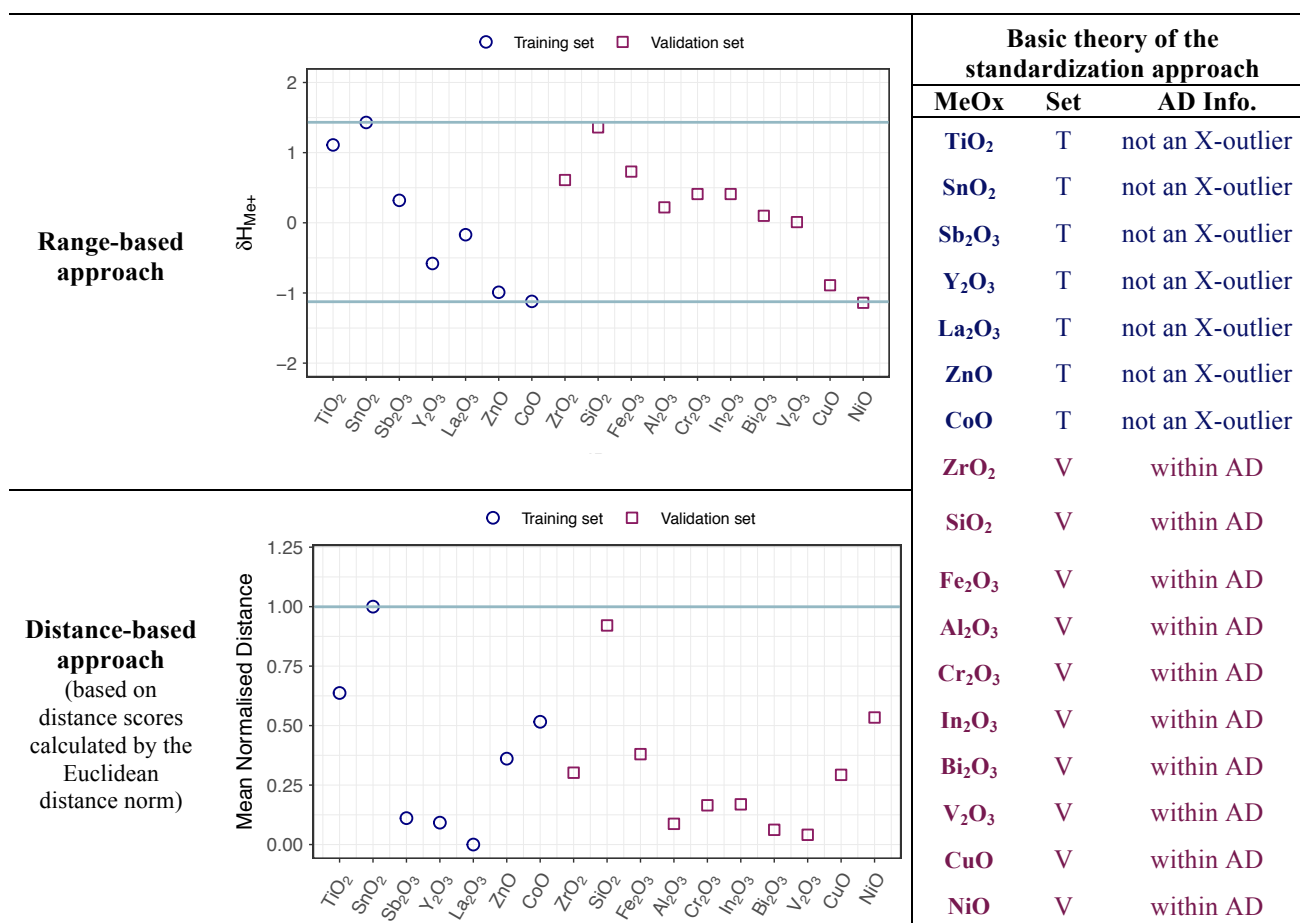


**Figure S4.** Euclidean distance matrix between the training (indicated in navy color) and validation (indicated in maroon color) compounds. The color intensity and shape of the ellipses displayed in the upper triangular part of a distance matrix reflect the values of the corresponding Euclidean distances expressed in numerical forms in the lower triangular part of this matrix. The smallest Euclidean distances (indicated as big circles) are considered as significant. In contrast, the highest distances that reflect the dissimilarity between compounds are considered as insignificant and shown as a thin ellipse.

**Table S11.** Input matrix data for 95% and 99% confidence interval boundaries estimation using the Student's *t*-distribution and developed in-house *R*-code

MeOx	Average Euclidean distance	Standardized residuals	Set
TiO <sub>2</sub>	1.40	-1.56	1
SnO <sub>2</sub>	1.67	1.06	1
Sb <sub>2</sub> O <sub>3</sub>	1.01	0.51	1
Y <sub>2</sub> O <sub>3</sub>	0.99	-1.24	1
La <sub>2</sub> O <sub>3</sub>	0.93	0.56	1
ZnO	1.20	0.11	1
CoO	1.31	0.55	1
ZrO <sub>2</sub>	0.99	-1.12	2
SiO <sub>2</sub>	1.38	1.10	2
Fe <sub>2</sub> O <sub>3</sub>	1.04	0.75	2
Al <sub>2</sub> O <sub>3</sub>	0.85	-0.65	2
Cr <sub>2</sub> O <sub>3</sub>	0.90	-0.51	2
In <sub>2</sub> O <sub>3</sub>	0.90	0.89	2
Bi <sub>2</sub> O <sub>3</sub>	0.83	0.53	2
V <sub>2</sub> O <sub>3</sub>	0.82	1.66	2
CuO	0.98	-1.10	2
NiO	1.14	-0.16	2

**Table S12.** Results on the applicability domain determination using: range-based method, geometrical-based method, distance-based method and standardization approach



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