

THE UNIVERSITY of NORTH CAROLINA at CHAPEL HILL



NTP Workshop on Predictive Models for Acute Oral Systemic Toxicity

NTP Collaboration on Modeling Acute Systemic Oral Toxicity: Results and Ideas

Vinicius M. Alves,^{a,b} Stephen J. Capuzzi,^a Sherif Farag,^a Joyce Borba,^b Arthur Carvalho,^b Carolina H. Andrade,^b <u>Eugene Muratov</u>, ^{a,c} and Alexander Tropsha.^a

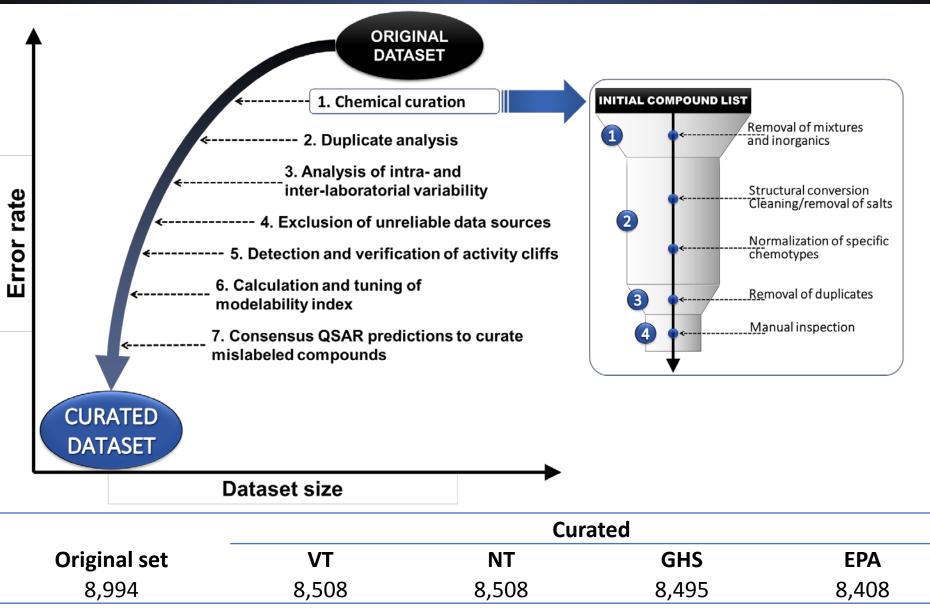
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HOW TO SUCCEED IN COLLABORATION?

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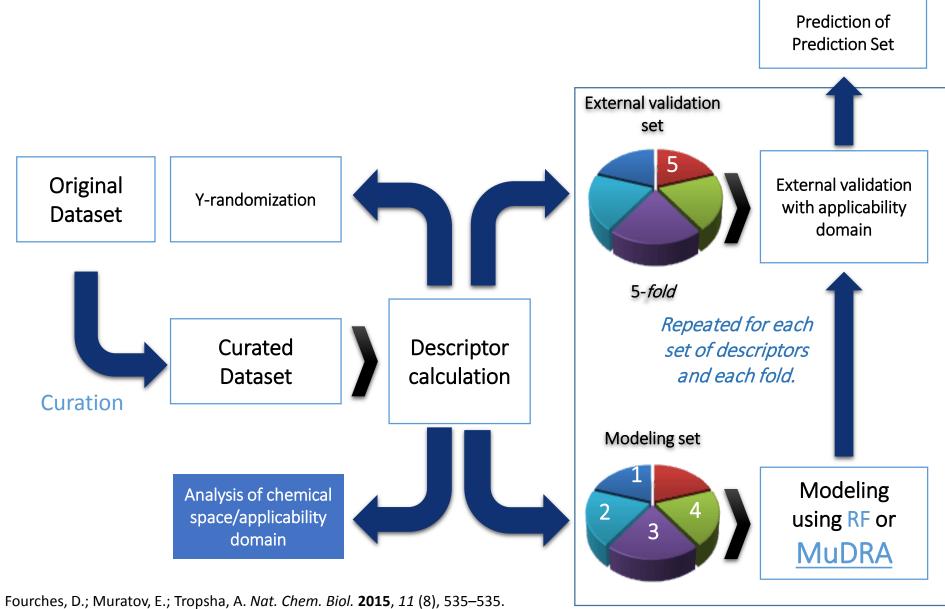
- Understanding of the problem;
- Use of additional data;
- Data curation;
- Rigorous External Validation;
- AD vs. 100% coverage;
- Consensus modeling;
- Experimental validation.

DATA CURATION



Fourches, D.; Muratov, E.; Tropsha, A. J. Chem. Inf. Model. **2016**, *56* (7), 1243–1252. Fourches, D.; Muratov, E.; Tropsha, A. Nat. Chem. Biol. **2015**, *11* (8), 535–535.

GENERAL WORFLOW



Fourches, D.; Muratov, E.; Tropsha, A. *Nat. Chem. Biol.* **2015**, *11* (8), 535–535. Fourches, D.; Muratov, E.; Tropsha, A. *J. Chem. Inf. Model.* **2010**, *50*, 1189–1204 Tropsha, A. *Mol. Inform.* **2010**, *29*, 476–488.

MuDRA

<u>Multi-Descriptor Read Across (MuDRA): a simple</u>

and transparent approach for developing accurate

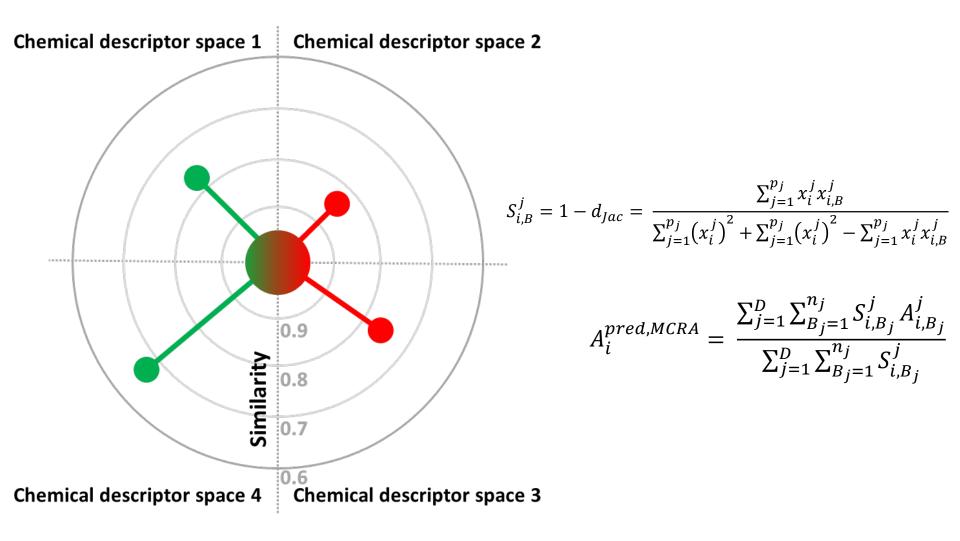
QSAR models

Vinicius M. Alves^{a,b}, Alexander Golbraikh^a, Stephen J. Capuzzi^a, Kammy Liu^c,

Wai In Lam^c, Daniel Robert Korn^c, Diane Pozefsky^c, Carolina Horta Andrade^b,

Eugene N. Muratov^{a,d}*, Alexander Tropsha^{a*}

MuDRA

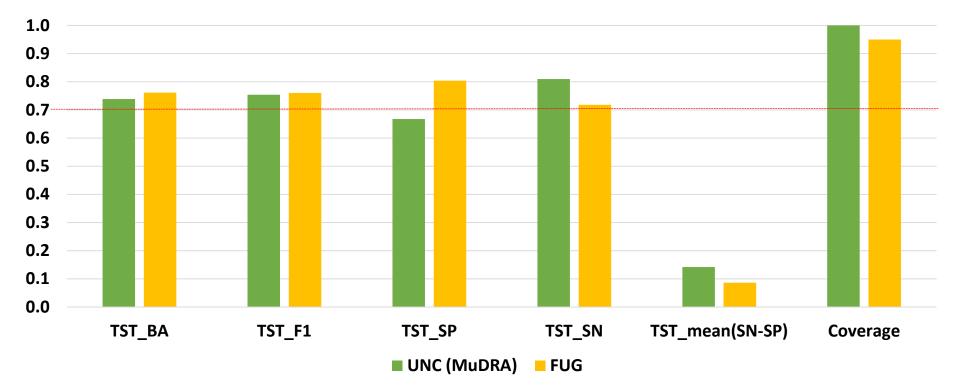


MuDRA vs. CERAPP MODELS

Agonist (n=6,319)				
Model	CCR	Sensitivity	Specificity	
CERAPP (n = 5)	0.73 (± 0.05)	0.51 (± 0.13)	0.95 (± 0.05)	
MuDRA	0.74	0.65	0.83	
Antagonist (n=6,532)				
Model	CCR	Sensitivity	Specificity	
CERAPP (n = 4)	0.53 (± 0.02)	0.11 (± 0.09)	0.95 (± 0.05)	
MuDRA	0.52	0.05	0.99	
Binding (n=7,283)				
Model	CCR	Sensitivity	Specificity	
CERAPP (n = 9)	0.57 (± 0.02)	0.27 (± 0.11)	0.85 (± 0.08)	
MuDRA	0.58	0.35	0.81	

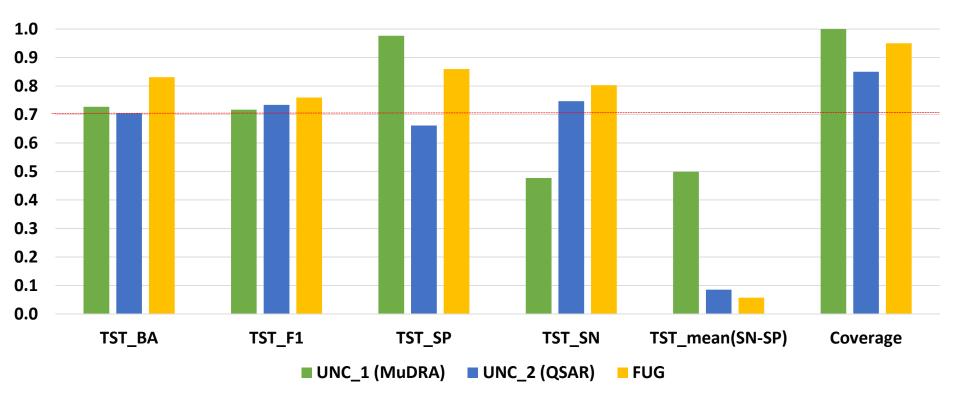
CERAPP compounds are provided by Dr. Mansouri.

NON TOXIC MODELS



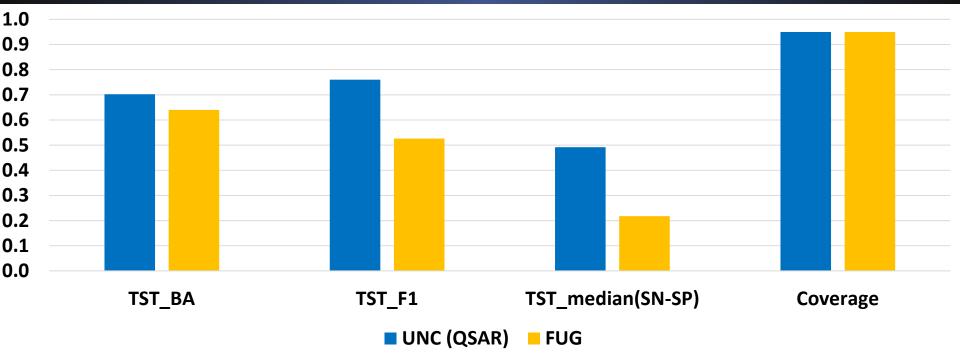
	UNC (MuDRA)	FUG
Descriptors	SiRMS, DRAGON, Morgan, RDKit	MACCS
Algorithm	MuDRA	RANDOM FOREST
No. of compounds in training set	4,834 <mark>toxic</mark> 3,661 not very toxic	2,298 <mark>toxic</mark> 2,298 not very toxic

VERY TOXIC MODELS



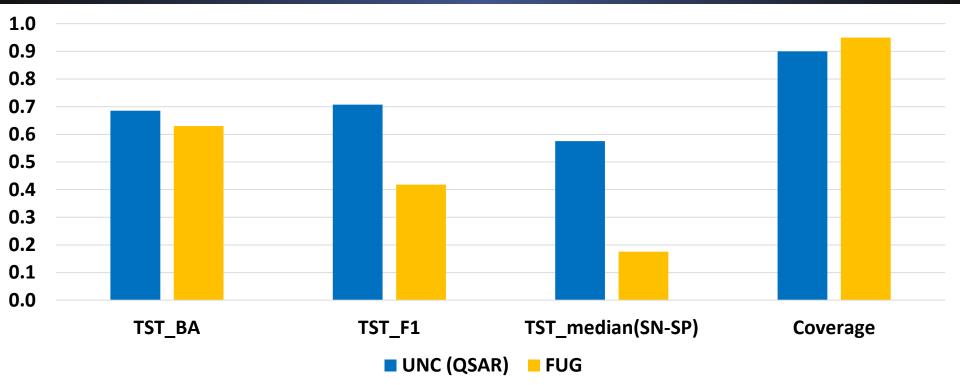
	UNC_1 (MuDRA)	UNC_2 (QSAR)	FUG	
Descriptors	SiRMS, DRAGON, DRAGON Morgan, RDKit		MACCS	
Algorithm	MuDRA	RANDOM FOREST	RANDOM FOREST	
No. of compounds in training set	716 <mark>very toxic</mark> 7,790 not very toxic	385 <mark>very toxic</mark> 385 not very toxic	572 <mark>very toxic</mark> 572 not very toxic	

EPA MODELS



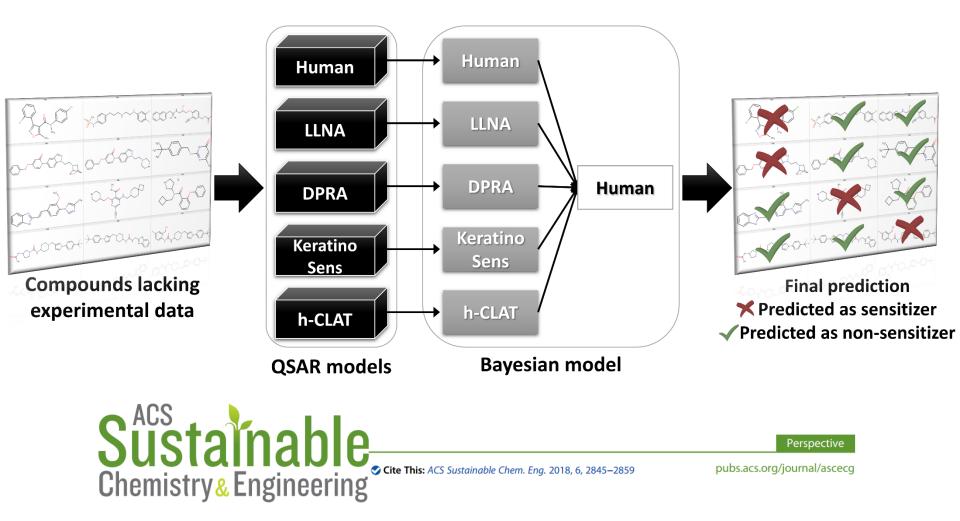
	UNC (QSAR)	FUG
Descriptors	SiRMS, DRAGON, Morgan, RDKit	MACCS
Algorithm	MuDRA	RANDOM FOREST
No. of compounds in	464 extreme (cat. I) 1,099 strong (cat. II) 5.840 moderate (cat. III)	1,048 extreme/strong (cat. I and II) 1,048 moderate (cat. III)
training set	5,840 moderate (cat. III) 1,005 non-toxic (cat. IV)	1,048 non-toxic (cat. IV)

GHS MODELS



	UNC (QSAR)	FUG
Descriptors	SiRMS, DRAGON, Morgan, RDKit	MACCS
Algorithm	MuDRA	RANDOM FOREST
No. of compounds in training set	458 extreme (cat. I) 334 strong (cat. II) 70 moderate (cat. III)	512 extreme/strong (cat. I and II) 512 moderate (cat. III) 512 non-toxic (cat. IV)
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SMART CONSENSUS MODELING



A Perspective and a New Integrated Computational Strategy for Skin Sensitization Assessment

Vinicius M. Alves,^{†,‡}[®] Stephen J. Capuzzi,[†][®] Rodolpho C. Braga,[‡][®] Joyce V. B. Borba,[‡][®] Arthur C. Silva,[‡] Thomas Luechtefeld,[§] Thomas Hartung,[§] Carolina Horta Andrade,[‡][®] Eugene N. Muratov,^{*,†,||}[®] and Alexander Tropsha^{*,†}[®]

- Correct identification and formulation of a problem is a must;
- Use of additional data is extremely helpful;
- Data curation and rigorous external validation is critical;
- MuDRA is a simple, fast, and reliable approach that yields similar accuracy with complex modeling ensembles with 100% coverage of the prediction set;
- We recommend use of AD for single models but 100% coverage for final consensus ensemble;
- Comparison of the accuracy of the models must be made using the same compounds only;
- Building smart consensus model is recommended let the models help each other;
- Only experimental validation could demonstrate predictivity and utility of a model.

ACKNOWLEDGMENT

- Organizers of NTP Collaboration;
- Organizers of the Workshop;
- Kamel Mansouri (NICEATM);
- Nicole Kleinstreuer (NICEATM);
- Alexey Zakharov (NCATS);
- Denis Fourches (NCSU).

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CHEMICAL INFORMATION

Application Note

pubs.acs.org/jcim

Chembench: A Publicly Accessible, Integrated Cheminformatics Portal

Stephen J. Capuzzi,[†] Ian Sang-June Kim,[†] Wai In Lam,[‡] Thomas E. Thornton,[†] Eugene N. Muratov,[†] Diane Pozefsky,^{*,‡} and Alexander Tropsha^{*,†,‡}®

[†]Laboratory for Molecular Modeling, Division of Chemical Biology and Medicinal Chemistry, UNC Eshelman School of Pharmacy, and [‡]Department of Computer Science, University of North Carolina, Chapel Hill, North Carolina 27599, United States



Home	My Bench	Datasets	Modeling	Prediction	MUDRA
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https://chembench.mml.unc.edu/mudra/

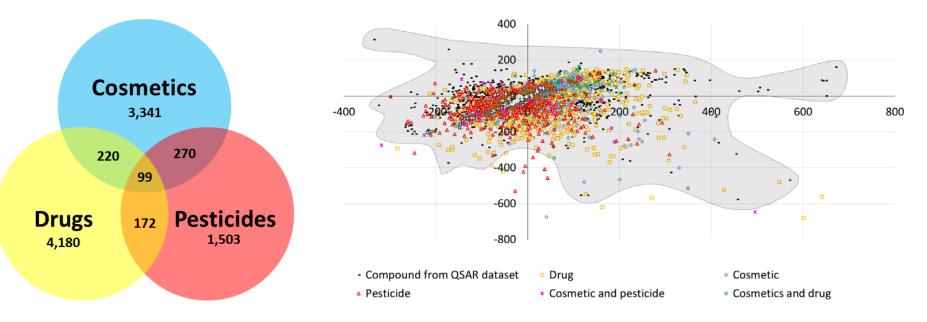
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Food and Chemical Toxicology 112 (2018) 526-534



Chemical toxicity prediction for major classes of industrial chemicals: Is it possible to develop universal models covering cosmetics, drugs, and pesticides?

Vinicius M. Alves ^{a, b}, Eugene N. Muratov ^{a, c}, Alexey Zakharov ^d, Nail N. Muratov ^c, Carolina H. Andrade ^b, Alexander Tropsha ^{a, *}



CrossMark

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Implementation of QSAR models for use of the scientific community

JOURNAL OF CHEMICAL INFORMATION -AND MODELING

Application Note

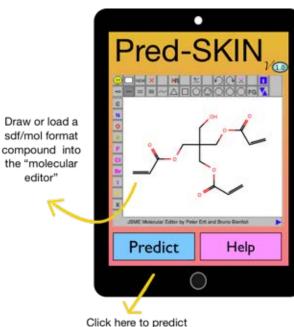
pubs.acs.org/jcim

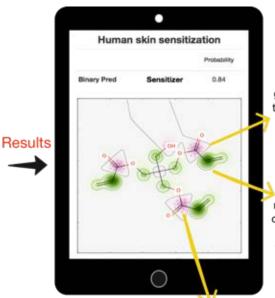
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Pred-Skin: A Fast and Reliable Web Application to Assess Skin Sensitization Effect of Chemicals

Rodolpho C. Braga,^{†,O} Vinicius M. Alves,^{†,‡,O} Eugene N. Muratov,^{‡,§} Judy Strickland,^{||} Nicole Kleinstreuer,[⊥] Alexander Trospsha,[‡] and Carolina Horta Andrade^{*,†}

www.labmol.com.br/predskin





gray lines delimit the region of split between the positive and the negative contributions

green color means a positive contribution of an atom or a fragment to skin sensitization



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pink color means a negative contribution to skin sensitization