Using Deep Learning and Active Learning Methods to Streamline Literature Curation for the ECOTOXicology knowledgebase

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ECOTOX Knowledgebase

Home   Search   Explore   Help   Contact Us

Data last updated

Sept 12, 2019

See update totals

Recent chemicals with full searches and coding completed

Acetochlor	Glyphosate	Prothioconazole
Dichlorobenzenes	HHCB	Simazine
trans-1,2- Dichloroethy…	Metaldehyde	Topramezone
1,2-Dichloropropane	Phthalic anhydride	Uranium
Dicyclohexyl phthalate	Picloram	
Forchlorfenuron	Propazine	

Total in database

11,756	12,906
Chemicals	Species
49,153	952,634
References	Results

WELCOME TO ECOTOX VERSION 5!
Please click here to provide feedback so that we can continue to improve your experience.

About ECOTOX

The ECOTOXicology knowledgebase (ECOTOX) is a comprehensive, publicly available knowledgebase providing single chemical environmental toxicity data on aquatic life, terrestrial plants and wildlife.

Learn More

Getting Started

- Use **Search** if you know exact parameters or search terms (chemical, species, etc.)
- Use **Explore** to see what data may be available in ECOTOX (including data plots)

- **ECOTOX Quick User Guide** (2 pp, 141 K)
- **ECOTOX User Guide** (84 pp, 1120 K)
- **ECOTOX Code Appendix (PDF)** (765 pp, 6447 K, About PDF)

Other Links

- Limitations
- Frequent Questions
- Other Tools/Databases
- Recent Additions

Get Updates via Email

Download

SWIFT-Active Screener

SWIFT-Active Screener is a web-based, collaborative systematic review software application. Active Screener was designed to be easy-to-use, incorporating a simple, but powerful, graphical user interface with rich project status updates. What makes Active Screener special, however, is its behind-the-scenes application of state-of-the-art statistical models designed to save screeners time and effort by automatically prioritizing articles as they are reviewed, using user feedback to push the most relevant articles to the top of the list.

IMPROVED RANKING MODEL
The computer suggests the next articles to screen based on previously included articles.

USER FEEDBACK
User screening decisions are used to continuously refine the machine learning model.

GET ACTIVE SCREENER

To use Active Screener, please contact us at swift-activescreener@sciome.com

If you already have an Active Screener account, click here to access the application.

ACTIVE SCREENER KNOWLEDGE BASE

The SWIFT-Active Screener Knowledge Base is designed to be a resource for users to review key elements of the software and locate responses to specific questions when learning to use the application. To quickly locate a response to a specific question, click on the question in the Knowledge Base index to move to the relevant section of the document.

ACTIVE SCREENER NEWS

Sciome was honored to participate in the **Society of Toxicology 58th Annual Meeting and ToxExpo on March 10-14, 2019 in Baltimore, MD.** Our team's work to create new tools and methods to automate and accelerate systematic review was featured in a number of presentations, including several posters that highlighted SWIFT-Review, SWIFT-Active Screener, and our rapid Evidence Mapping (rEM) efforts. It was a pleasure to meet many of our active users in person and to learn more about your systematic review projects!

Screen Reference

Add New Review

You have reached the predicted inclusion threshold and can stop screening.

Currently Screening: Level 1 - Title & Abstract

62.1%

Inclusion Color
Exclusion Color

2021913: Functionality of sugars: physicochemical interactions in foods

Davis, E. A.; Am J Clin Nutr; 1995

Basic and selected functional properties of s[...]
and maple syrups, honey, and high-fructose[...]
Properties that relate to sweetness and pro[...]
component interaction as a basis for produ[...]
implications of such functionality are illustra[...]
energy foods and for the microwave heating of foods. Among the properties
discussed are solubility, hygroscopicity, crystallinity, and viscosity. Interrelations
among water mobility, water activity, and hydration of proteins, lipids, and
carbohydrates are described in the context of food formulation. Application of
polymer chemistry principles to explain functional properties of amorphous
molecules is reviewed.

Active Screener can reduce required
screening by 50% on most projects with
more than 1,000 references

Main

Notes

Welcome EcoTox User!

👤 brian.howard ⌄

Screen Reference

⊕ Add New Review

Currently Screening: Level 1 - Title & Abstract

1.9%

Inclusion Color
Exclusion Color

3044610: Monte-Carlo-derived insights into dose-kerma-collision kerma inter-relationships for 50 keV-25 MeV photon beams in water, aluminum and copper

Kumar, S., Deshpande, D. D., Nahum, A. E.; Physics in Medicine and Biology; Pg501-519; 2015

Abstract: The relationships between D, K and K-col are of fundamental importance in radiation dosimetry. These relationships are critically influenced by secondary electron transport, which makes Monte- Carlo (MC) simulation indispensable; we have used MC codes DOSRZnrc and FLURZnrc. Computations of the ratios D/K and D/K-col in three materials (water,

Active Screener for EcoTox

Include/Exclude Question

Include this reference? *

◯ Yes, retain the reference for full-text review

⦿ No, exclude the reference from full-text review

Exclusion Reasons

If the reference is excluded, why?

☑ CHEM METHODS

☐ HUMAN HEALTH

☐ FATE

☐ REVIEW

☐ BACTERIA

☐ NON-ENGLISH

Save and Next

Display Instructions

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1.9% ◔

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Active Screener for EcoTox

1. **Improved prioritization with Deep Learning / Transfer Learning**

Save and Next

Display Instructions

Welcome EcoTox User!

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Active Screener for EcoTox

1. Improved prioritization with Deep Learning / Transfer Learning

2. Customized EcoTox Forms

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☐ NON-ENGLISH

Active Screener for EcoTox

1. Improved prioritization with Deep Learning / Transfer Learning

2. Customized EcoTox Forms

3. Automatic Detection of Exclusion Reason

Save and Next

Display Instructions

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1.9%

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Active Screener for EcoTox

1. Improved prioritization with Deep Learning / Transfer Learning

2. Customized EcoTox Forms

3. Automatic Detection of Exclusion Reason

4. Exclusion Reason Keyword Highlighting

Save and Next

Display Instructions

Existing Datasets

Each reference in the libraries was annotated with one of four statuses:

Excluded	65,553
Not Acceptable	3,028
Acceptable	19,181
Unreviewed	1,138
Total:	88,900

Existing Datasets

• Excluded articles also were associated with a reason for exclusion.

• The top 20 reasons make up over 95% of the data. The remaining terms were combined as an "Other" category.

Exclusion Reason	Refs	Percentage
HUMAN HEALTH	19609	30.41%
CHEM METHODS	16745	25.97%
NO TOXICANT	8074	12.52%
FATE	5184	8.04%
BACTERIA	2961	4.59%
REVIEW	2251	3.49%
SURVEY	1696	2.63%
MIXTURE	1101	1.71%
NON-ENGLISH	1003	1.56%
ABSTRACT	939	1.46%
IN VITRO	805	1.25%
OTHER	701	1.09%
……..	……..	……..
BIOLOGICAL TOXICANT	105	0.16%
	64,480	

Deep Learning

ULMFit Classifier (Howard and Ruder, 2018)

(a) LM pre-training   (b) LM fine-tuning   (c) Classifier fine-tuning

Adding Attention to ULMFit

Results: Acceptable / Not Acceptable

Evaluated whether machine learning can be used to classify documents as Acceptable vs Not Acceptable / Excluded and found that:

– Using Active Screener can save users 50% of screening effort for many datasets.

– Augmenting standard model with pretrained model via transfer learning provides additional benefits (mean improvement of 6.5% WSS over the standard Active Screener prioritization model, but several datasets had significantly larger gains).

Results: Exclusion Reason

Label	Total Refs	% Refs	Accuracy	Recall	Precision	F1
CHEM METHODS	7022	27.53%	89.77%	74.09%	77.47%	75.74%
HUMAN HEALTH	4752	18.63%	86.43%	69.70%	61.95%	65.60%
FATE	2875	11.27%	95.56%	69.86%	61.39%	65.35%
REVIEW	1800	7.06%	94.23%	60.78%	62.85%	61.80%
BACTERIA	1359	5.33%	95.58%	54.64%	35.69%	43.18%
NON-ENGLISH	940	3.68%	97.53%	68.33%	74.38%	71.23%
SURVEY	914	3.58%	95.89%	49.80%	59.15%	54.08%
MIXTURE	809	3.17%	97.12%	61.54%	55.03%	58.10%
IN VITRO	805	3.16%	95.73%	47.37%	54.55%	50.70%
ABSTRACT	791	3.10%	98.06%	61.19%	54.67%	57.75%
REFS CHECKED	697	2.73%	97.51%	68.84%	57.93%	62.91%
NO SOURCE	370	1.45%	96.25%	57.84%	64.46%	60.97%
NO CONC	336	1.32%	97.89%	51.49%	46.43%	48.83%
MODELING	284	1.11%	98.78%	63.64%	30.43%	41.18%
NO EFFECT	253	0.99%	97.93%	6.86%	17.95%	9.93%
METHODS	249	0.98%	99.47%	60.00%	68.57%	64.00%
FOOD	220	0.86%	99.03%	35.71%	38.46%	37.04%
PUBL AS	157	0.62%	99.25%	37.84%	41.18%	39.44%
NO DURATION	142	0.56%	99.49%	50.00%	60.00%	54.55%
YEAST	111	0.44%	99.55%	70.59%	83.72%	76.60%
OTHER	625	2.44%	95.76%	10.73%	27.85%	15.49%

Adding Attention (IMDB Example)

Positive Review

184	legend	1.77E-05		184	...	0.008636
185	,	0.000843		185	which	0.00083
186	and	0.001618		186	made	0.001627
187	the	0.000135		187	laugh	0.005802
188	director	0.001799		188	the	1.83E-05
189	plays	0.000435		189	whole	0.000199
190	on	0.000696		190	theater	0.000761
191	this	0.017107		191	tk_rep	0.003114
192	with	0.003877		192	4	0.01618
193	the	0.000181		193	.	0.00739
194	style	0.008258		194	this	0.002407
195	and	0.003755		195	movie	0.004775
196	pace	0.015138		196	is	0.001468
197	of	0.006156		197	a	0.000584
198	the	0.000833		198	must	0.044615
199	action	0.01235		199	see	0.057459
200	,	0.003604		200	for	0.026079
201	making	0.003004		201	everyone	0.034127
202	it	0.002469		202	!	0.017458
203	more	0.00192				

Negative Review

73	for	0.001087		73	effect	0.000544
74	a	0.000145		74	on	2.13E-05
75	proper	0.000367		75	palestinians	2.85E-05
76	translation	0.001085		76	,	8.99E-05
77	,	0.000924		77	and	0.000252
78	so	0.000231		78	consider	0.009548
79	they	2.48E-05		79	it	0.00769
80	decided	0.001516		80	unnecessarily	0.001456
81	to	0.000219		81	divisive	0.00026
82	_unk_	0.000296		82	and	0.001019
83	it	0.002551		83	/	6.69E-05
84	.	0.004328		84	or	2.91E-05
85	with	0.001731		85	a	2.92E-05
86	sometimes	0.000357		86	waste	0.064926
87	hilarious	0.021297		87	of	0.060844
88	results	0.001107		88	money	0.064973
89	.	0.001372		89	.	0.018019
90		0.004582		90	oh	0.005169
91	do	0.000368		91	yes	0.002627
92	n't	0.003227		92	,	0.001174

Summary

- Standard Active Screener application saves users 50% screening time

- EcoTox Active Screener uses Deep Learning to:

  - Save an additional 6.5+% screening time

  - Accurately predict exclusion reasons

  - Explain its predictions using attention-highlighting

- Models will continue to improve with more data, and several methodological enhancements are planned

Next Steps

Phase II of the project…

- **Aim 2.1**: **Additional refinements to machine learning models** that can be used to automatically identify, with high precision, those references that can be deemed **non-acceptable/non-applicable** for the EcoTox database, and to **categorize excluded references** according to a selection from a list of pre-defined rationales.

- **Aim 2.2**: Modify **Active Screener** to **operationalize the above models** and to better serve EcoTox data curation pipeline.

- **Aim 2**.3: **Publish** results in a suitable journal or conference.

- **Aim 2.4**: Investigate feasibility of developing models to extract the approximately 4,000 Effects Groups and Measurement Codes from full-text documents.

∫Ciome

WHAT WE DO   WHO WE ARE   WHO WE SERVE   SOFTWARE   PUBLICATIONS   NEWS   CAREERS   CONTACT

Bioinformatics

- ✓ Next-Generation Sequence data analysis
- ✓ Microarray data analysis
- ✓ Structural & Functional genomics
- ✓ SNP/Genotype analysis & GWAS
- ✓ Biostatistics and Mathematical Modeling

Cheminformatics

- ✓ Quantitative Structure-Activity Relationship (QSAR) modeling
- ✓ Computational Toxicity Predictions
- ✓ Active site and Protein-Protein Docking
- ✓ Pharmacophore Modeling

Text-Mining and Literature Review

- ✓ Document Tagging and Visualization
- ✓ Full-Text Conversion and Search
- ✓ Document Clustering, Ranking & Classification
- ✓ Literature Prioritization and Screening
- ✓ Data extraction
- ✓ rapid Evidence Mapping (rEM) and systematic reviews
- ✓ Web mining and information retrieval

Data Science and Analytics

- ✓ Integration and visualization of large volumes of heterogeneous data
- ✓ Development and implementation of Deep Learning methodologies for predictive science
- ✓ Automated Image analysis using artificial intelligence
- ✓ Natural Language Processing (NLP) methods using Deep Learning

Software Development

- ✓ Requirements gathering
- ✓ Software architecture design
- ✓ User interface design
- ✓ Implementation, deployment
- ✓ User support

More info about Sciome and Active Screener at our website:

www.sciome.com