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

# **Brian E Howard (Sciome)**

Ruchir Shah (Sciome)
Jennifer Olker (EPA)
Colleen Elonen (EPA)
Dale Hoff (EPA)





**Environmental Topics Laws & Regulations About EPA** Search EPA.gov **ECOTOX** Knowledgebase Home Search **Explore** Help Contact Us Total in database Recent chemicals with full searches and coding completed Data last updated 11,756 12,906 Acetochlor Glyphosate Prothioconazole Sept 12, Dichlorobenzenes ННСВ Simazine Chemicals **Species** 2019 trans-1,2- Dichloroethy... Metaldehyde Topramezone 1,2-Dichloropropane Phthalic anhydride Uranium Dicyclohexyl phthalate Picloram 49,153 952,634 See update totals Forchlorfenuron Propazine Results References

### **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



# **ECOTOX Pipeline**

Develop literature search

Conduct searches

Identify potentially applicable studies

Acquire potentially applicable studies

Apply ECOTOX applicability criteria

Code Data into ECOTOX

Systematic Review and Data Curation









**USER** 

**FEEDBACK** 

### **SWIFT-Active Screener**

# SWIFT-ACTIVESCREENER

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.

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

### **IMPROVED RANKING MODEL**

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

### **GET ACTIVE SCREENER**

To use Active Screener, please contact us at swiftactivescreener@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!



Inclusion Color

Exclusion Color

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

Currently Screening: Level 1 - Title & Abstract

62.1%

2021913: Functionality of sugars: physicochemical interactions in foods

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

Basic and selected functional properties of sand maple syrups, honey, and high-fructose Properties that relate to sweetness and proposed component interaction as a basis for productional properties of such functionality are illustrated.

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

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.

Main

Notes

• Add New Review

Currently Screening: Level 1 - Title & Abstract — Inclusion Color 1.9% 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 Include/Exclude Question Abstract: The relationships between D, K and K-col are of fundamental importance in radiation dosimetry. These relationships are critically Include this reference? \* influenced by secondary electron transport, which makes Monte- Carlo (MC) Yes, retain the reference for full-text review simulation indispensable; we have used MC codes DOSRZnrc and FLURZnrc. No, exclude the reference from full-text review Computations of the ratios D/K and D/K-col in three materials (water, \*\*Active Screener for EcoTox\*\* **Exclusion Reasons** If the reference is excluded, why? CHEM METHODS HUMAN HEALTH FATE REVIEW BACTERIA NON-ENGLISH **Display Instructions** 

• Add New Review

Currently Screening: Level 1 - Title & Abstract Inclusion Color 1.9% 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 Include/Exclude Question Abstract: The relationships between D, K and K-col are of fundamental importance in radiation dosimetry. These relationships are critically Include this reference? \* influenced by secondary electron transport, which makes Monte- Carlo (MC) Yes, retain the reference for full-text review simulation indispensable; we have used MC codes DOSRZnrc and FLURZnrc. No, exclude the reference from full-text review Computations of the ratios D/K and D/K-col in three materials (water, \*\*Active Screener for EcoTox\*\* Improved prioritization with Deep **Exclusion Reasons** Learning / Transfer Learning If the reference is excluded, why? CHEM METHODS HUMAN HEALTH FATE REVIEW BACTERIA NON-ENGLISH **Display Instructions** 

SWIFT ACTIVESCREENER

• Add New Review

Currently Screening: Level 1 - Title & Abstract Inclusion Color 1.9% 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 Include/Exclude Question Abstract: The relationships between D, K and K-col are of fundamental importance in radiation dosimetry. These relationships are critically Include this reference? \* influenced by secondary electron transport, which makes Monte- Carlo (MC) Yes, retain the reference for full-text review simulation indispensable; we have used MC codes DOSRZnrc and FLURZnrc. No, exclude the reference from full-text review Computations of the ratios D/K and D/K-col in three materials (water, \*\*Active Screener for EcoTox\*\* Improved prioritization with Deep **Exclusion Reasons** Learning / Transfer Learning If the reference is excluded, why? 2. Customized EcoTox Forms CHEM METHODS HUMAN HEALTH FATE REVIEW BACTERIA NON-ENGLISH **Display Instructions** 

SWIFT ACTIVESCREENER

• Add New Review

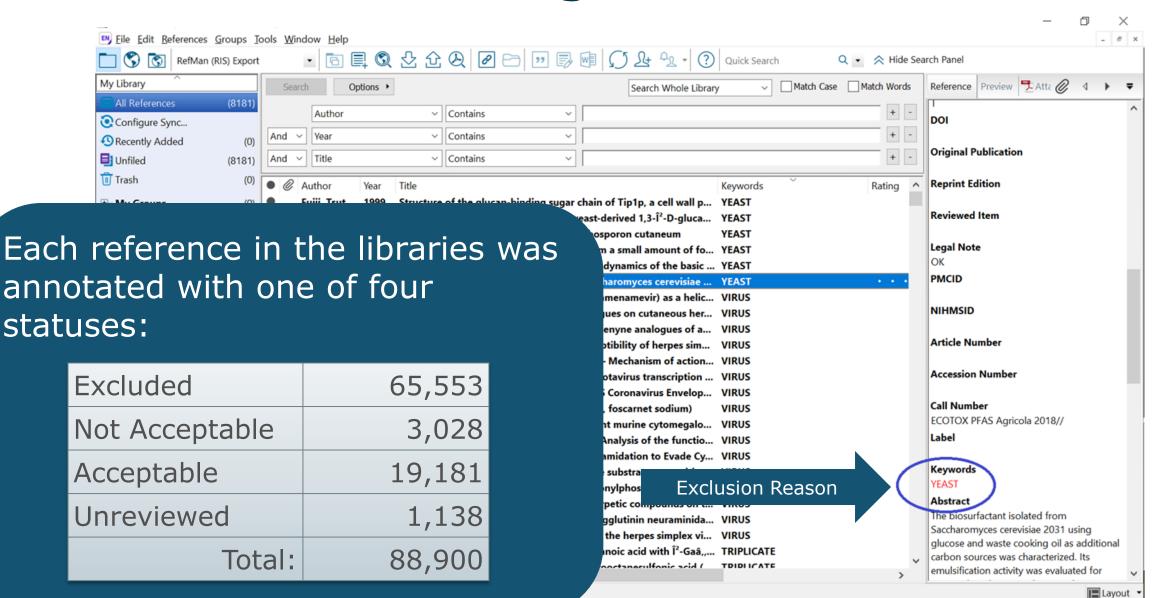
Currently Screening: Level 1 - Title & Abstract Inclusion Color 1.9% Exclusion Color 3044610: Monte-Carlo-derived insights into dose-kerma-collision kerma inter-relationships for 50 keV-25 MeV photon beams in water, aluminum and copper Kumar, S., Deshpande, D. D., Nahum, A. E.; Physics in Medicine and Biology; Pg501-519; 2015 Include/Exclude Question Abstract: The relationships between D, K and K-col are of fundamental importance in radiation dosimetry. These relationships are critically Include this reference? \* influenced by secondary electron transport, which makes Monte- Carlo (MC) Yes, retain the reference for full-text review simulation indispensable; we have used MC codes DOSRZnrc and FLURZnrc. No, exclude the reference from full-text review. Computations of the ratios D/K and D/K-col in three materials (water, \*\*Active Screener for EcoTox\*\* Improved prioritization with Deep **Exclusion Reasons** Learning / Transfer Learning If the reference is excluded, why? 2. Customized EcoTox Forms CHEM METHODS HUMAN HEALTH FATE 3. Automatic Detection of Exclusion REVIEW Reason BACTERIA NON-ENGLISH **Display Instructions** 

SWIFT ACTIVESCREENER



Currently Screening: Level 1 - Title & Abstract Inclusion Color 1.9% 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 Include/Exclude Question Abstract: The relationships between D, K and K-col are of fundamental importance in radiation dosimetry. These relationships are critically Include this reference? \* influenced by secondary electron transport, which makes Monte- Carlo (MC) -Yes, retain the reference for full-text review simulation in tispensable; we have used MC codes DOSRZnrc and FLURZnrc. No, exclude the reference from full-text review. Computations of the ratios D/K and D/K-col in three materials (water, \*\*Active Screener for EcoTox\*\* Improved prioritization with Deep **Exclusion Reasons** Learning / Transfer Learning If the reference is excluded, why? 2. Customized EcoTox Forms CHEM METHODS HUMAN HEALTH FATE 3. Automatic Detection of Exclusion REVIEW Reason BACTERIA NON-ENGLISH 4. Exclusion Reason Keyword Highlighting **Display Instructions** 

# **Existing Datasets**



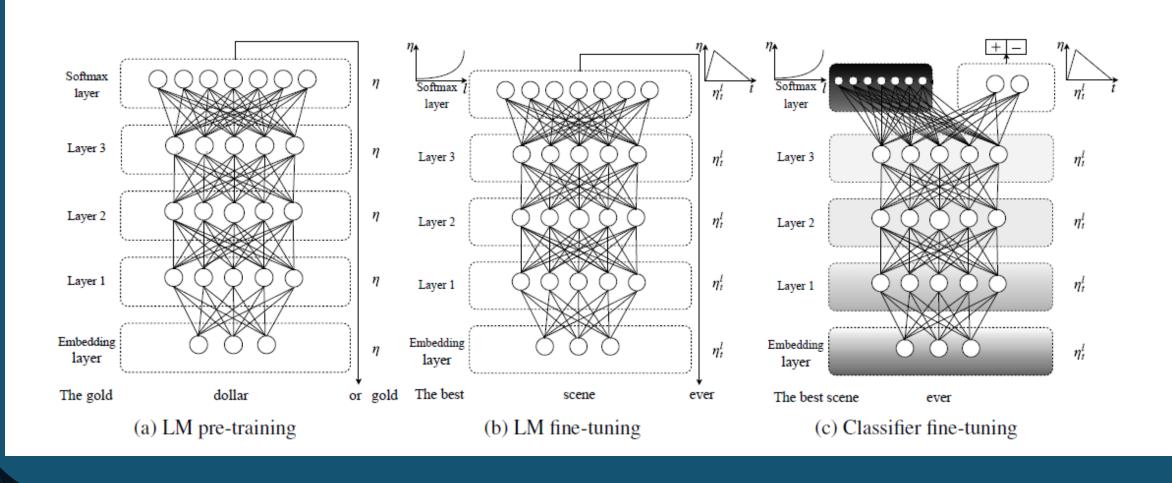
# **Existing Datasets**

- •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.

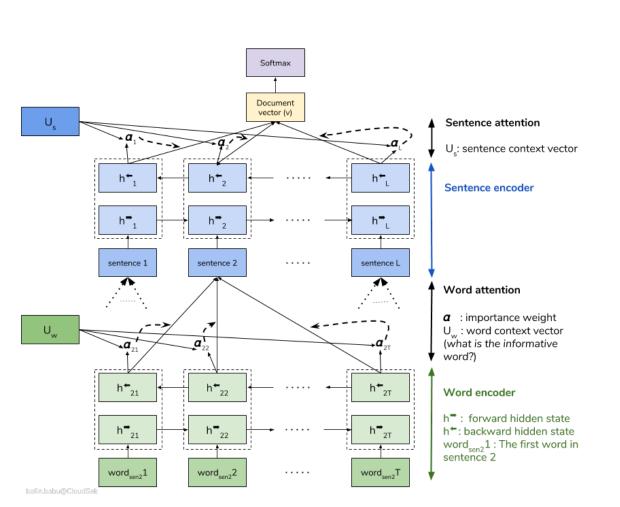
<b>Exclusion Reason</b>	Refs	Percentage
HUMAN HEALTH	19609	30.41%
CHEM METHODS	16745	25.97%
NO TOXICANT	8074	12.52%
FATE	5184	8.04%
BACTERIA	2961	4.59%
REVIEW	2251	3.49%
SURVEY	1696	2.63%
MIXTURE	1101	1.71%
NON-ENGLISH	1003	1.56%
ABSTRACT	939	1.46%
IN VITRO	805	1.25%
OTHER	701	1.09%
BIOLOGICAL TOXICANT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

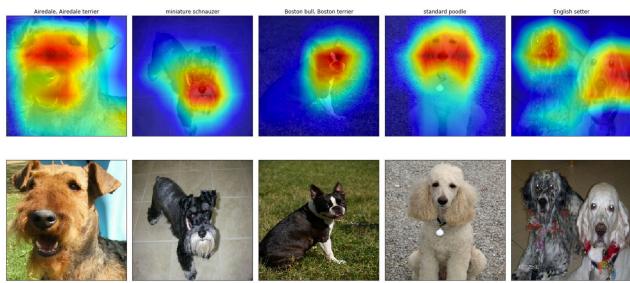
# Deep Learning

# **ULMFit Classifier (Howard and Ruder, 2018)**



# **Adding Attention to ULMFit**

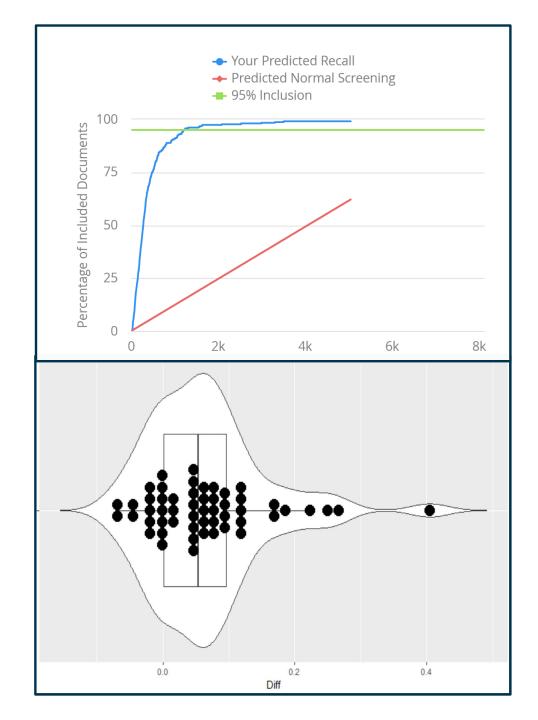




# Results: Acceptable / Not Acceptable

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	<b>Total Refs</b>	% Refs	Accuracy	Recall	<b>Precision</b>	F1
CHEM METHODS	7022	<mark>27.53%</mark>	<mark>89.77%</mark>	<mark>74.09%</mark>	<mark>77.47%</mark>	<mark>75.74%</mark>
HUMAN HEALTH	<mark>4752</mark>	18.63%	<mark>86.43%</mark>	<mark>69.70%</mark>	<mark>61.95%</mark>	<mark>65.60%</mark>
FATE	<b>2875</b>	11.27%	<mark>95.56%</mark>	<mark>69.86%</mark>	<mark>61.39%</mark>	<mark>65.35%</mark>
REVIEW	1800	<mark>7.06%</mark>	<mark>94.23%</mark>	<mark>60.78%</mark>	62.85%	<mark>61.80%</mark>
BACTERIA	<b>1359</b>	<b>5.33%</b>	<mark>95.58%</mark>	<mark>54.64%</mark>	<mark>35.69%</mark>	<mark>43.18%</mark>
NON-ENGLISH	940	3.68%	<mark>97.53%</mark>	<mark>68.33%</mark>	<mark>74.38%</mark>	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	а	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			90	oh	0.005169
		0.004582			
91	do	0.000368	91	yes	0.002627
92	<u>n't</u>	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 predefined 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.







HAT WE DO WH

RE WHO WE SE

SOFTWARE

PUBLICATIONS

S CARE

CON

### **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

# QUESTIONS