# Estimating the uncertainty of AI predictions of the Kováts retention index Lewis Y. Geer, Douglas J. Slotta, and Stephen E. Stein

## INTRODUCTION

The Kováts Retention Index (RI) is a molecular property widely used to help identify chemicals.

The NIST/EPA/NIH EI-MS spectral library includes measured and estimated RIs. Estimated RIs are predicted using a deep neural network called **AIRI** (Geer, Stein & Wallace, ASMS 2019).

AIRI takes a molecular structure as input and outputs a predicted RI. Previous AIRI predictions used a single fixed uncertainty calculated over the entire test set.

We investigate the use of an ensemble of networks to estimate a per structure uncertainty with the goal of improving the overall average confidence limits of the **AIRI** predictions.

## METHOD



The models were trained with an AdamW optimizer and a mean average error loss. The software libraries used are Masskit (https://github.com/usnistgov/masskit) for computation on spectra and structures, and Masskit AI (https://github.com/usnistgov/masskit\_ai), used for AI computation on spectra and structures.

#### Data sets

The models were trained on structures from the NIST 2020 library with experimentally measured RIs for standard semi-nonpolar columns.

The test set consisted of structures from the NIST 2023 library with experimentally measured RIs and with a different connectivity than any structure in the training set.

### **Uncertainty calculation**

To create uncertainties for each prediction, RI values were predicted for each structure using the 8 models trained. These values were used to generate an average and standard deviation for each RI value.

It was observed that the uncertainties for some predictions were too small. To compensate for this, a minimum error of 22 was used as it resulted in the lowest average value at the 90% confidence interval.

ransformer Model							
	Attention						
	Atom #	1	2	3			
	1		0.5	0.5			
	2	0.5		0.25			
	3	0.5	0.25				
×							



## RESULTS

The prediction of per RI value uncertainties resulted in an *improvement in the mean 50% confidence limit* from 27 to 22 and of the mean 90% confidence limit from 102 to 70.

A selection of test structures with experimentally measured RIs and predicted RIs with error bars:



## CONCLUSIONS

The use of model ensembles results in a better estimation of uncertainty.

Incorporating AIRI predictions in the NIST MS Search reduces false positives by the following method:

- All library structures have a measured or predicted RI.
- If the query spectrum of an unknown chemical has an RI, a match factor is computed for each hit that incorporates the difference of the query RI and the library RI, which can be measured or predicted.
- are performed on a spectrum of *o-Xylene*, one using RI and one not:

	#	Lib	MF RM	F Prol	o RI	DBs	Name		
No AIRI	<b>⊞ 1</b>	Μ	941	942	29.7	865	32	p-Xyle	ne
	<b>⊞ 2</b>	Μ	937	939	26.5	866	32	Benze	ne, 1,3-dimethy
		<u>M</u>	<u>924</u>	<u>925</u>	<u>18.0</u>	<u>888</u>	<u>36</u>	<u>o-Xyle</u>	ne
	<b>⊕ 4</b>	Μ	888	895	6.09	855	35	Ethylb	enzene
	<b>⊞</b> 5	Μ	878	881	4.55	121		Benze	neethanol, α,β-
	#	Lib	MF RN	ЛFPro	ob RI	origMF	DBs	Name	
With AIRI	<b>⊕</b> 1	Μ	924	925	90.6	888	924	36	o-Xylene
	2	M	826	877	6.70	878*	876		Cyclopentene
	<b>⊞</b> 3	M	767	939	1.22	866	937	32	Benzene, 1,3
	<b>⊕ 4</b>	M	761	<b>942</b>	1.03	865	941	32	p-Xylene
	<b>⊞</b> 5	Μ	699	750	0.17	898	749	1 G	1,3,5-Cyclood





• The resulting match factor can help suppress false positives. For example, the following two searches

