

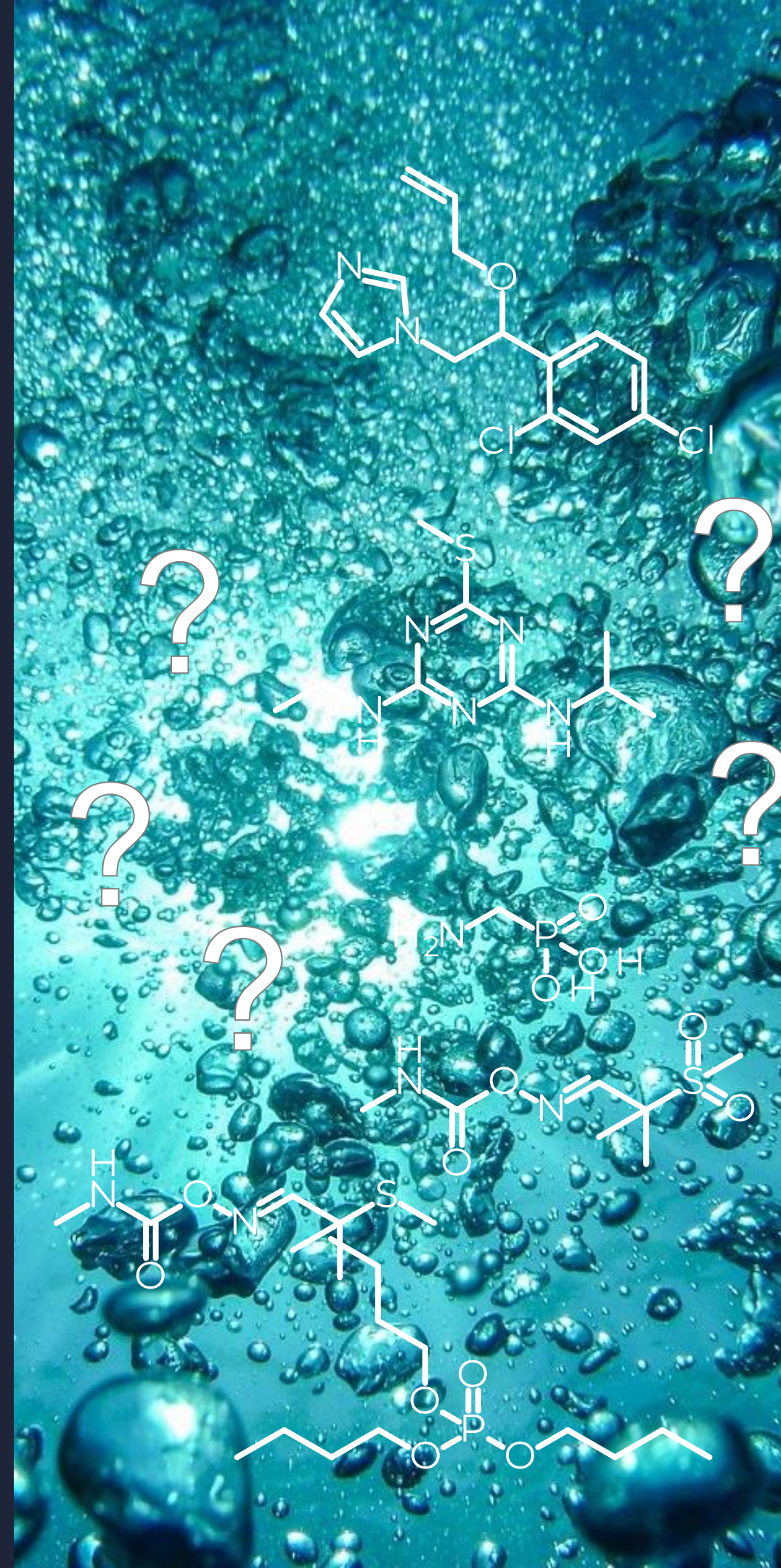
**prioritizing & elucidating the
structure of toxic chemicals
based on complementary
empirical analytical
information from LC/HRMS**

anneli kruve

Kruvelab.com

water analysis

thousands of chemicals
detected
in the environment



how to ...



PRIORITIZE

risk



IDENTIFY

structure

how to ...

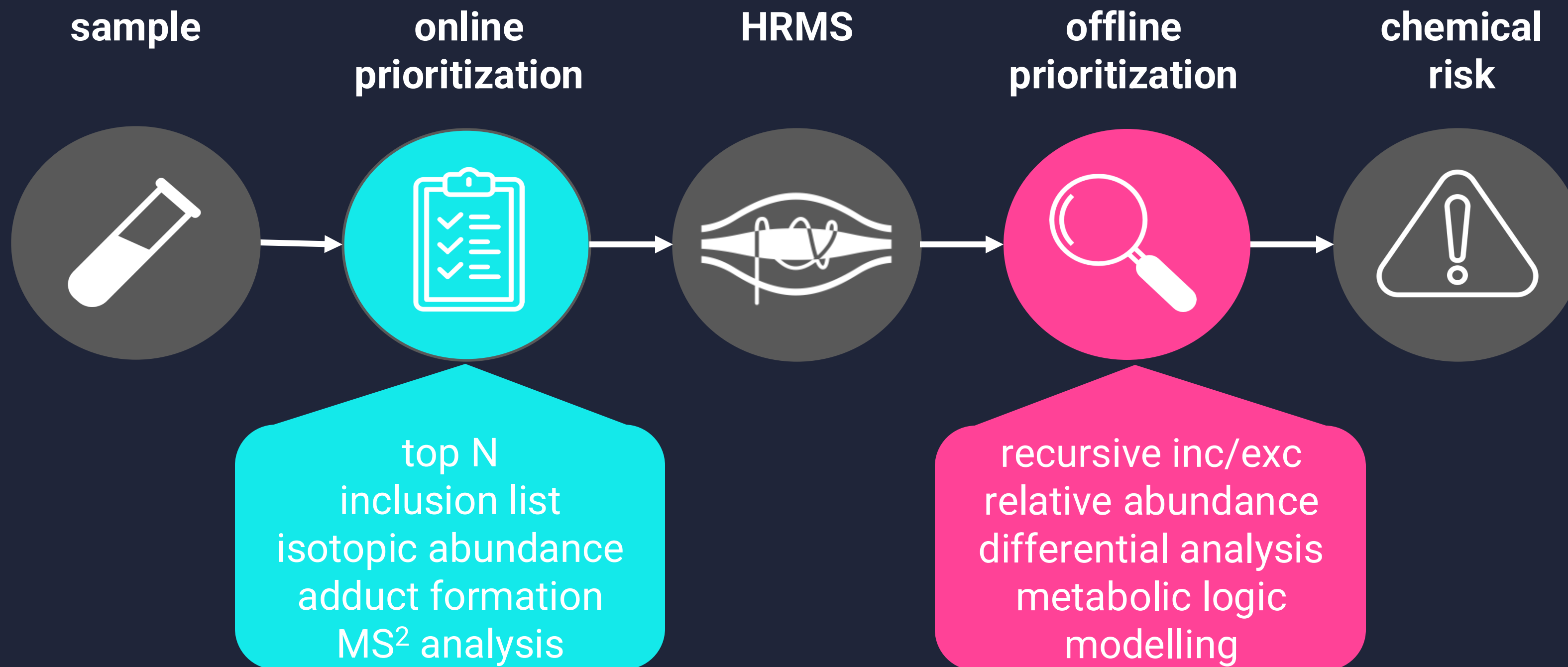


PRIORITIZE

risk

prioritization

Szabo et al. Anal Chem 2024



prioritization of chemicals



TOXICITY

ecotoxicity and endocrine
disruptors



CONCENTRATION

exposure to potentially toxic
chemicals



RISK

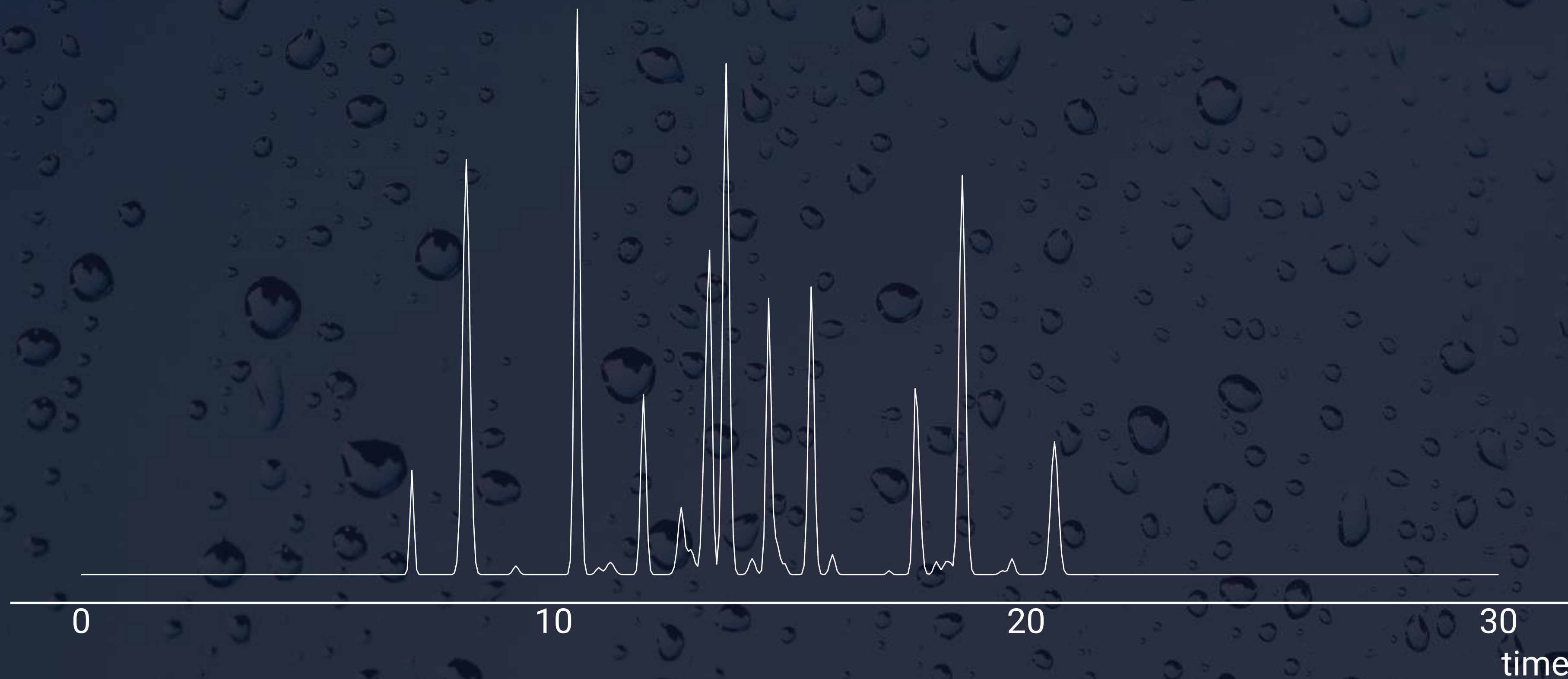
$$\text{PriorityScore} = \frac{C_{\text{predicted}}}{AC_{50}^{\text{5th percentile}}}$$

toxicity

of detected chemicals

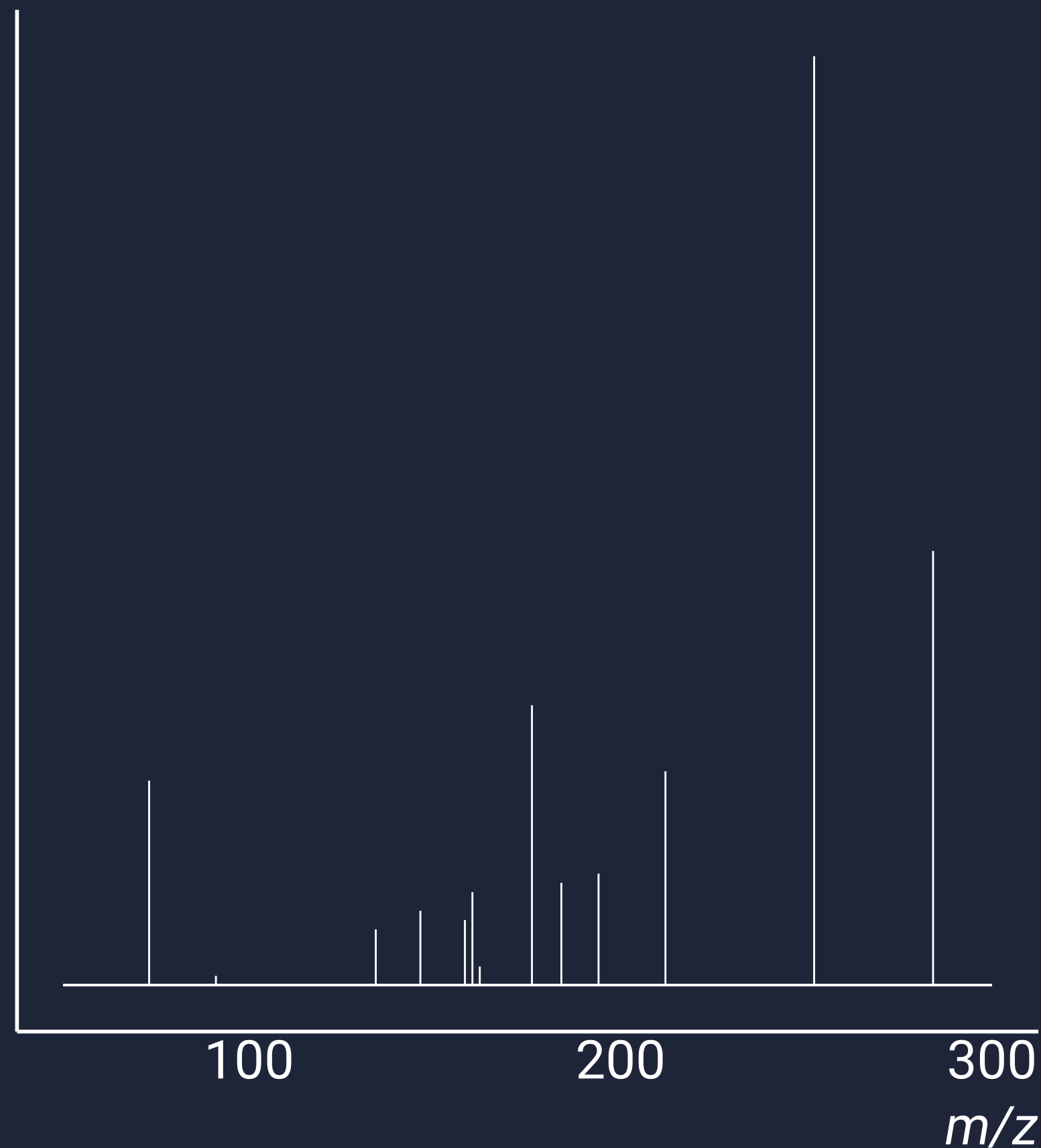
nontarget screening

with LC/HRMS

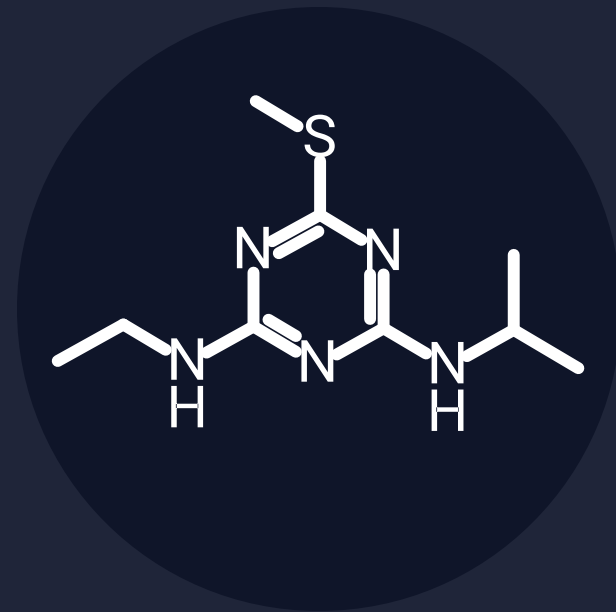


toxicity assessment

from spectra
to structure
to toxicity

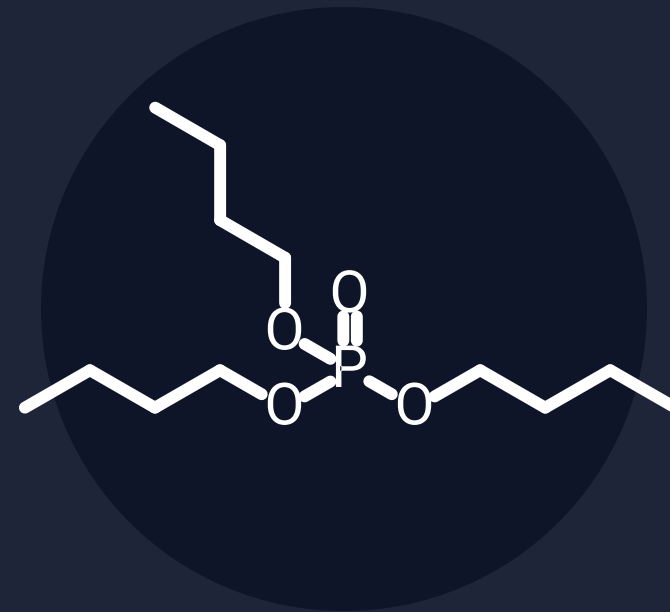


toxicity assessment



LC₅₀ = 9.3 mg/L

known structure
known toxicity



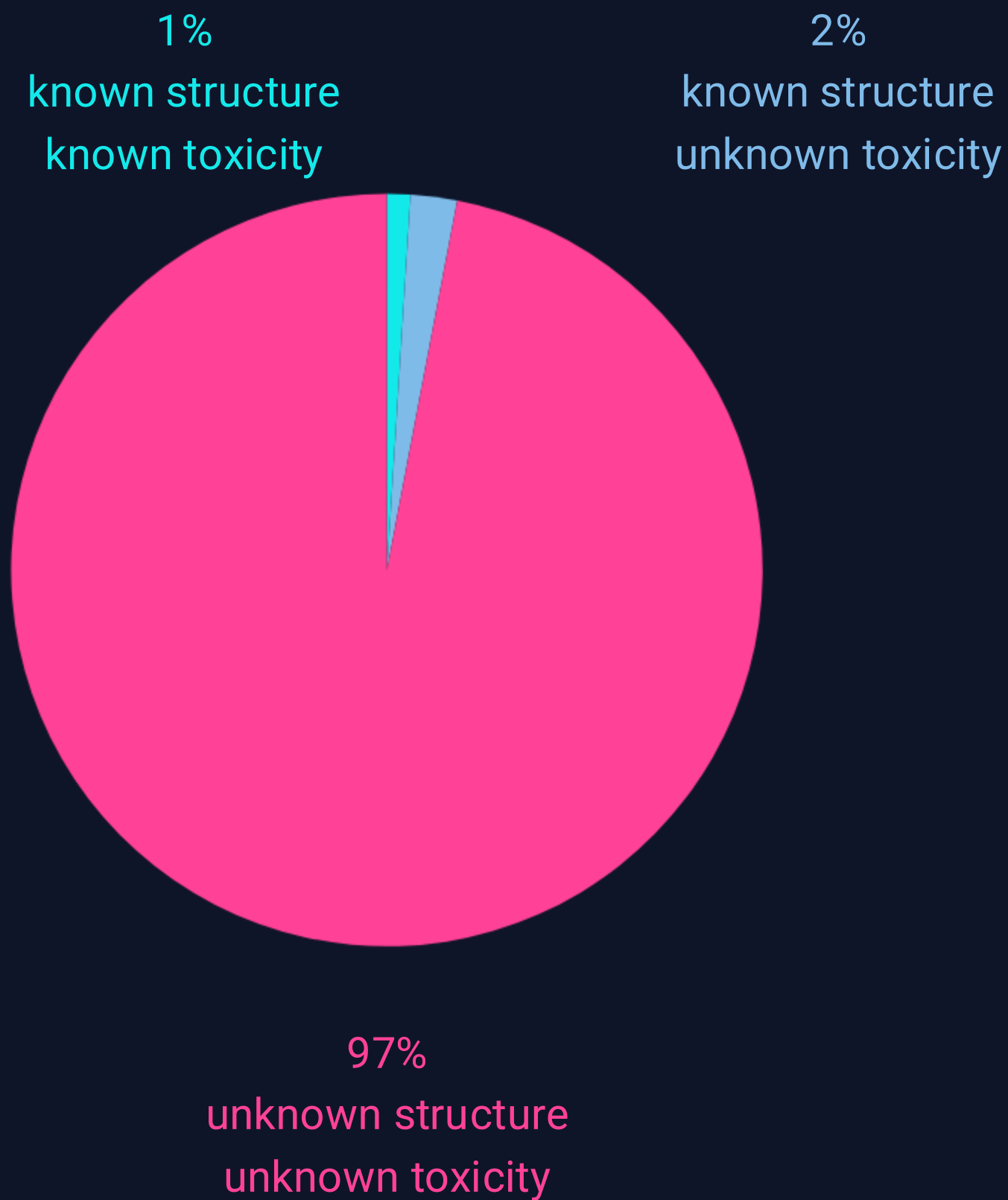
LC₅₀ = ? mg/L

known structure
unknown toxicity



LC₅₀ = ? mg/L

unknown structure
unknown toxicity



toxicity assessment

vast majority of detected
chemicals remain
unknown

predicting toxicity

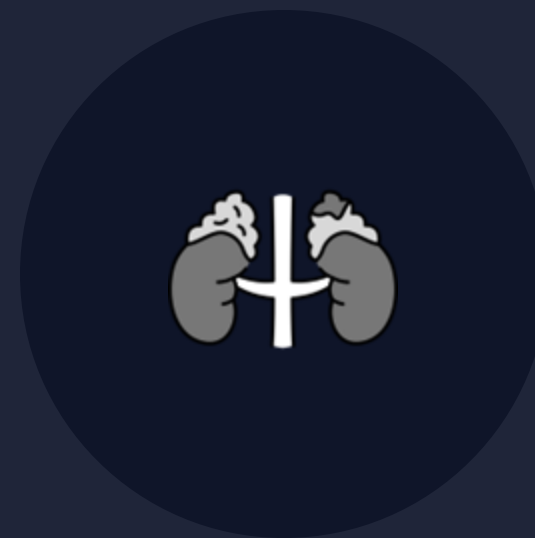
of detected chemicals

endpoints



ECOTOXICITY

fathead minnow, bluegill, and
rainbow trout



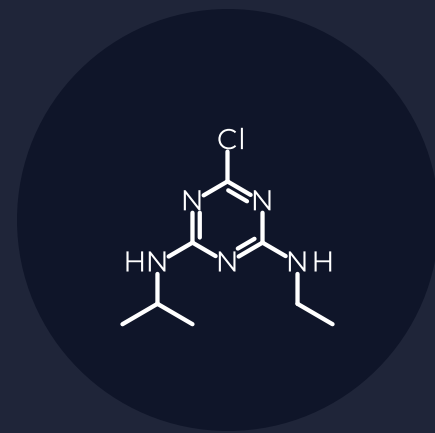
ENDOCRINE DISRUPTION

AhR, AR, ER, MMP, P53, ...

workflow



MS² SPECTRA



STRUCTURE AS SMILES



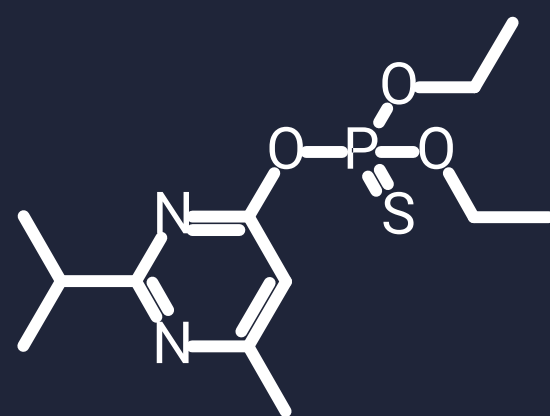
**MOLECULAR
DESCRIPTORS**



PREDICT TOXICITY

structural fingerprints

Peets et al. ES&T 2022



R:
rcdk



0	
1	
1	
0	
1	

model training

Peets et al. ES&T 2022

mass (Da)	fp1	...	fp243
317.32000	0	...	0
208.26100	1	...	0
240.21499	1	...	0
300.57998	0	...	0
201.22500	0	...	0

model training

Peets et al. ES&T 2022

mass (Da)	fp1	...	fp243
317.32000	0	...	0
208.26100	1	...	0
240.21499	1	...	0
300.57998	0	...	0
201.22500	0	...	0

training set
517
chemicals

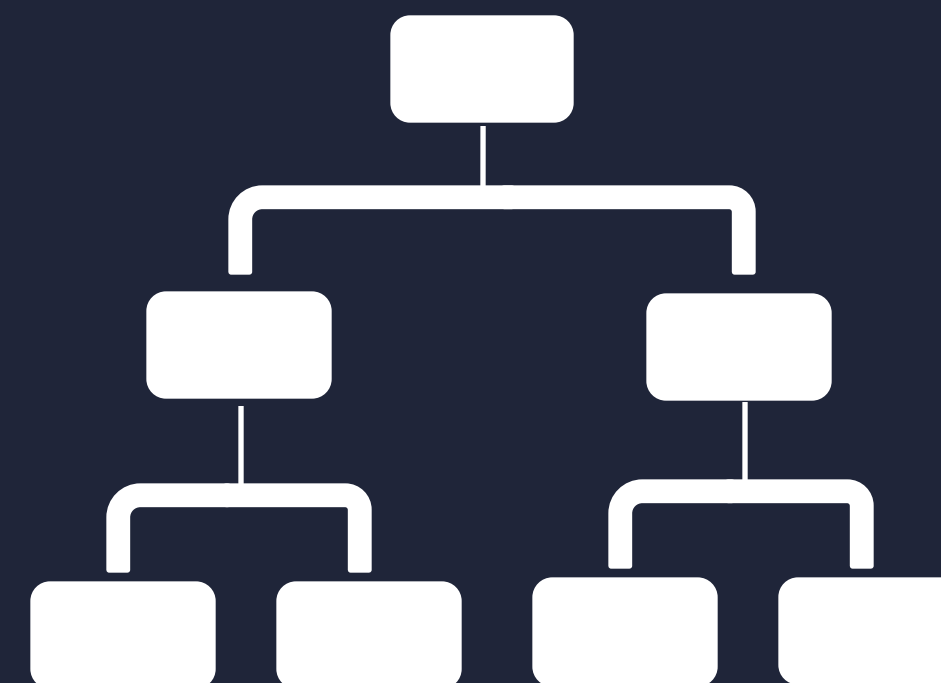
test set
130
chemicals

model training

Peets et al. ES&T 2022

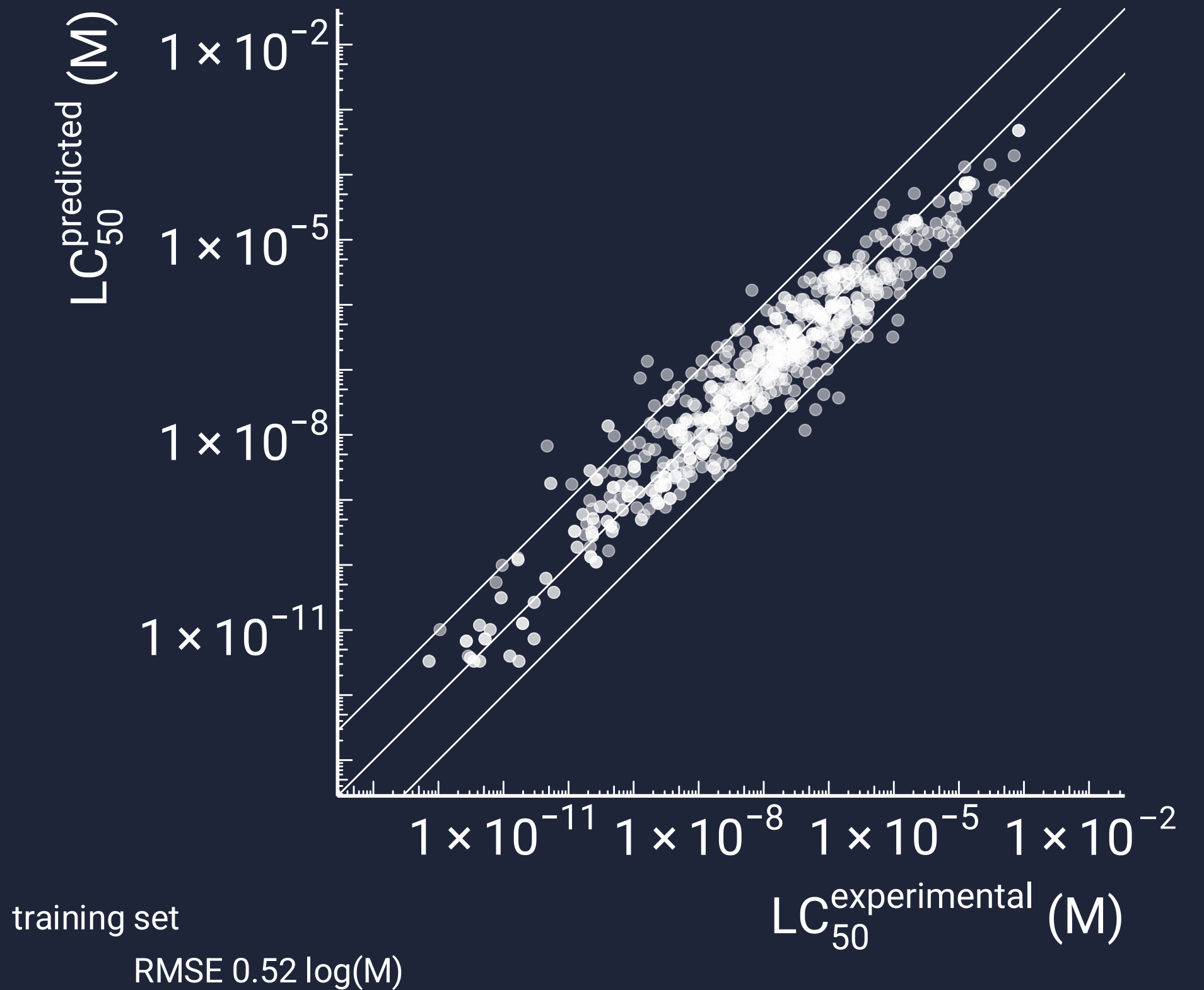
mass (Da)	fp1	...	fp243
317.32000	0	...	0
208.26100	1	...	0
240.21499	1	...	0
300.57998	0	...	0
201.22500	0	...	0

gradient
→
boosting



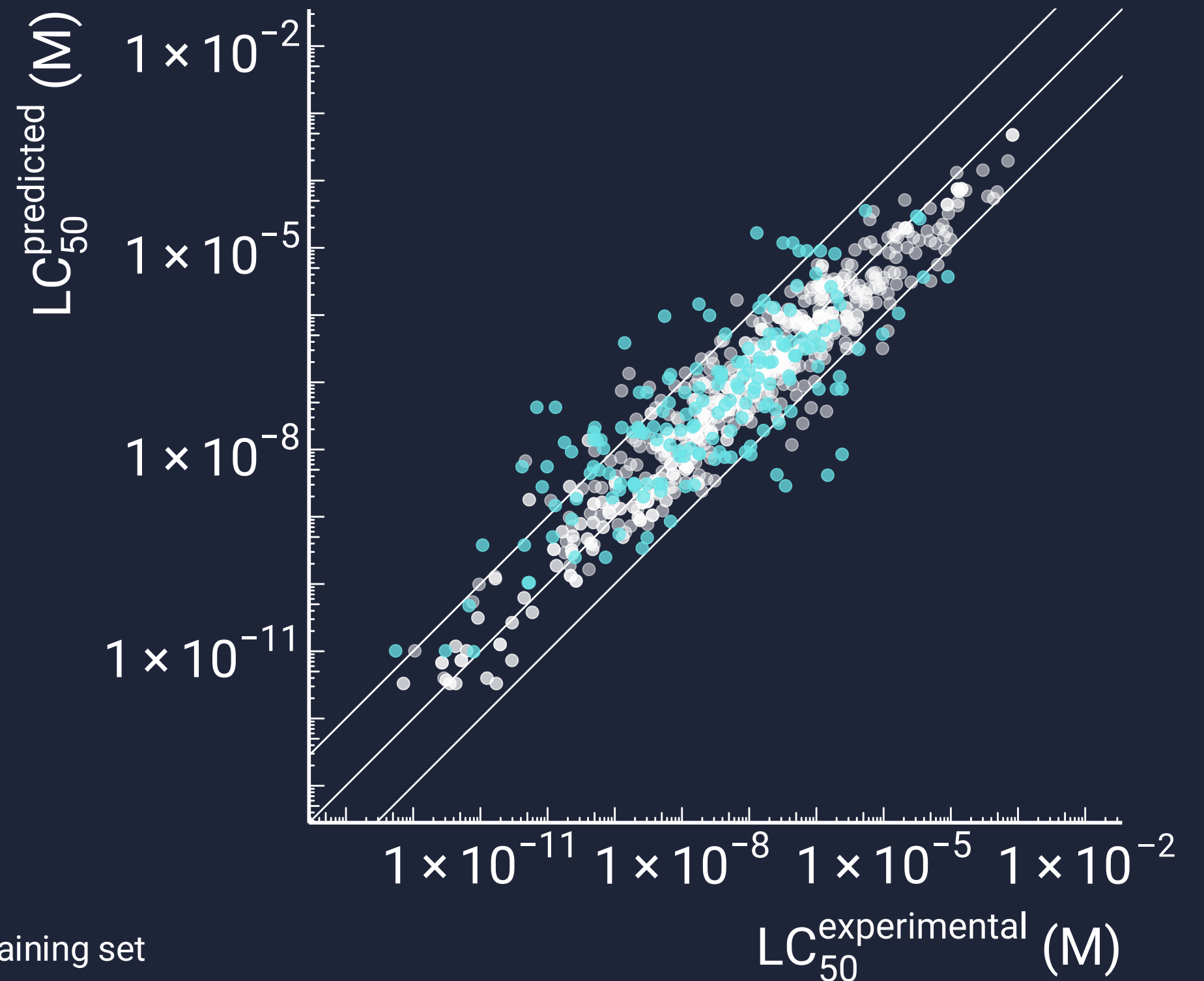
LC₅₀ predictions

Peets et al. ES&T 2022
fish LC₅₀



LC₅₀ predictions

Peets et al. ES&T 2022
fish LC₅₀



training set
RMSE 0.52 log(M)

test set
RMSE 0.78 log(M)

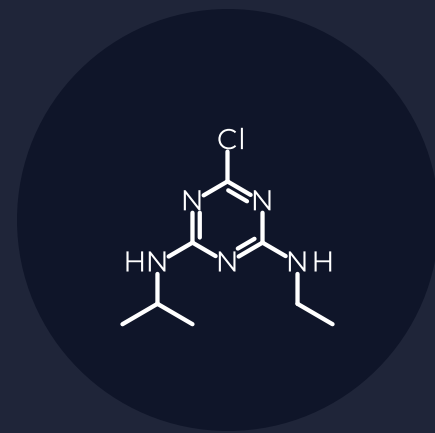
predicting toxicity

of unidentified chemicals

workflow



MS² SPECTRA



STRUCTURE AS SMILES



**MOLECULAR
DESCRIPTORS**



PREDICT TOXICITY

workflow



MS² SPECTRA



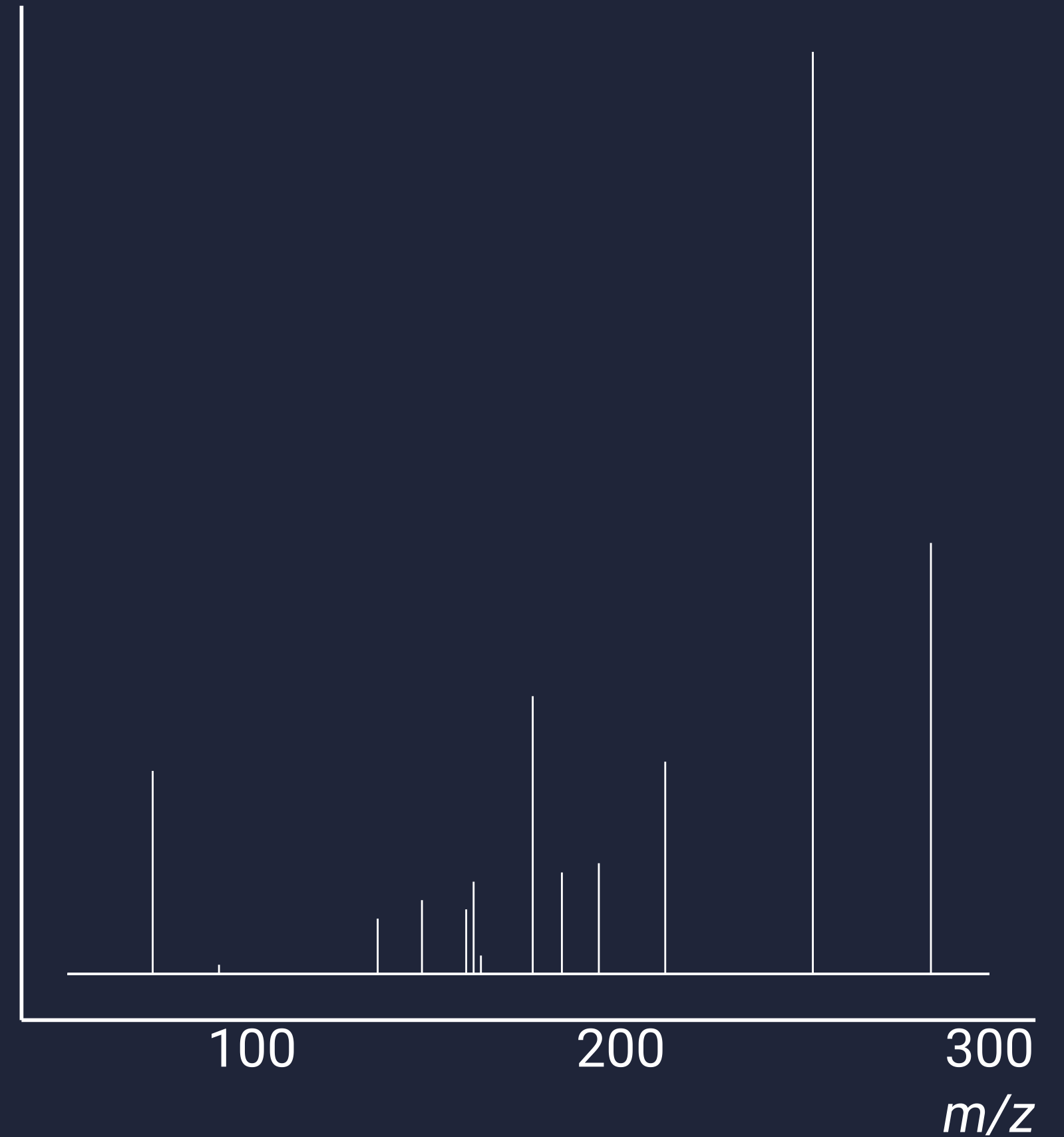
**MOLECULAR
DESCRIPTORS**



PREDICT TOXICITY

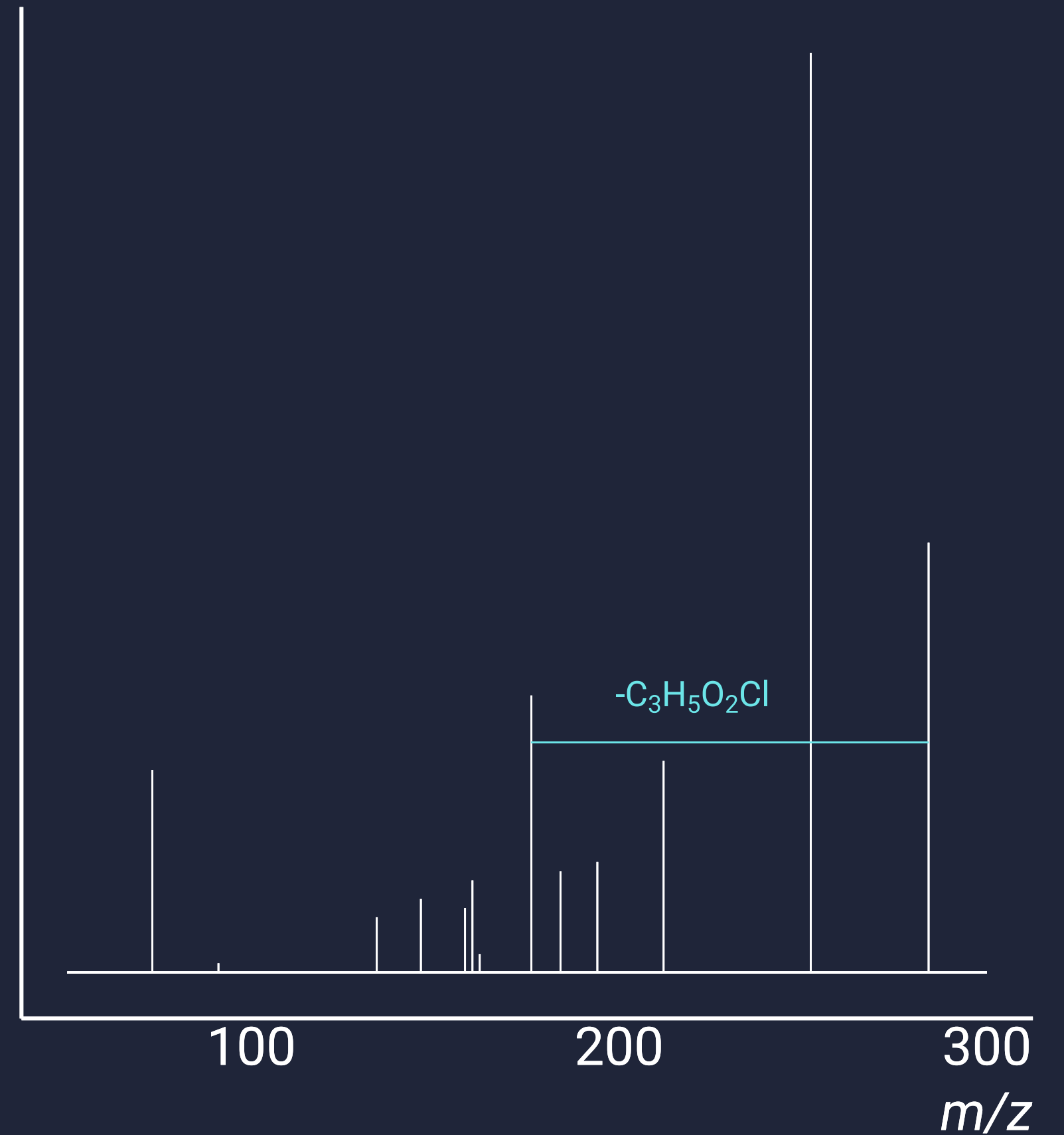
information available

in MS² spectra



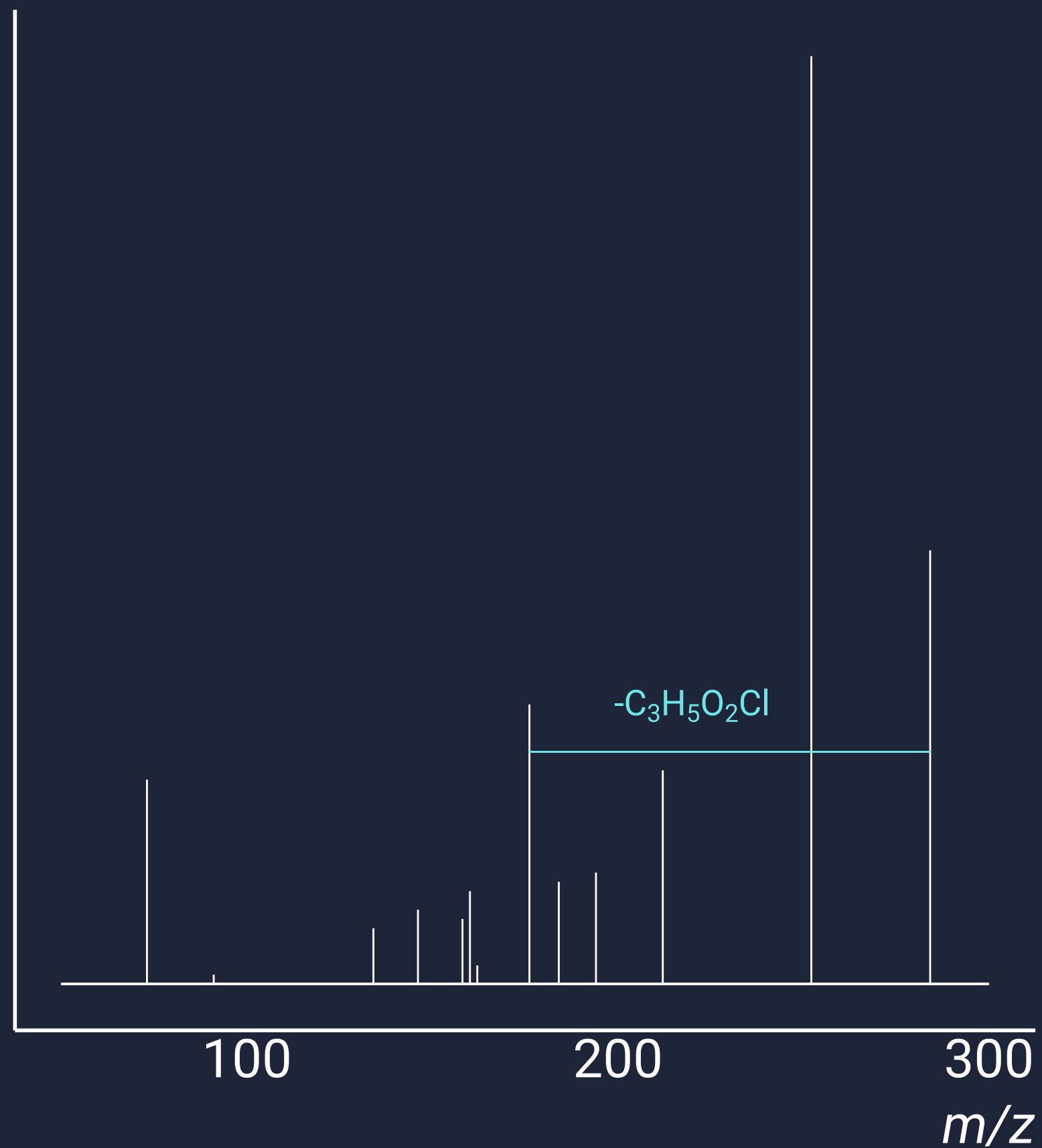
information available

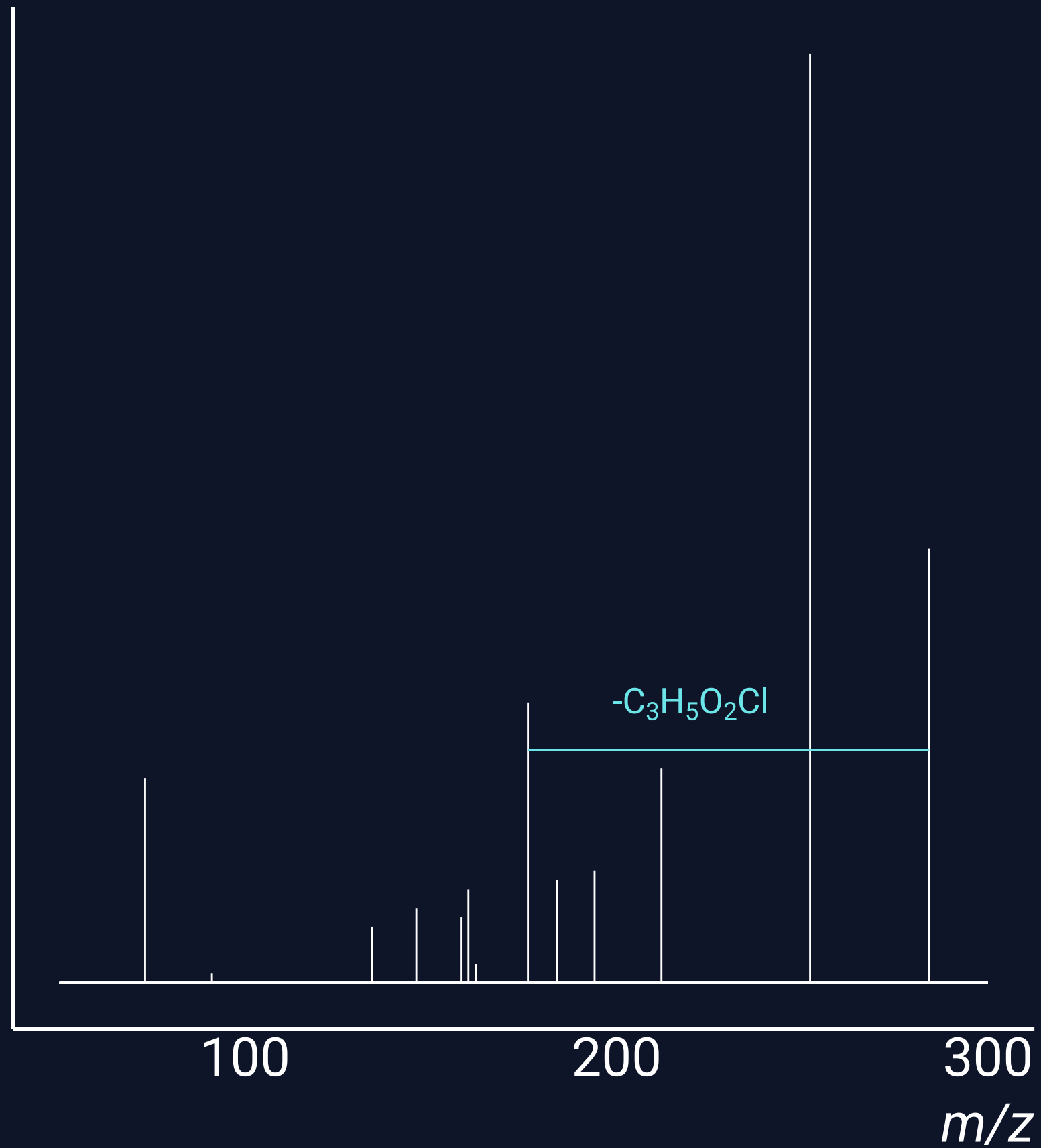
in MS² spectra



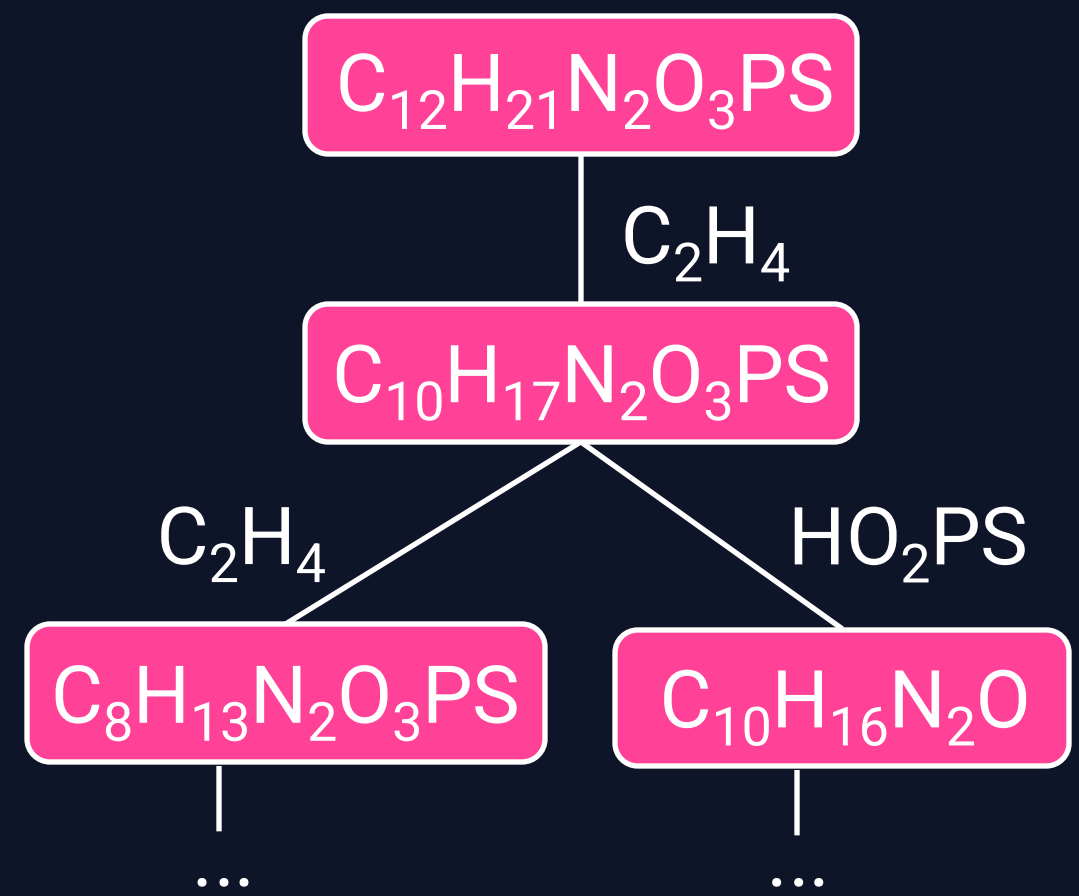
predict toxicity

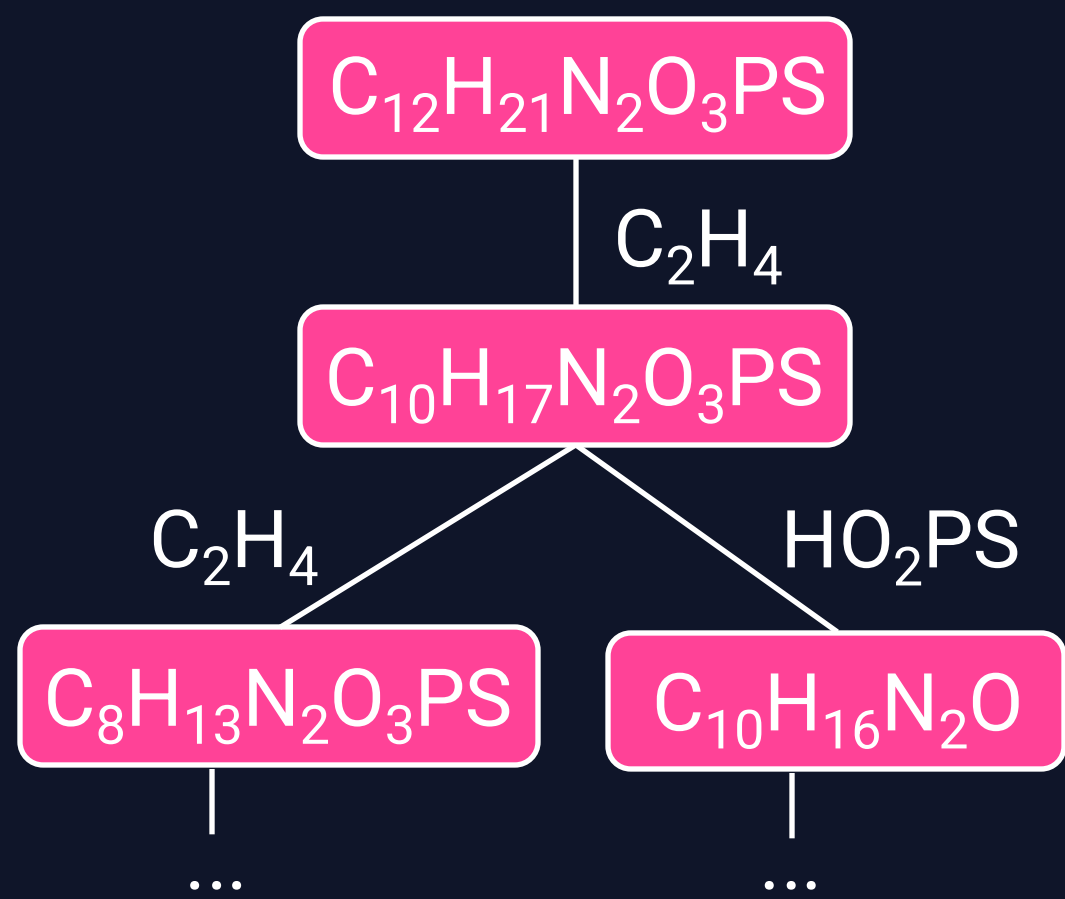
for unknown chemicals










SIRIUS+
CSI:FingerID





SIRIUS+
CSI:FingerID
→

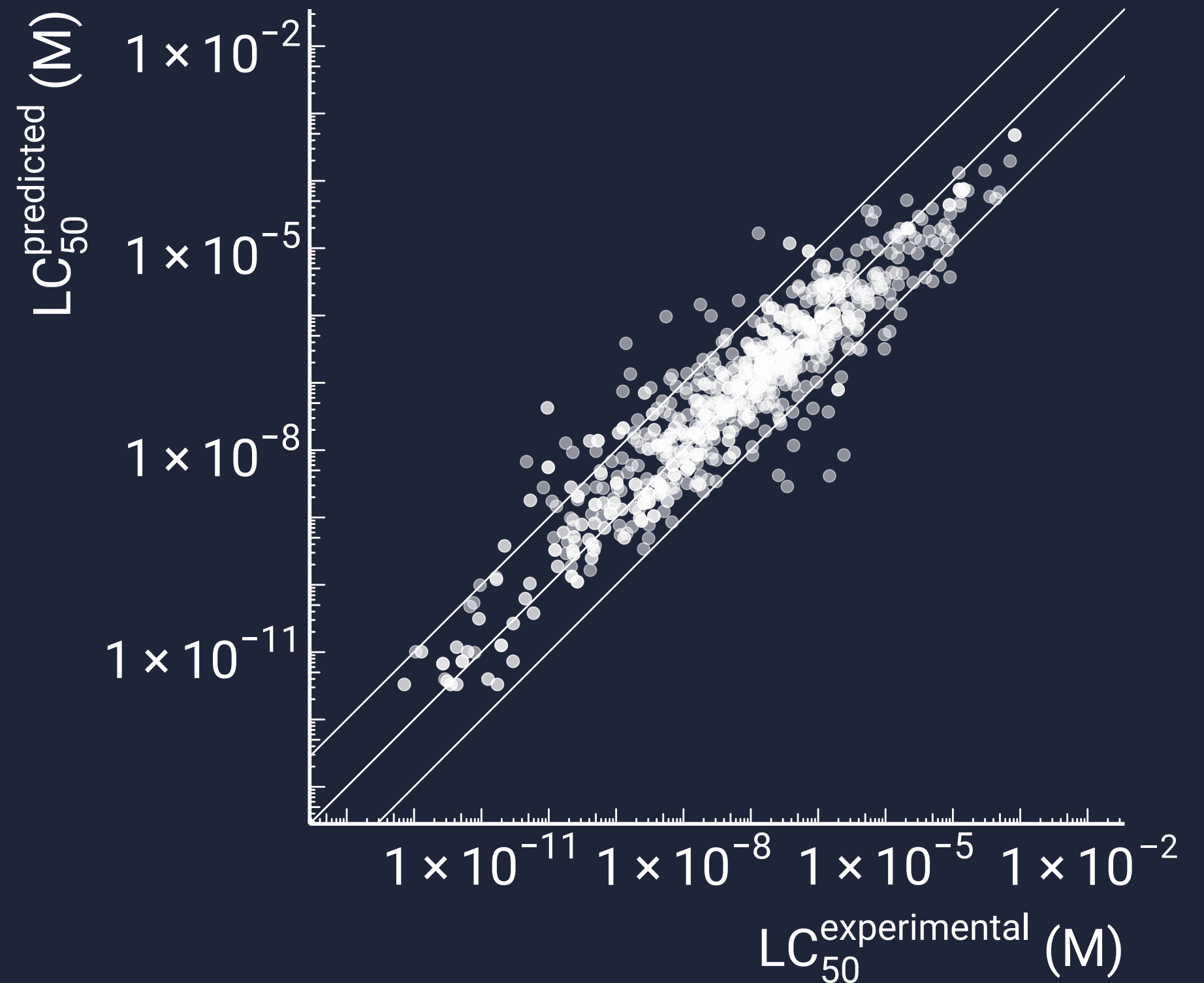
0	
1	
1	
0	
1	

gradient
boosting
→

$$LC_{50} = -2.2 \log(\text{mM})$$

LC₅₀ predictions

Peets et al. ES&T 2022
fish LC₅₀

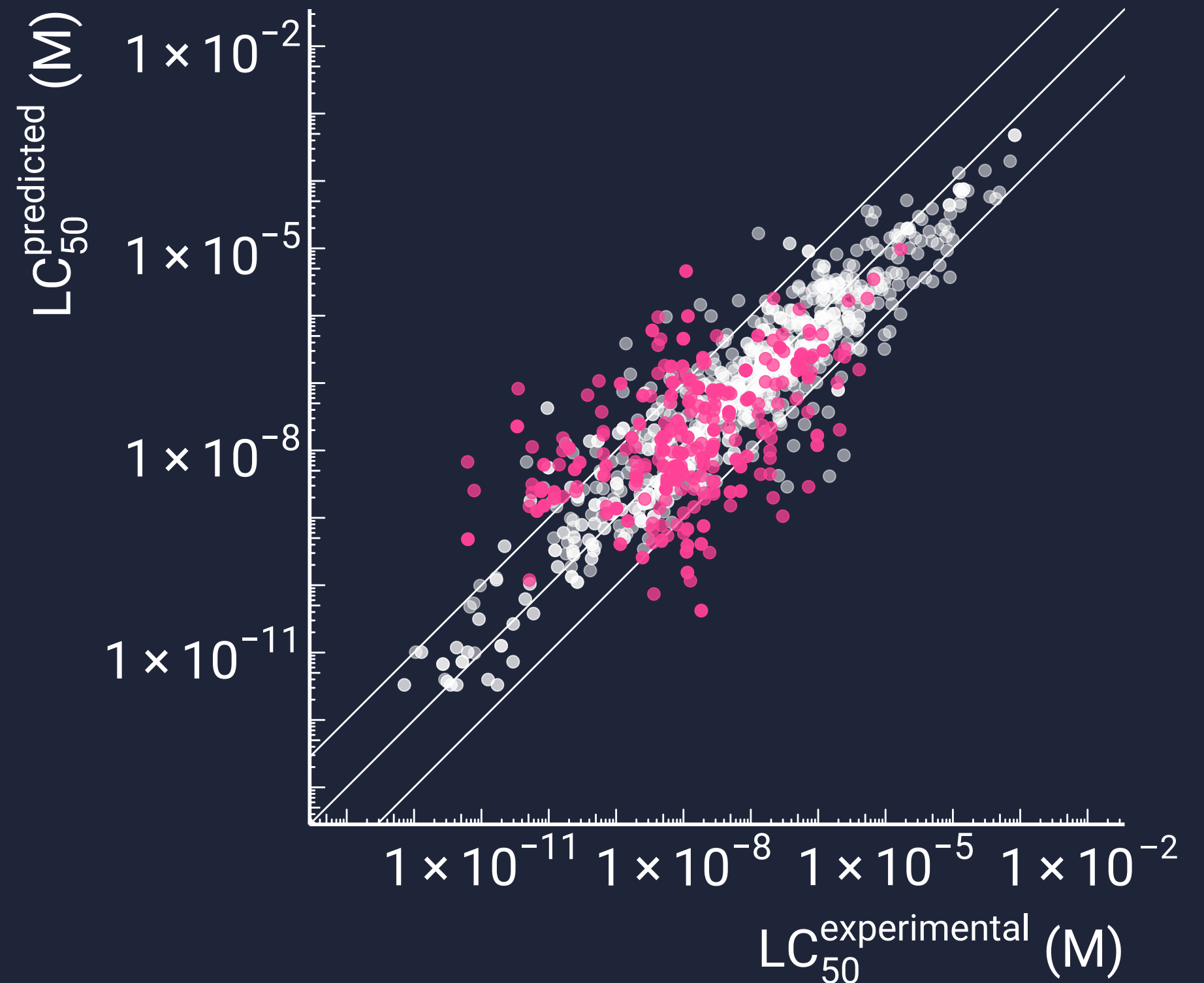


test set

RMSE 0.78 log(M)

LC₅₀ predictions

Peets et al. ES&T 2022
fish LC₅₀



validation on MassBank

RMSE_{model} 0.88 log(M)

SD_{experimental} 0.44 log(M)

endocrine disruption

Rahu et al. JCIM 2024
Tox21 endpoints

endocrine disruption

Rahu et al. JCIM 2024
Tox21 endpoints

true label	
active	non-active

endocrine disruption

Rahu et al. JCIM 2024
Tox21 endpoints

		true label	
		active	non-active
prediction	active		
	non-active		

endocrine disruption

Rahu et al. JCIM 2024
Tox21 endpoints

		true label	
		active	non-active
prediction	active	TP	
	non-active		TN

endocrine disruption

Rahu et al. JCIM 2024
Tox21 endpoints

		true label	
		active	non-active
prediction	active	TP	FP
	non-active		TN

endocrine disruption

Rahu et al. JCIM 2024
Tox21 endpoints

		true label	
		active	non-active
prediction	active	TP	FP
	non-active	FN	TN

endocrine disruption

Rahu et al. JCIM 2024
Tox21 endpoints

		true label	
		active	non-active
prediction	active	TP	FP
	non-active	FN	TN

which is more dramatic:
FP or FN?

endocrine disruption

Rahu et al. JCIM 2024
Tox21 endpoints

		true label	
		active	non-active
prediction	active	TP	FP
	non-active	FN	TN

FPR @ TPR = 0.9

endocrine disruption

Rahu et al. JCIM 2024
Tox21 endpoints

bioassay	FPR
sr.mmp	25.1%
sr.p53	25.4%
nr.ahr	41.8%
...	...
nr.ar	82.4%
nr.er	85.0%

MassBank & MoNA

748 compounds with MS² & tox

case study: interlaboratory comparison



interlaboratory comparison

Sandberg, Rahu, in preparation



SAMPLES

spiked water samples



ANALYSIS

HRMS, etc. characterization



DATA PROCESSING

AhR activity

results

4700

LC/HRMS features detected

238

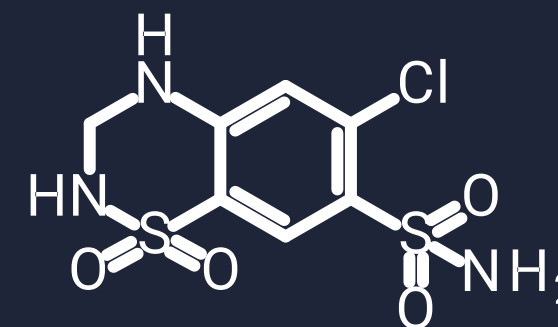
features with MS² spectra

55

features predicted active

AhR active

hydrochlorothiazide

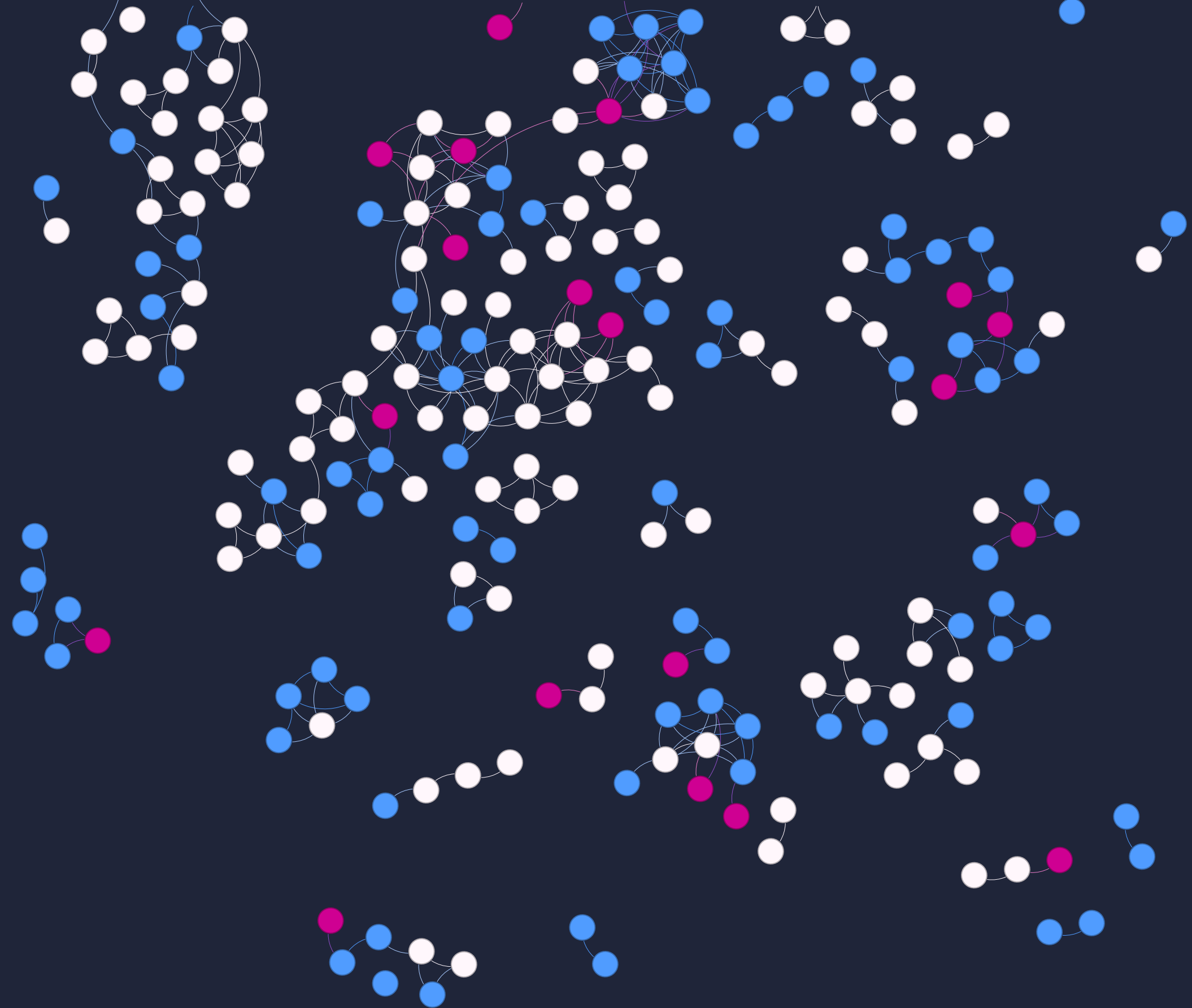


alternative approaches

Kreutzer et al. in preparation
Tox21 endpoints

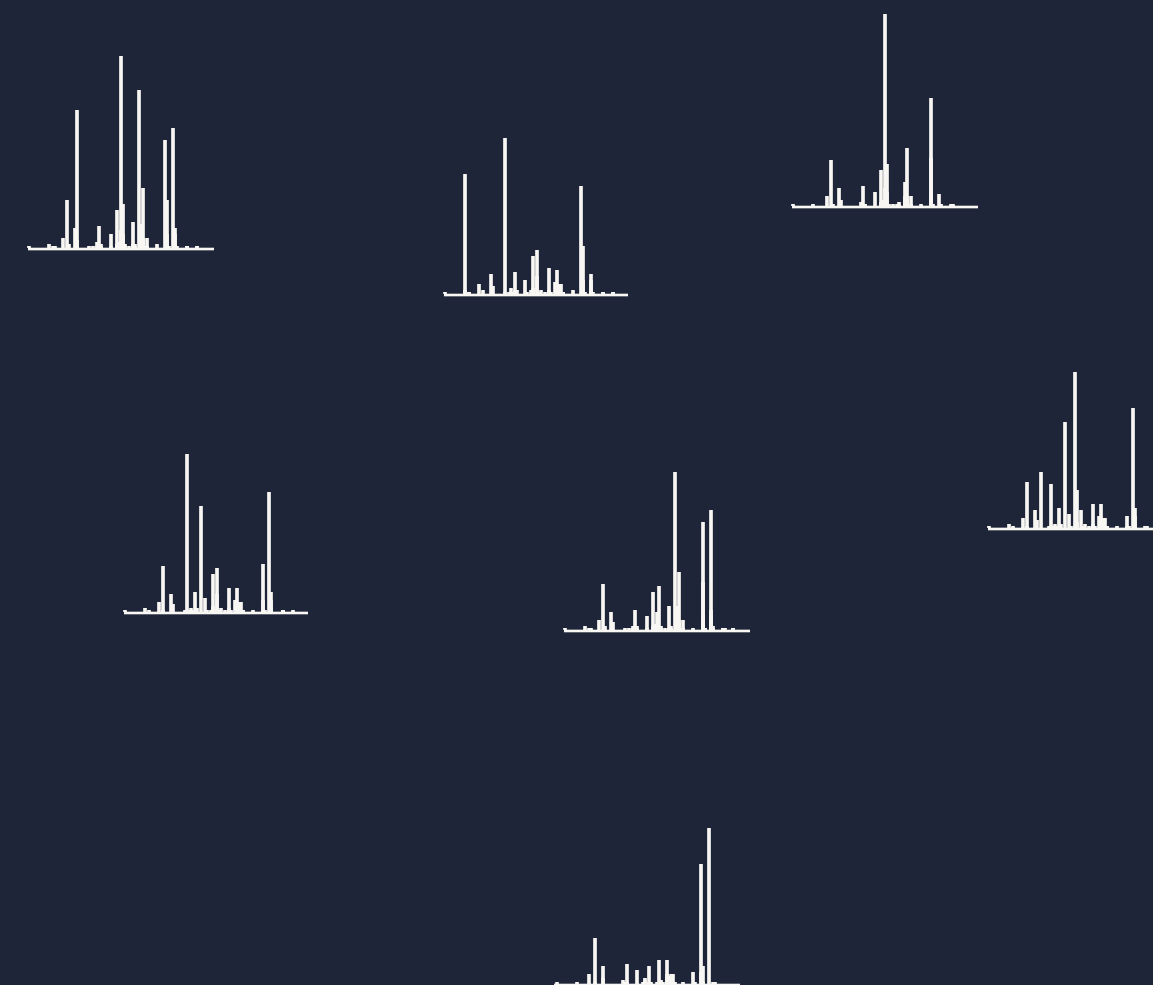
molecular networks

Kreutzer et al. in preparation
Tox21 endpoints



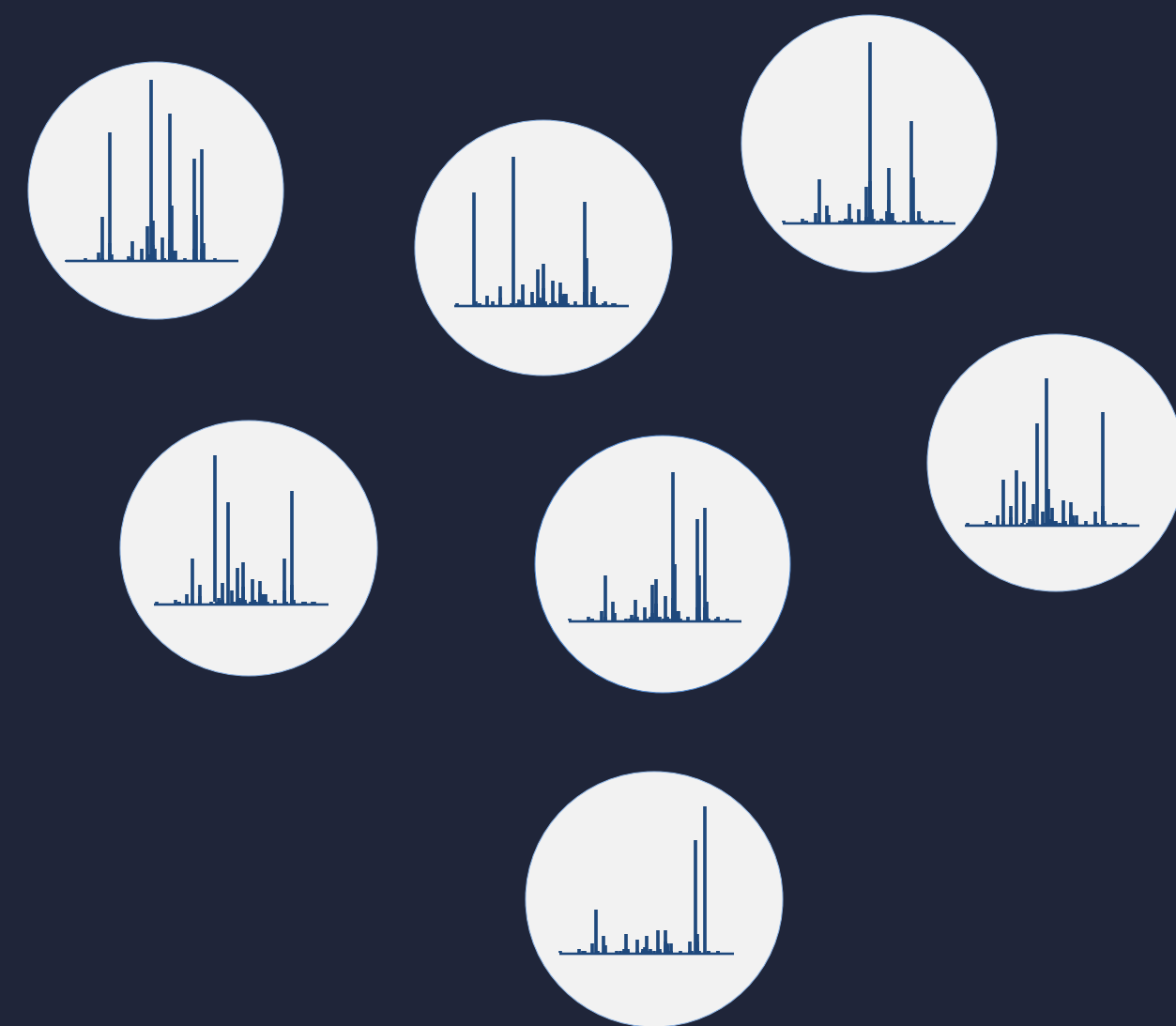
molecular networks

Kreutzer et al. in preparation
Tox21 endpoints



molecular networks

Kreutzer et al. in preparation
Tox21 endpoints



molecular networks

Kreutzer et al. in preparation
Tox21 endpoints



molecular networks

Kreutzer et al. in preparation
Tox21 endpoints



- Active
- Inactive
- Inconclusive

molecular networks

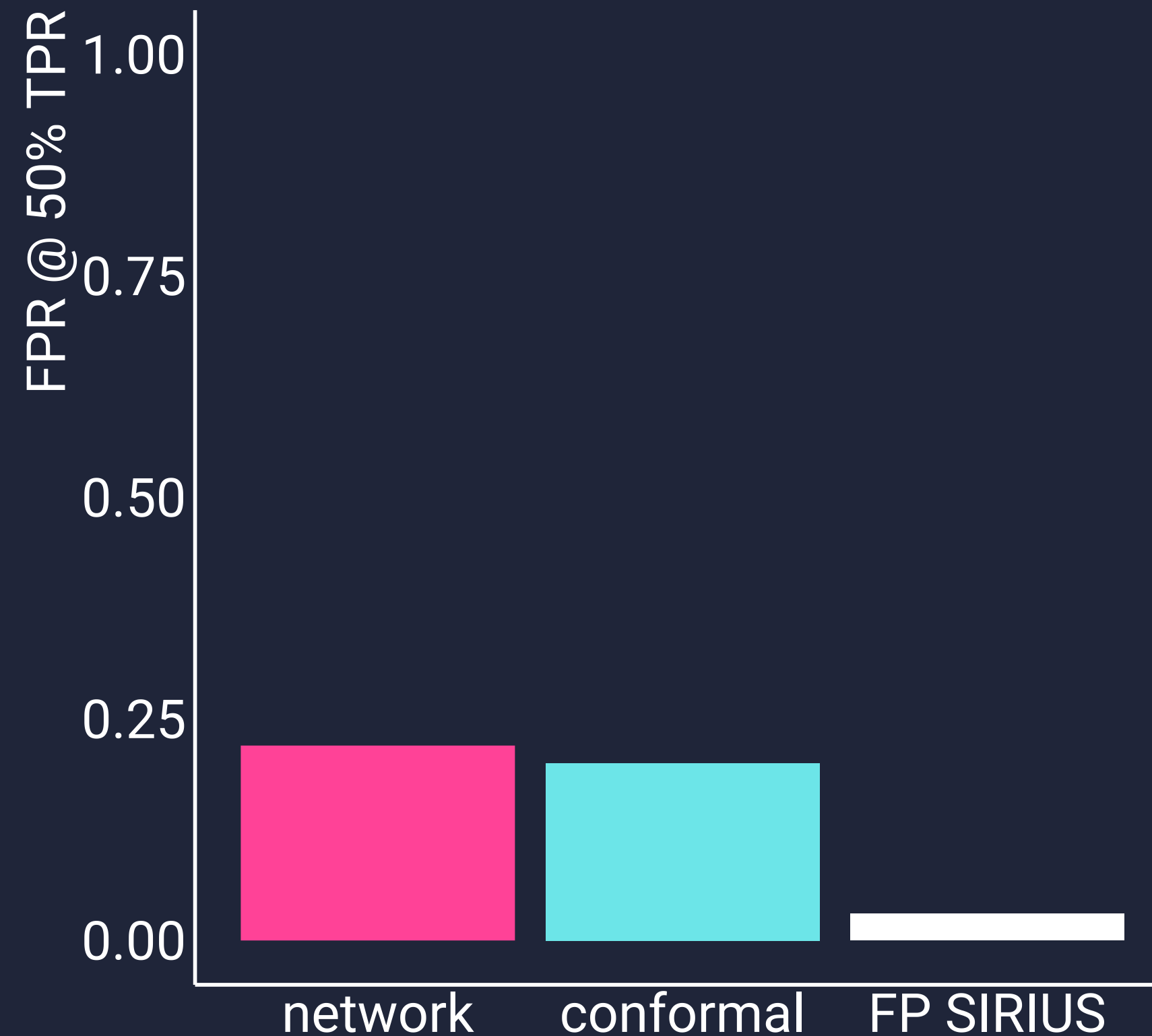
Kreutzer et al. in preparation
Tox21 endpoints



- Active
- Inactive
- Inconclusive
- Unknown

AhR activity predictions

Kreutzer et al. in preparation
Tox21 endpoints

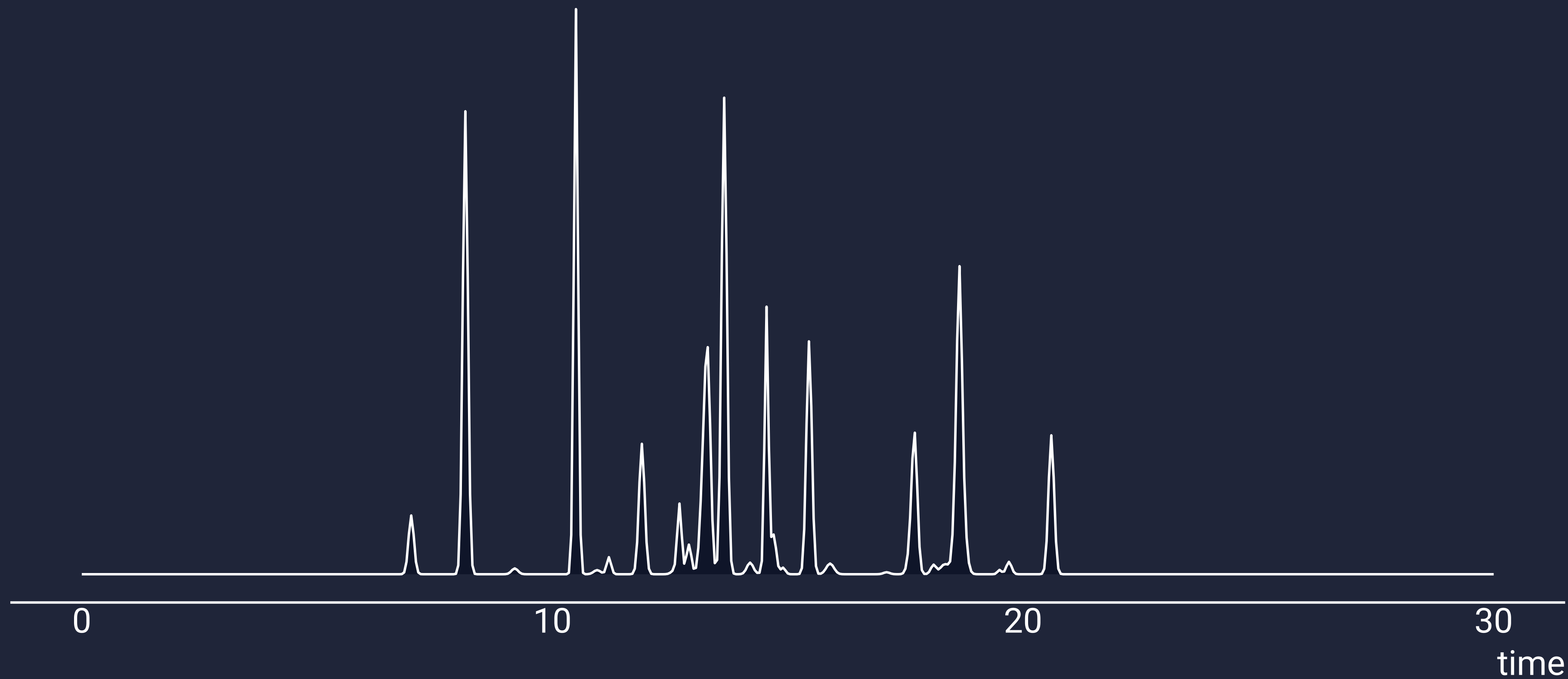


quantification

of detected chemicals

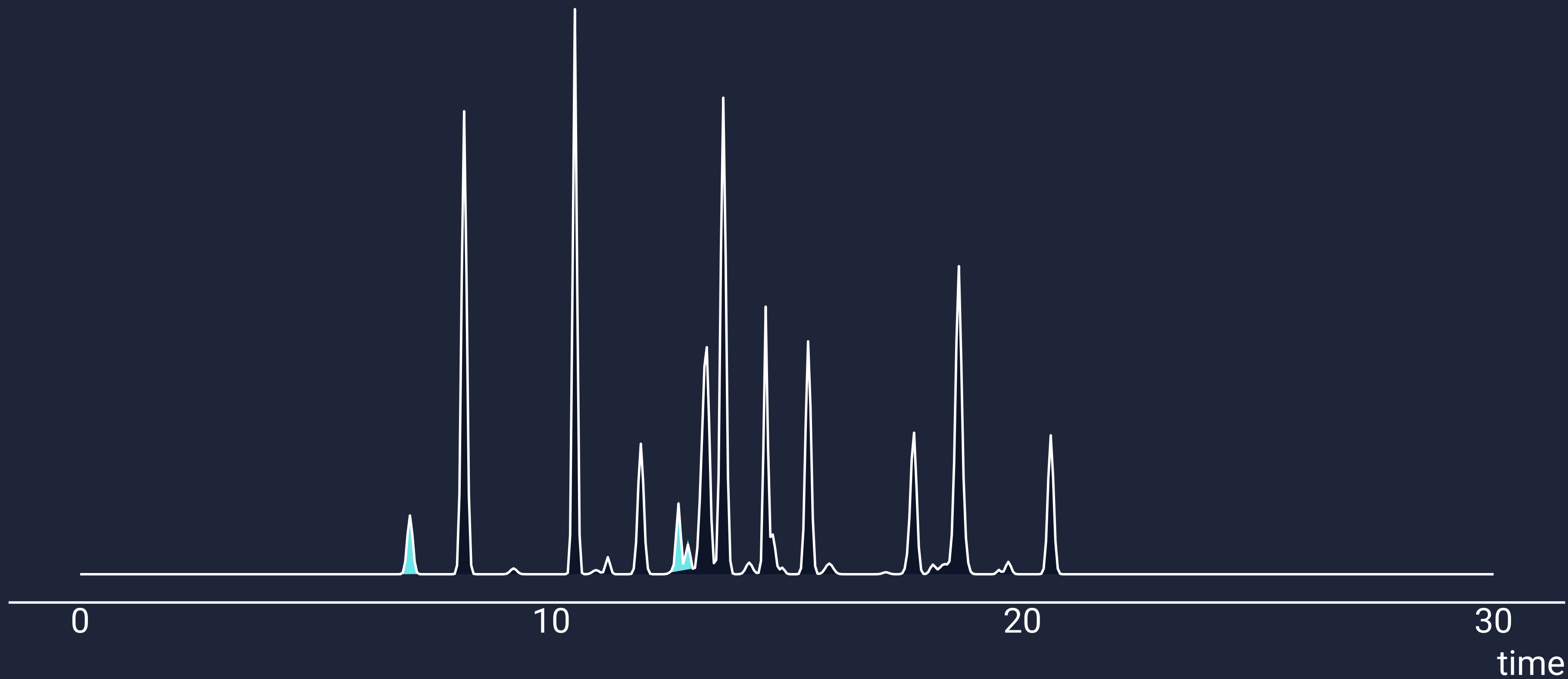
quantification in ESI/HRMS

Malm et al. Molecules 2021



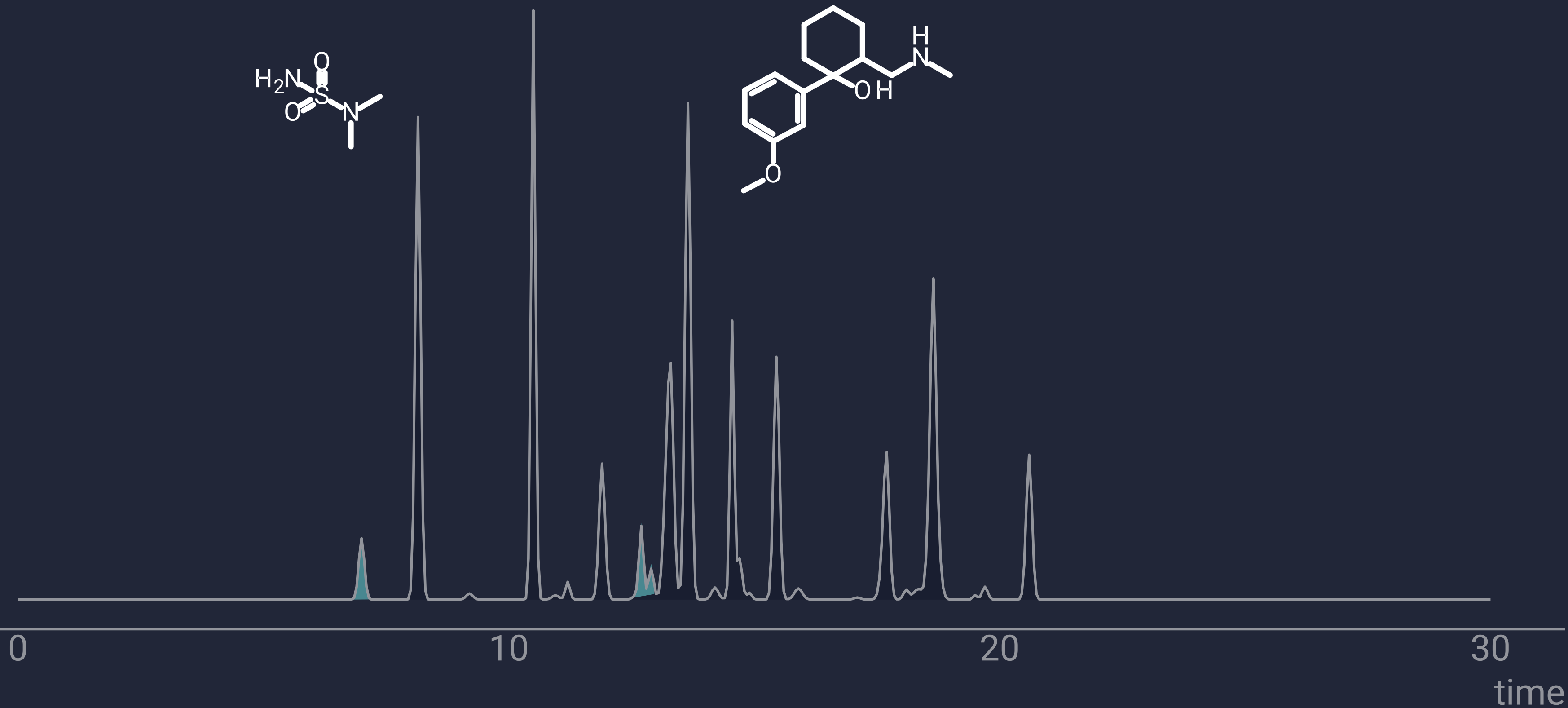
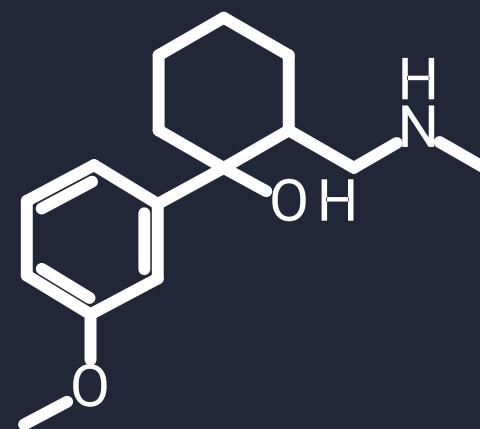
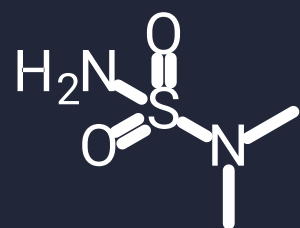
quantification in ESI/HRMS

Malm et al. Molecules 2021



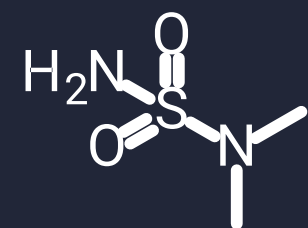
quantification in ESI/HRMS

Malm et al. Molecules 2021

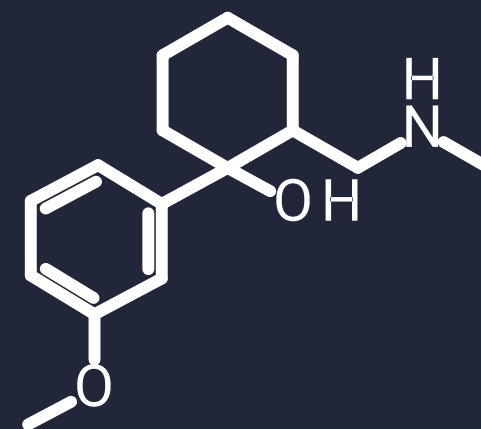


quantification in ESI/HRMS

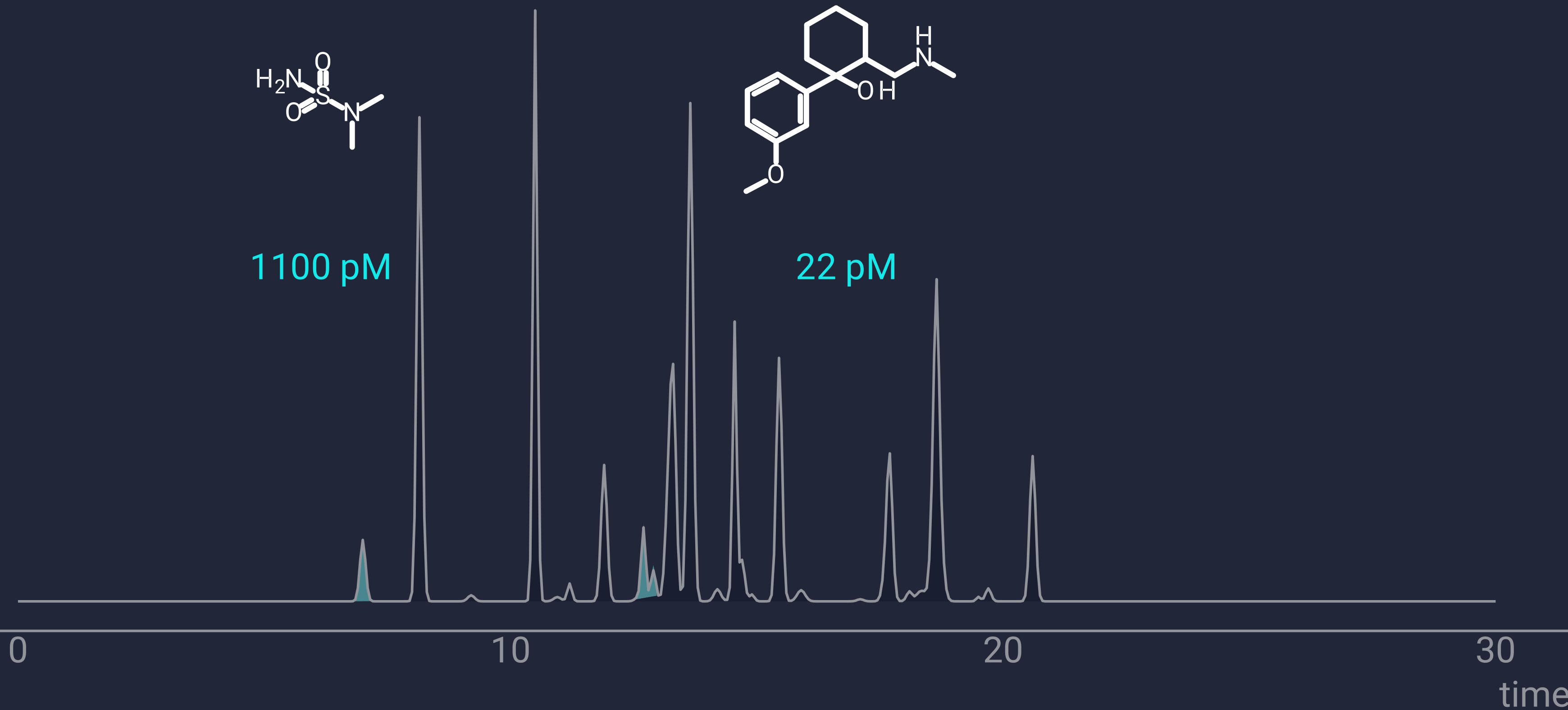
Malm et al. Molecules 2021



1100 pM



22 pM



electrospray

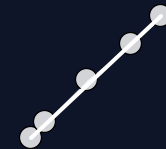


ionization efficiency



ANALYSIS

flow injections or LC



DATA

calibration graphs

$$\frac{\text{slope}_1}{\text{slope}_2} \rightarrow IE$$

IONIZATION EFFICIENCY

relative measurements

ionization efficiency

$1 \times 10^{+5}$

$1 \times 10^{+3}$

$1 \times 10^{+1}$

ionization efficiency

one solvent, purely analyte properties

377 chemicals

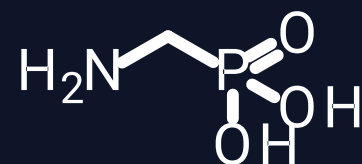
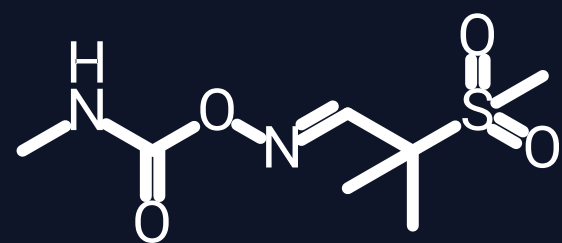
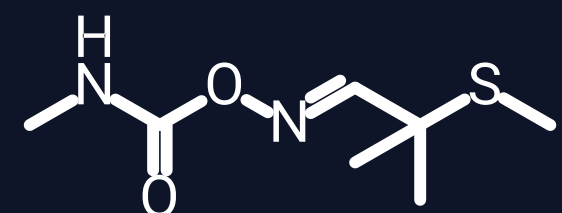
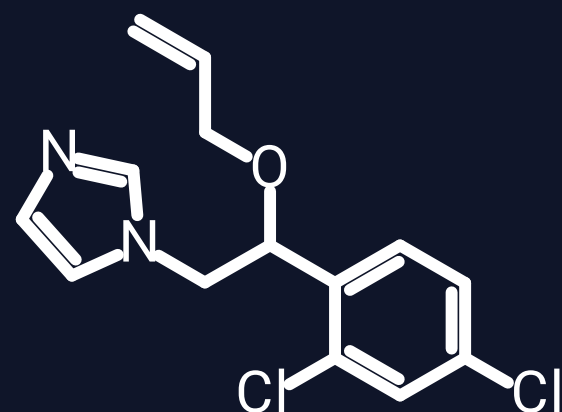
10,000,000x difference in *IE*

ionization efficiency

$1 \times 10^{+5}$

$1 \times 10^{+3}$

$1 \times 10^{+1}$



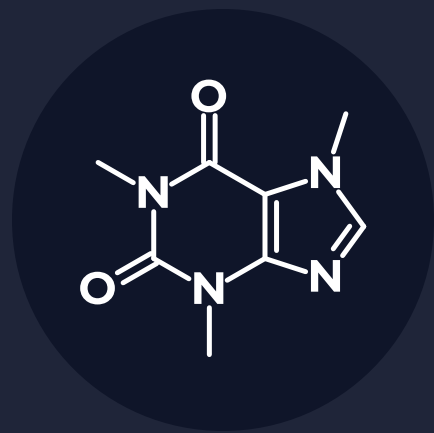
ionization efficiency

one solvent, purely analyte properties

377 chemicals

10,000,000x difference in *IE*

ionization efficiency



STRUCTURE



MOBILE PHASE



INSTRUMENT

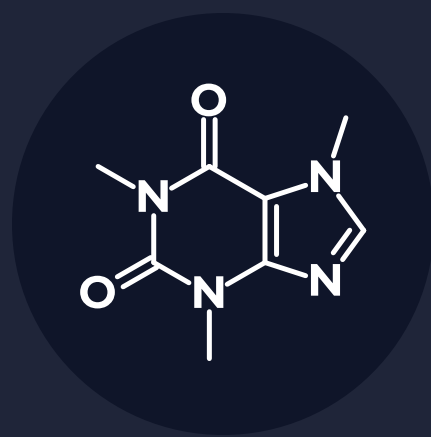


MATRIX

quantification

with machine learning

workflow



SMILES or MS²

+ mobile phase
composition



**MOLECULAR
DESCRIPTORS**

PaDEL, Mordred, SIRIUS
fingerprints



**MODEL
TRAINING**

RandomForest, xgbTree



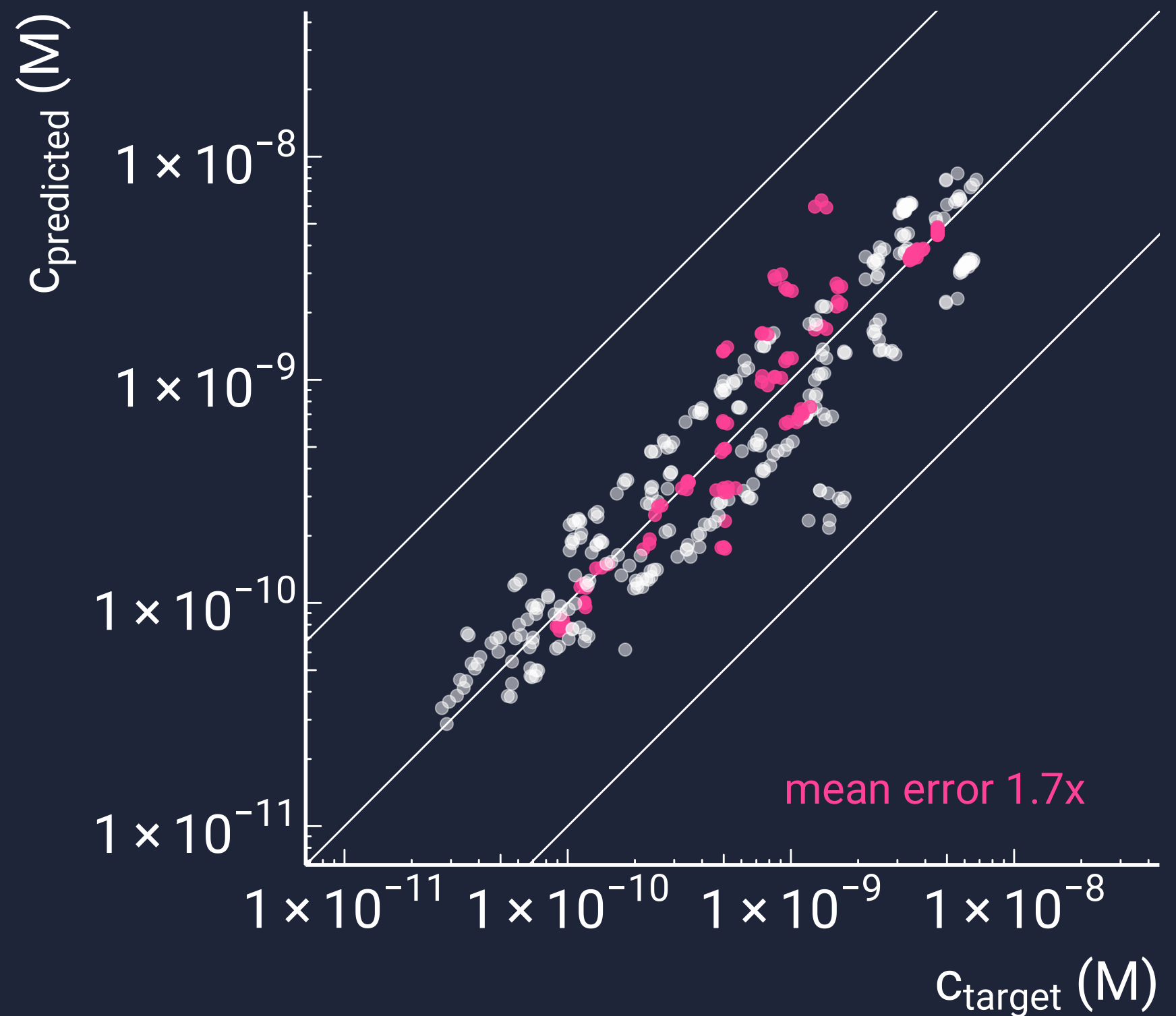
**MODEL
SELECTION**

Root Mean Square Error
(RMSE)

quantification

from structure

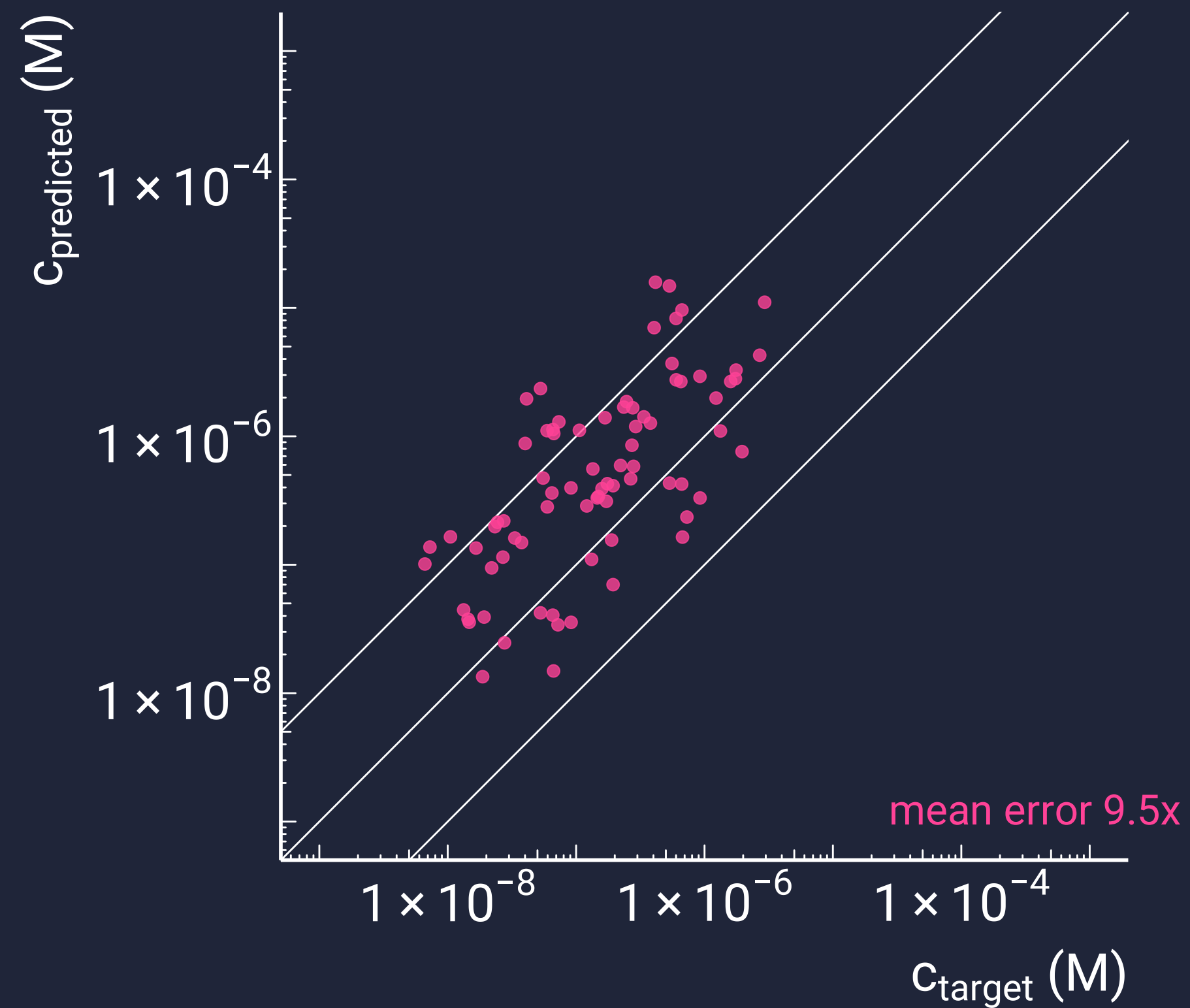
Been et al. Water Research 2021



quantification

from MS² spectra

Sepman et al. Anal Chem 2023



how to ...



PRIORITIZE

risk



IDENTIFY

structure

how to ...

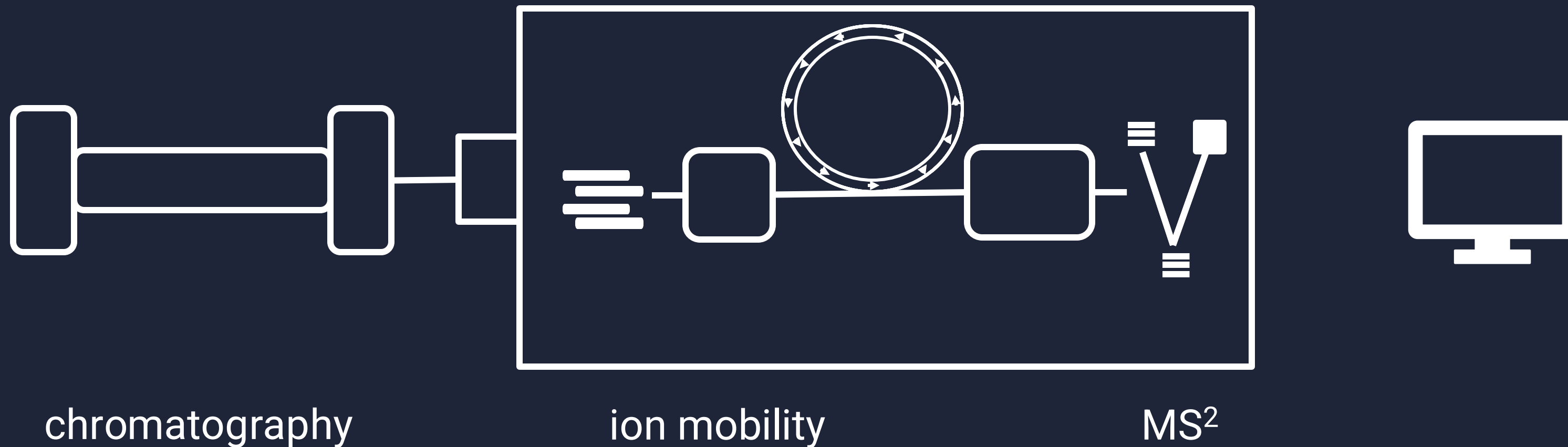
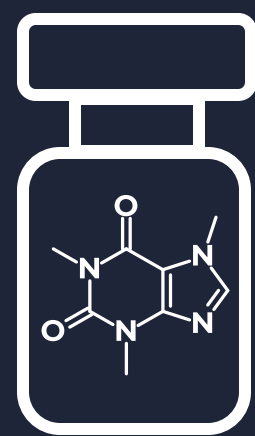


IDENTIFY

structure

orthogonal separation

Akhlaqi et al. Anal Bioanal Chem 2023

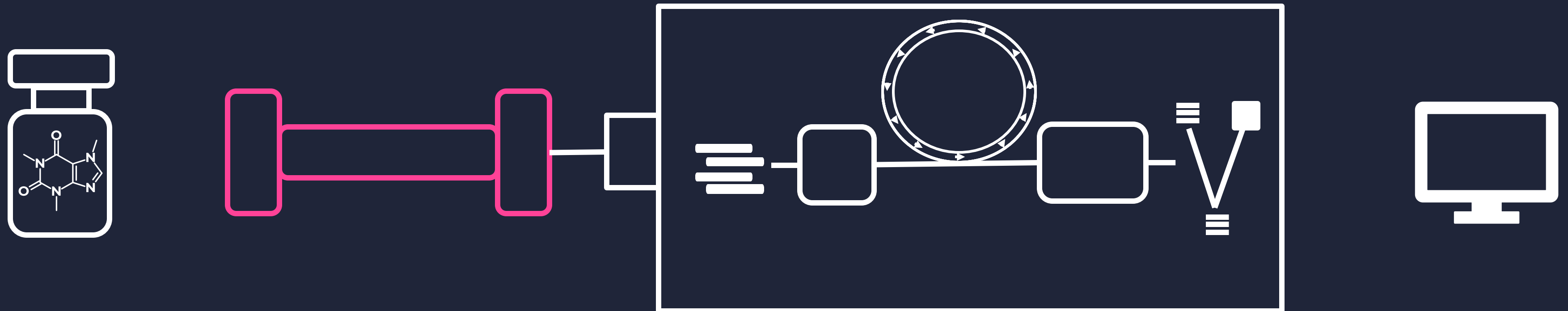


chromatography

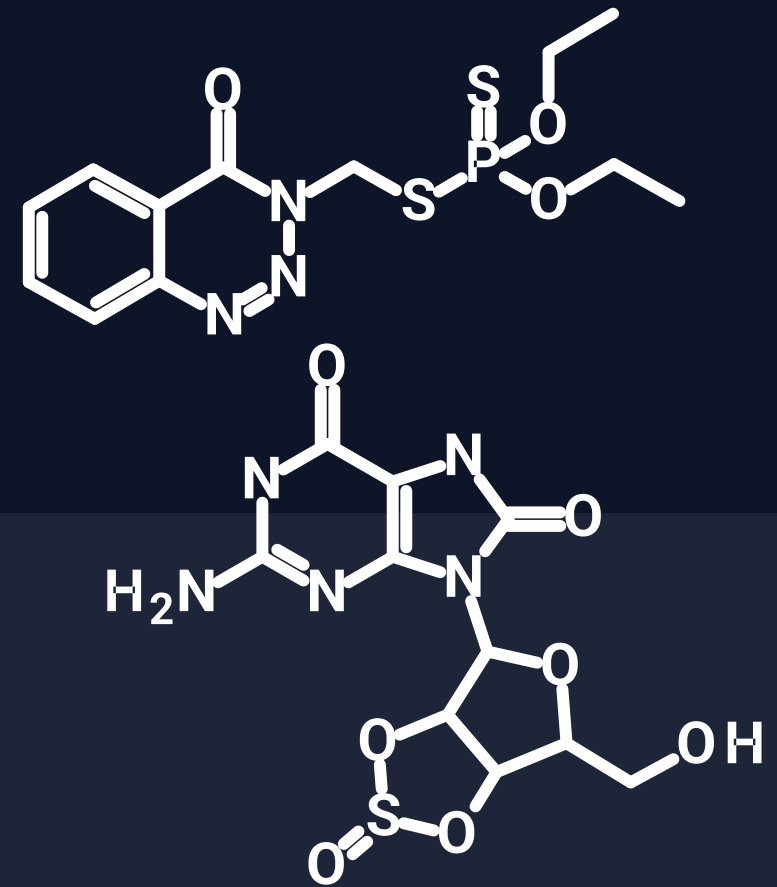
predicting retention time

chromatography

Kruve et al. in preparation



where is the problem

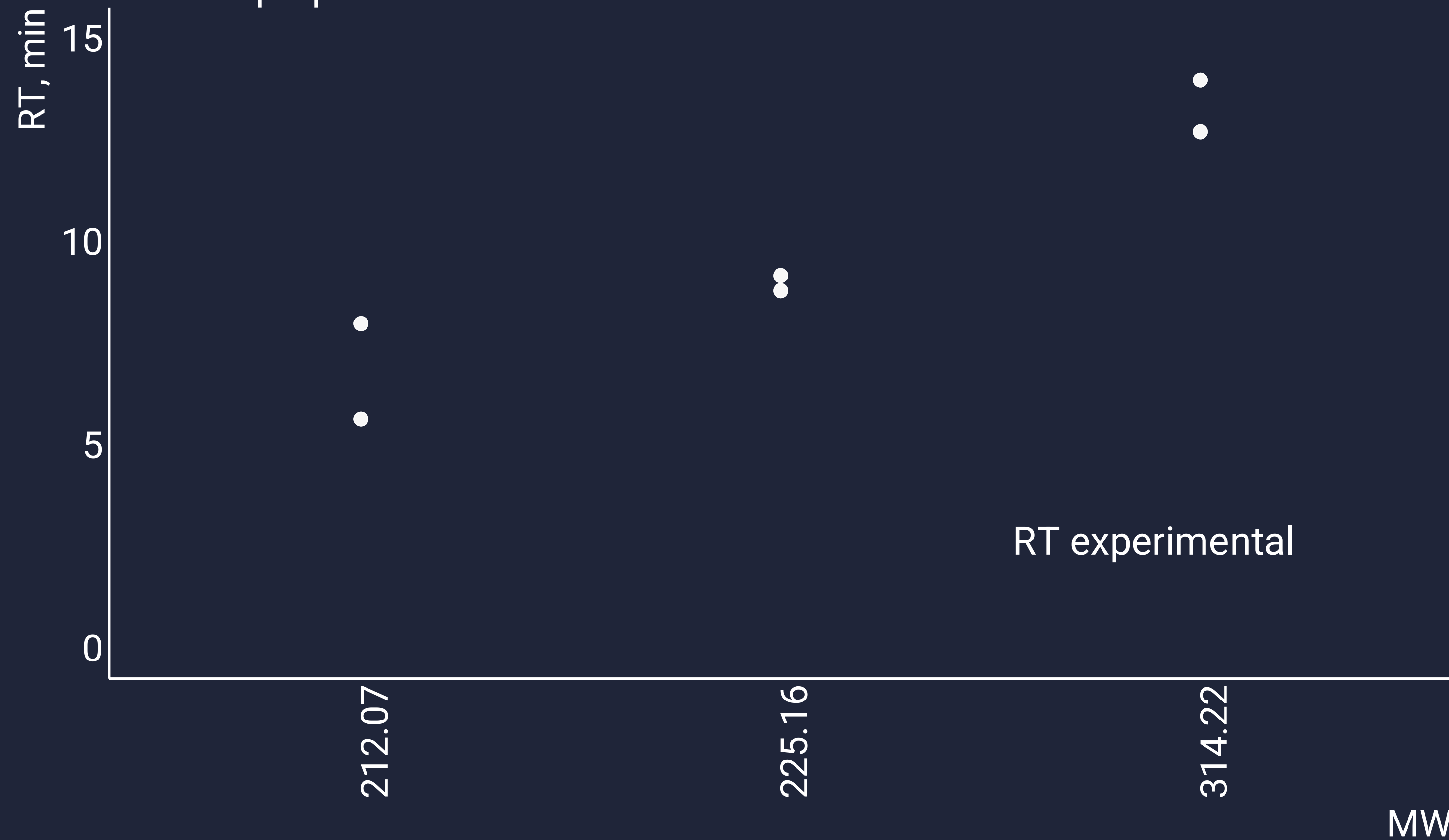


CANDIDATE STRUCTURES

ability to distinguish isomers

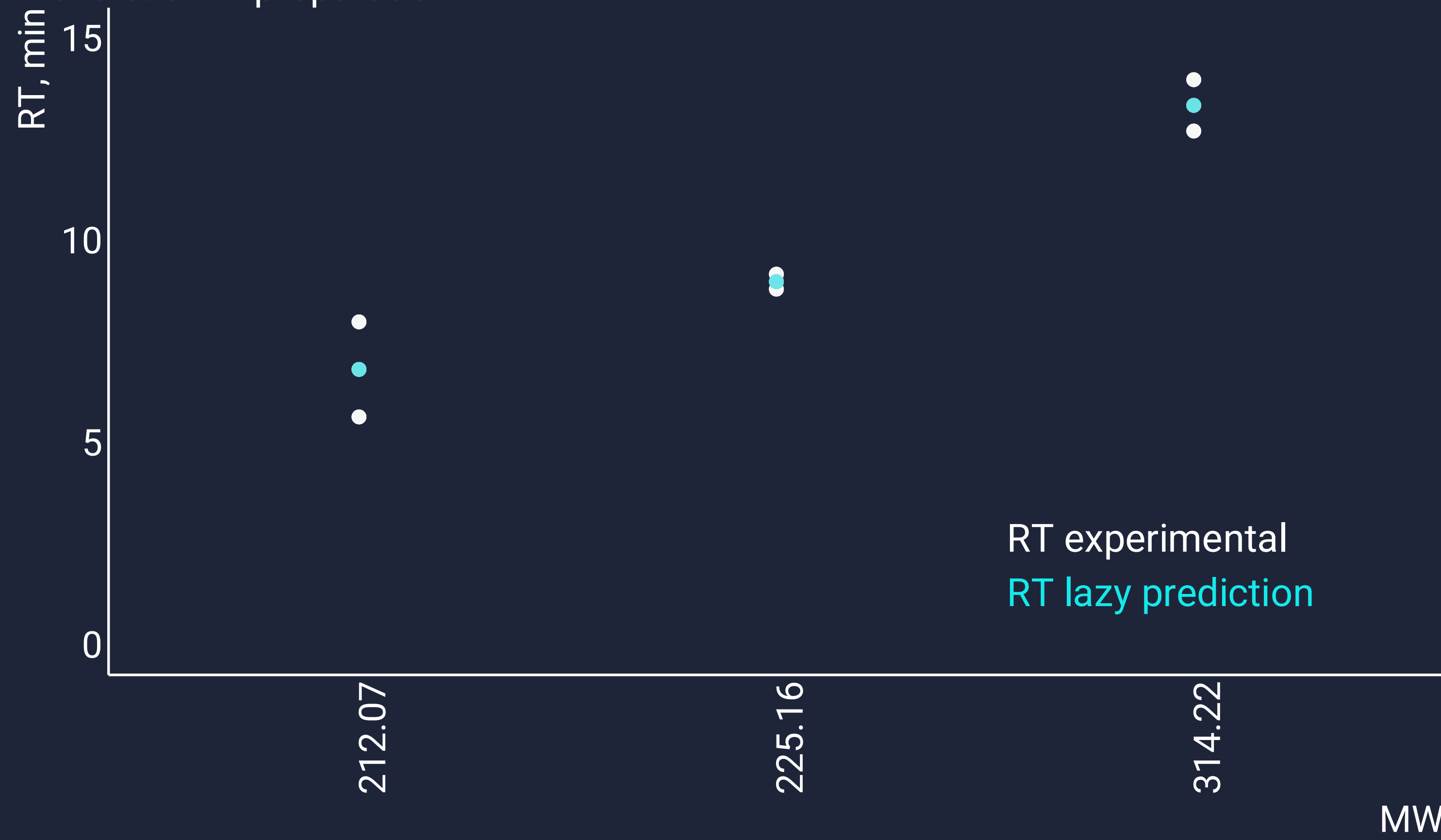
chromatography

Kruve et al. in preparation



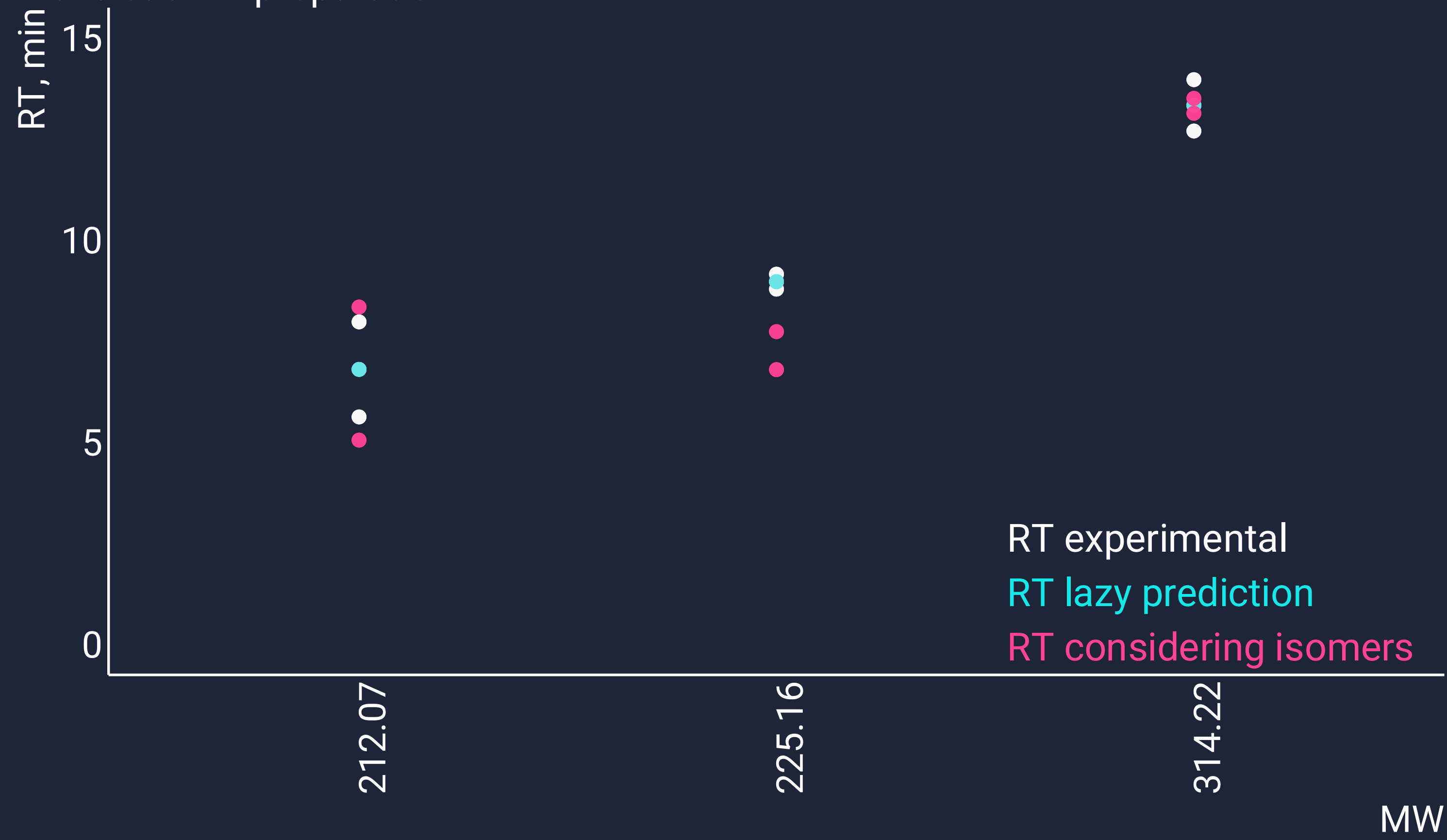
chromatography

Kruve et al. in preparation



chromatography

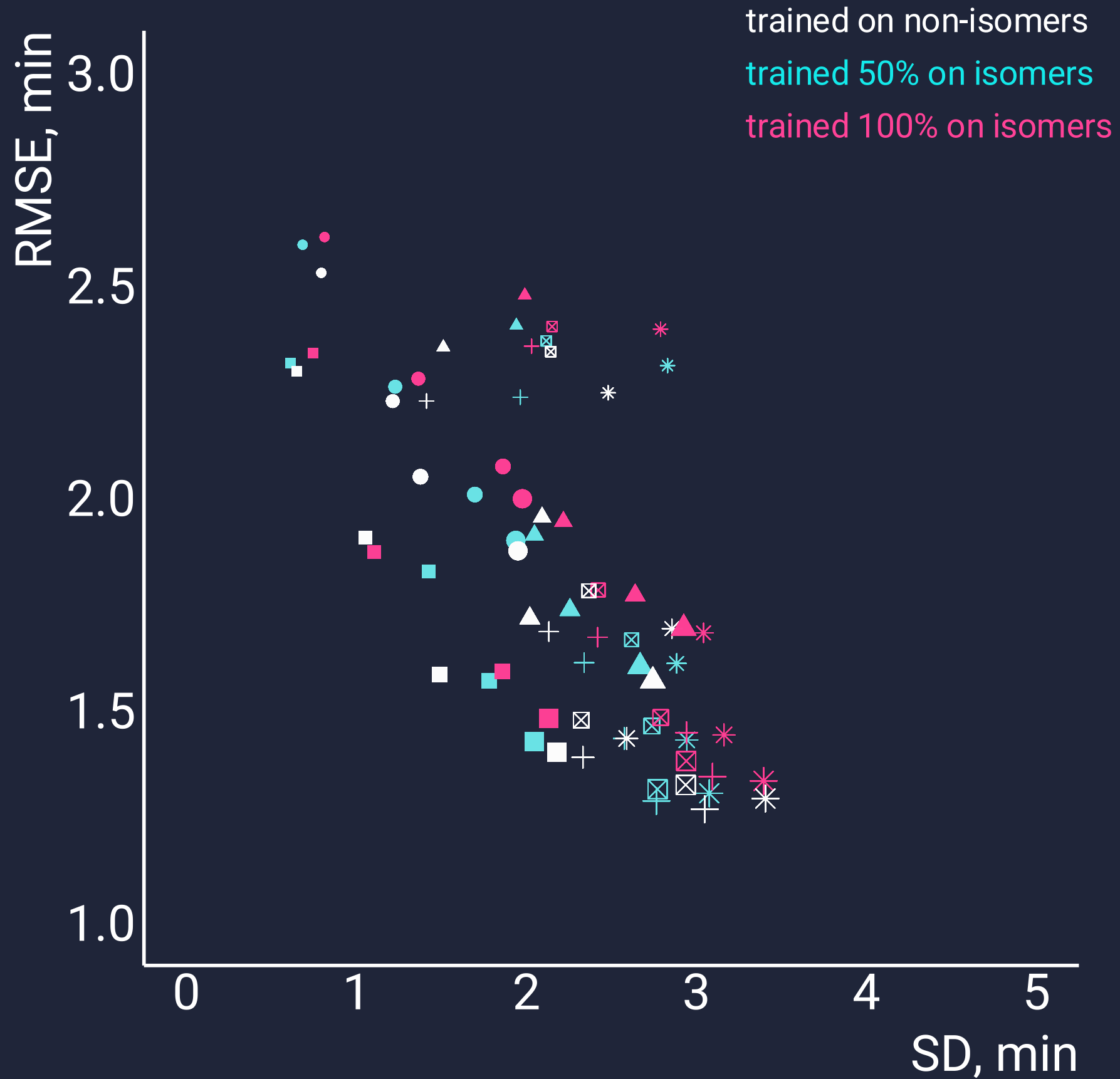
Kruve et al. in preparation



chromatography

for isomer identification

Kruve et al. in preparation

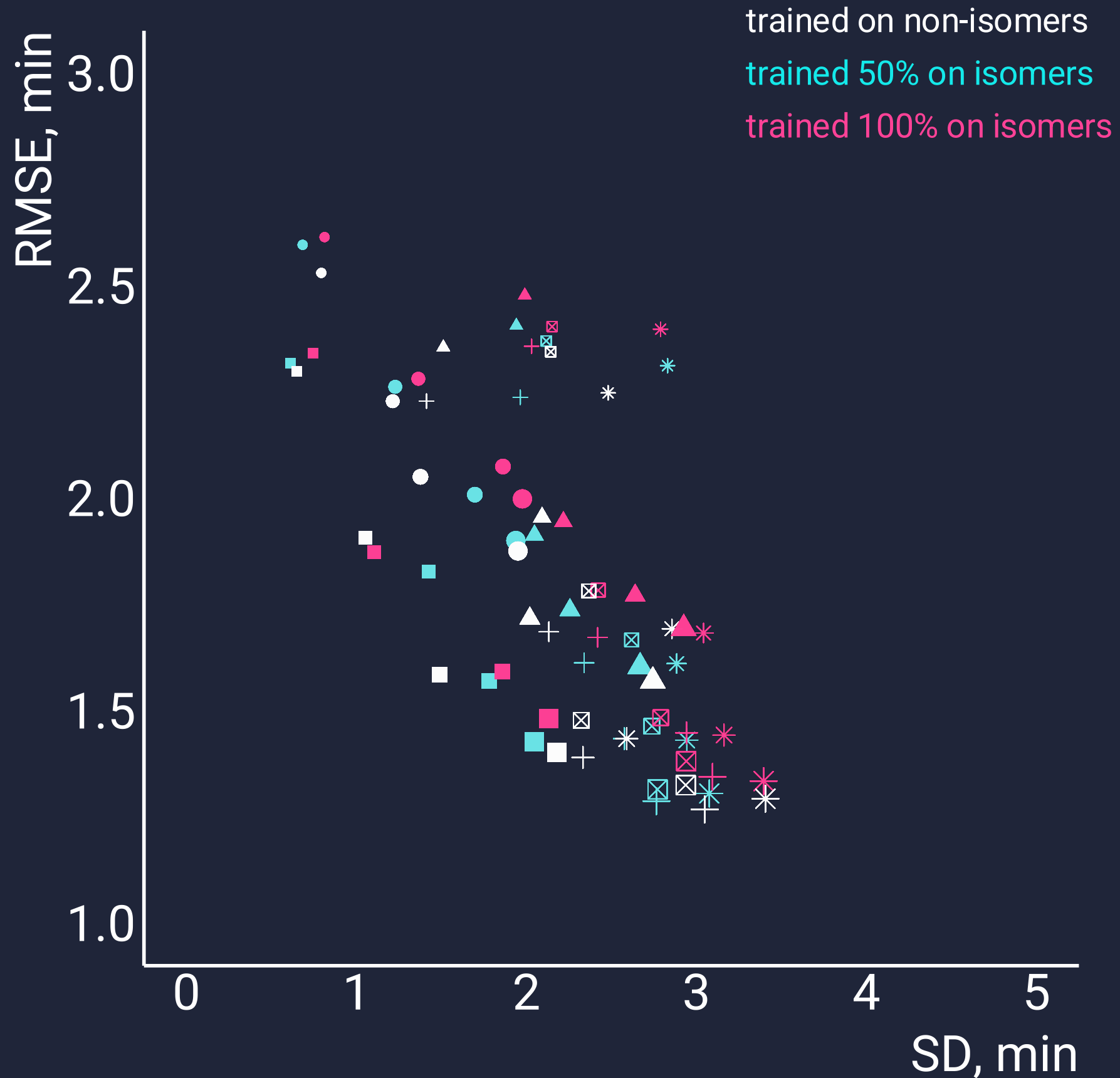


chromatography

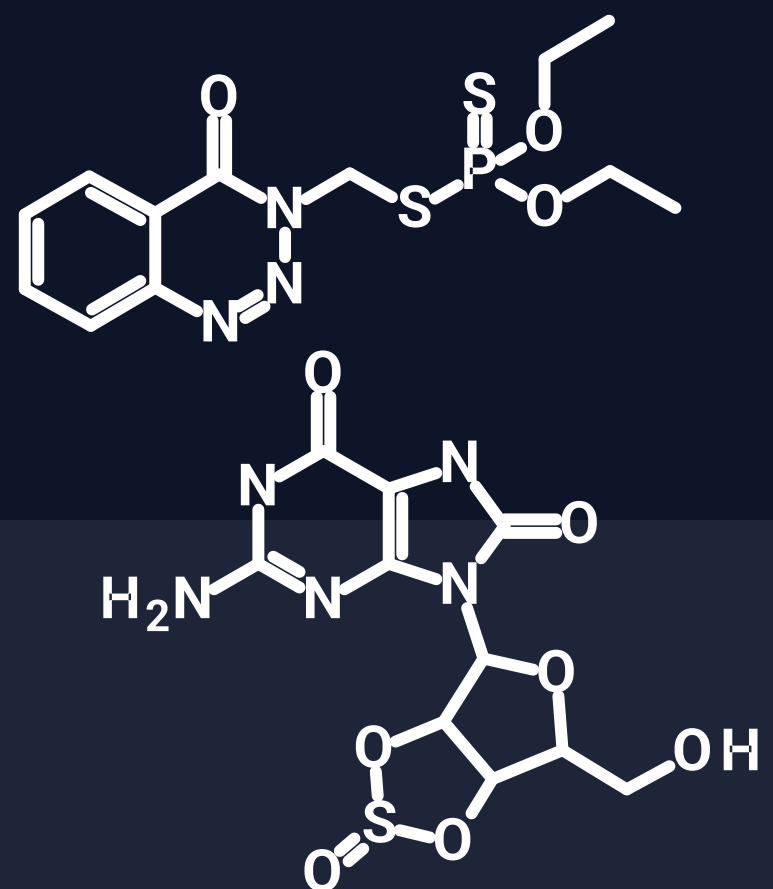
for isomer identification

Kruve et al. in preparation

isomers need to be considered in
choosing RT prediction model!

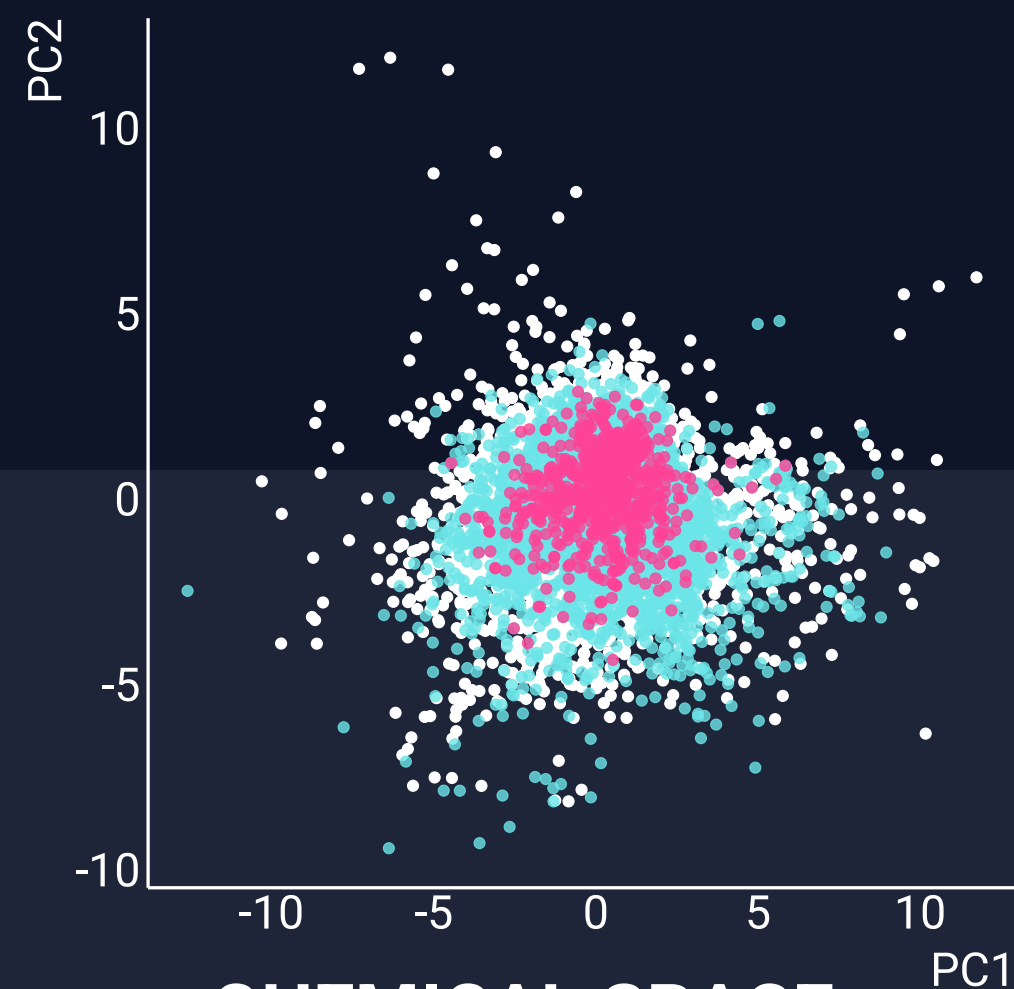


where is the problem



CANDIDATE STRUCTURES

ability to distinguish isomers

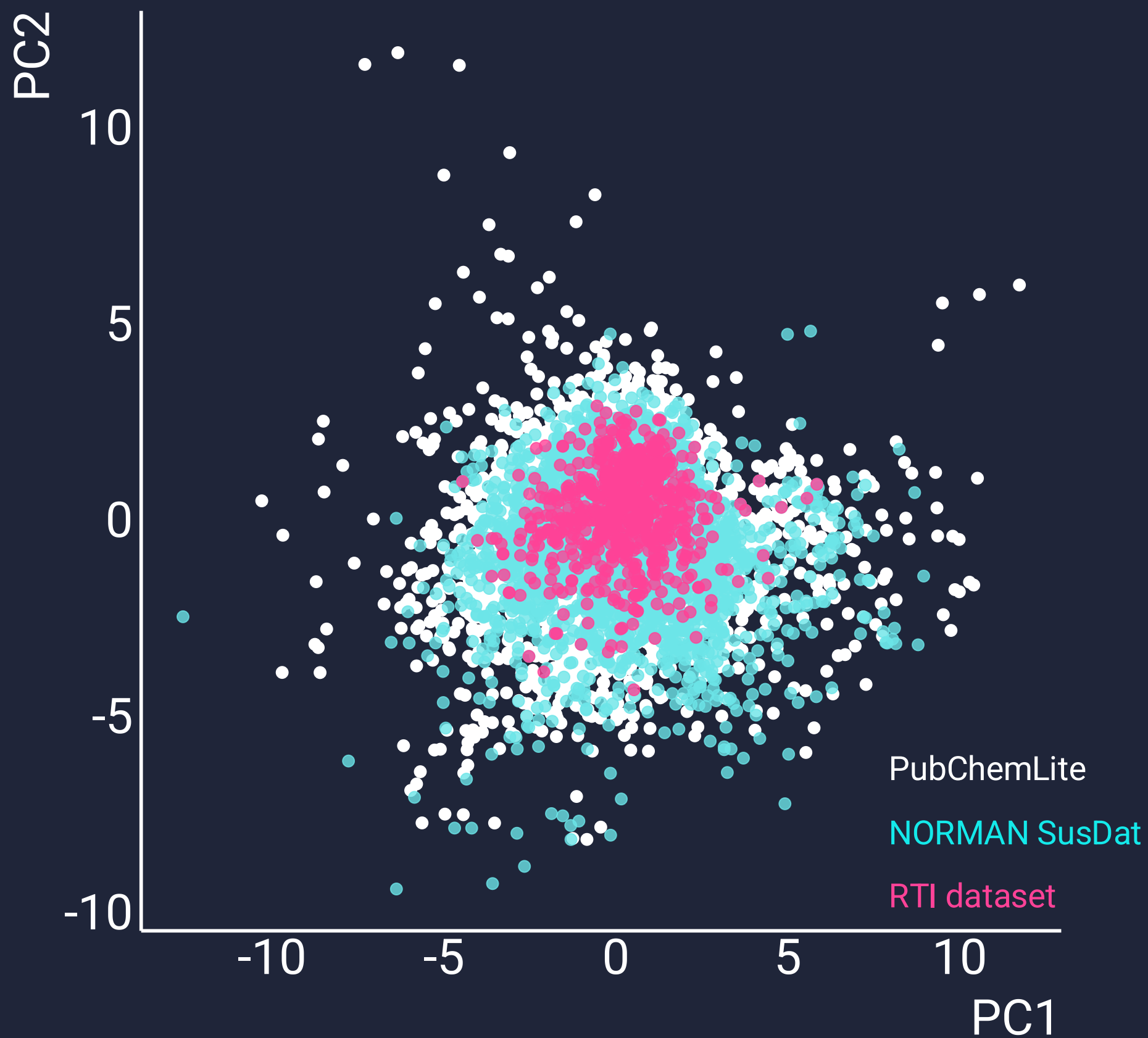


CHEMICAL SPACE

low chemical space coverage

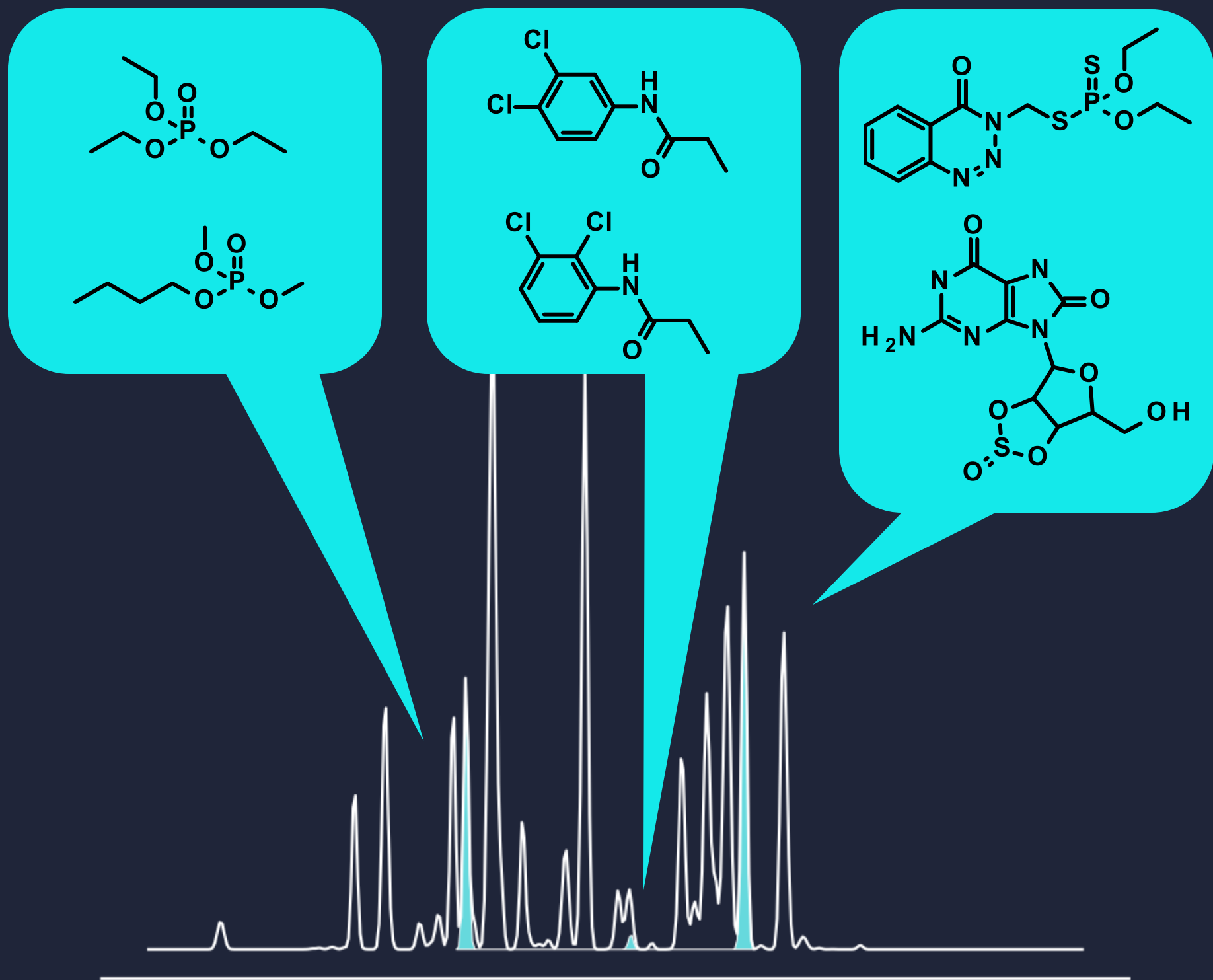
chemical space

covered by prediction algorithms



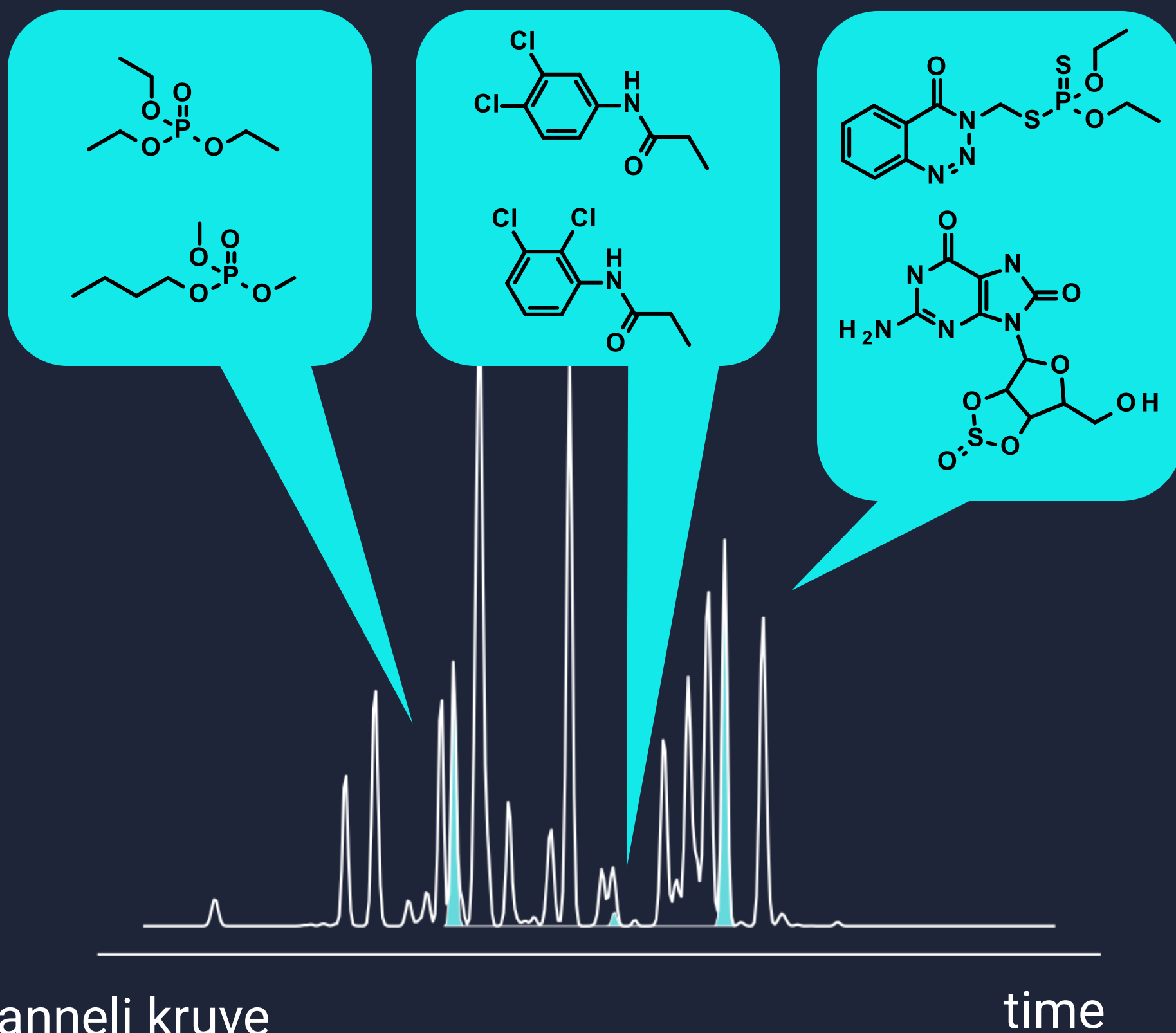
semi-supervised learning for RT

Sandberg et al. in preparation

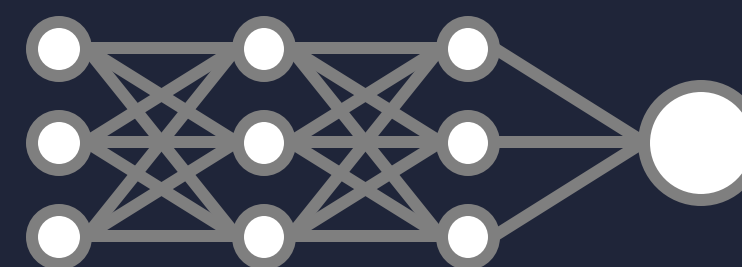


semi-supervised learning for RT

Sandberg et al. in preparation



$\log P$
 pK_a
 $C=O$

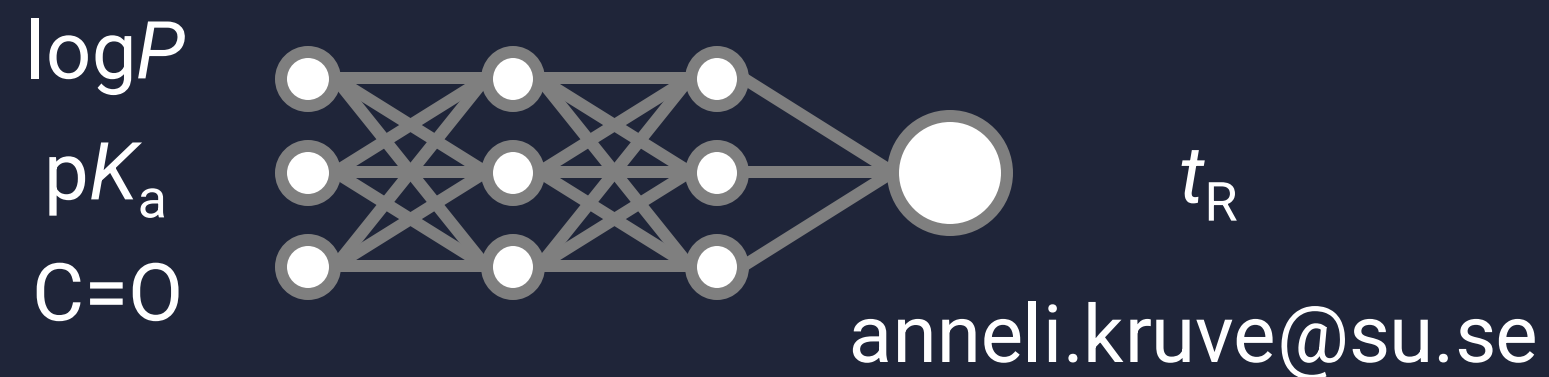
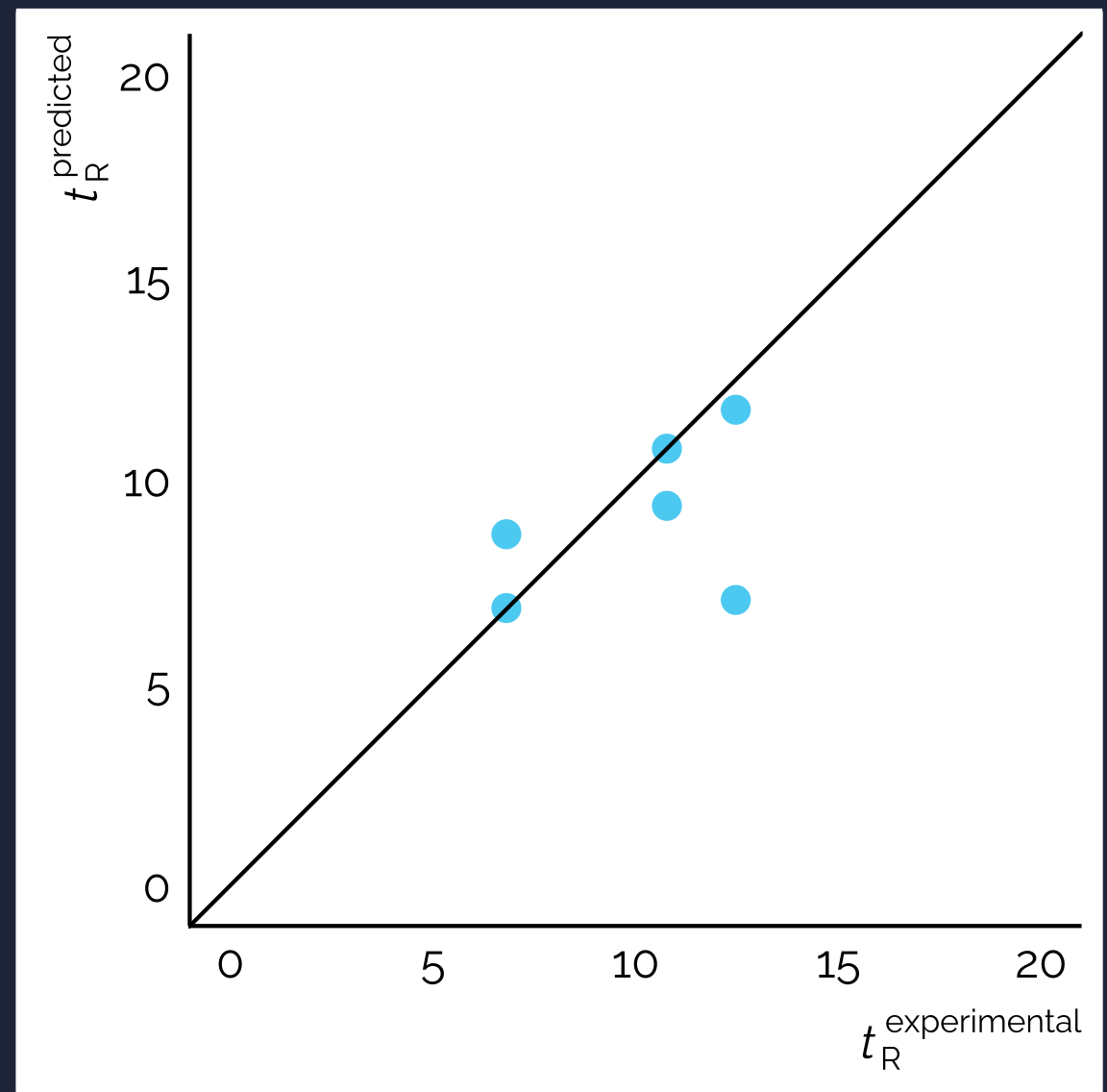
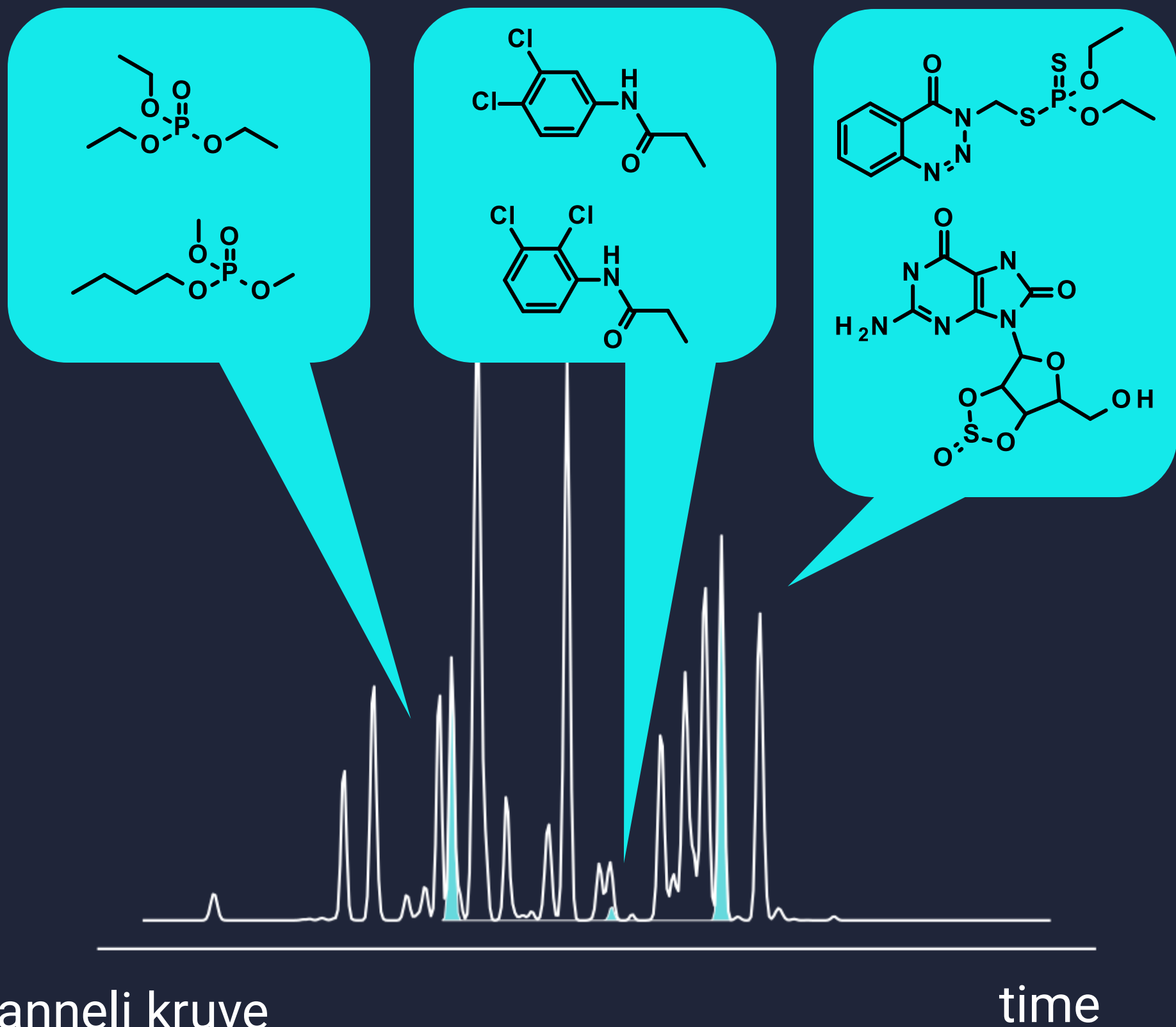


t_R

anneli.kruve@su.se

semi-supervised learning for RT

Sandberg et al. in preparation



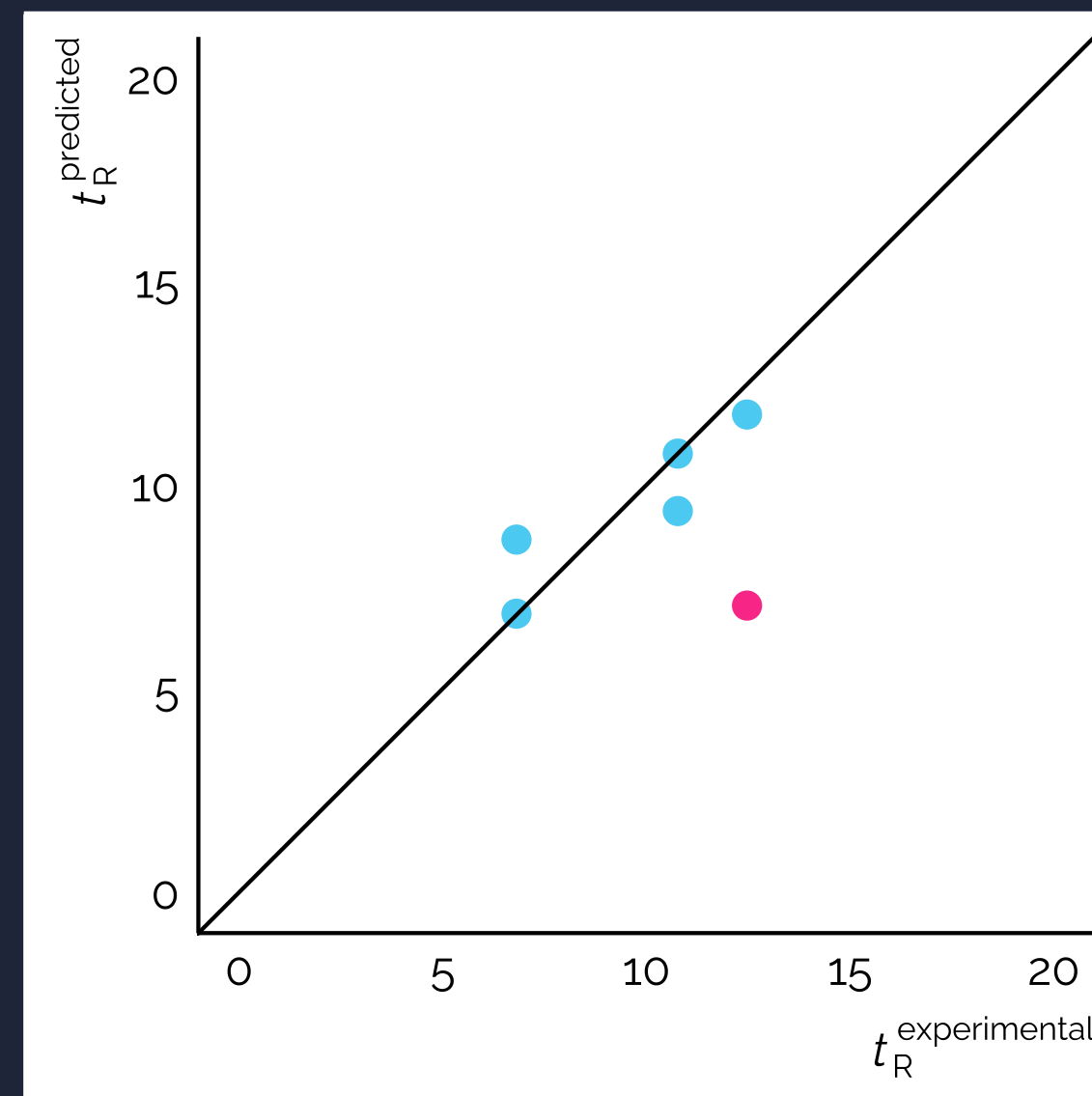
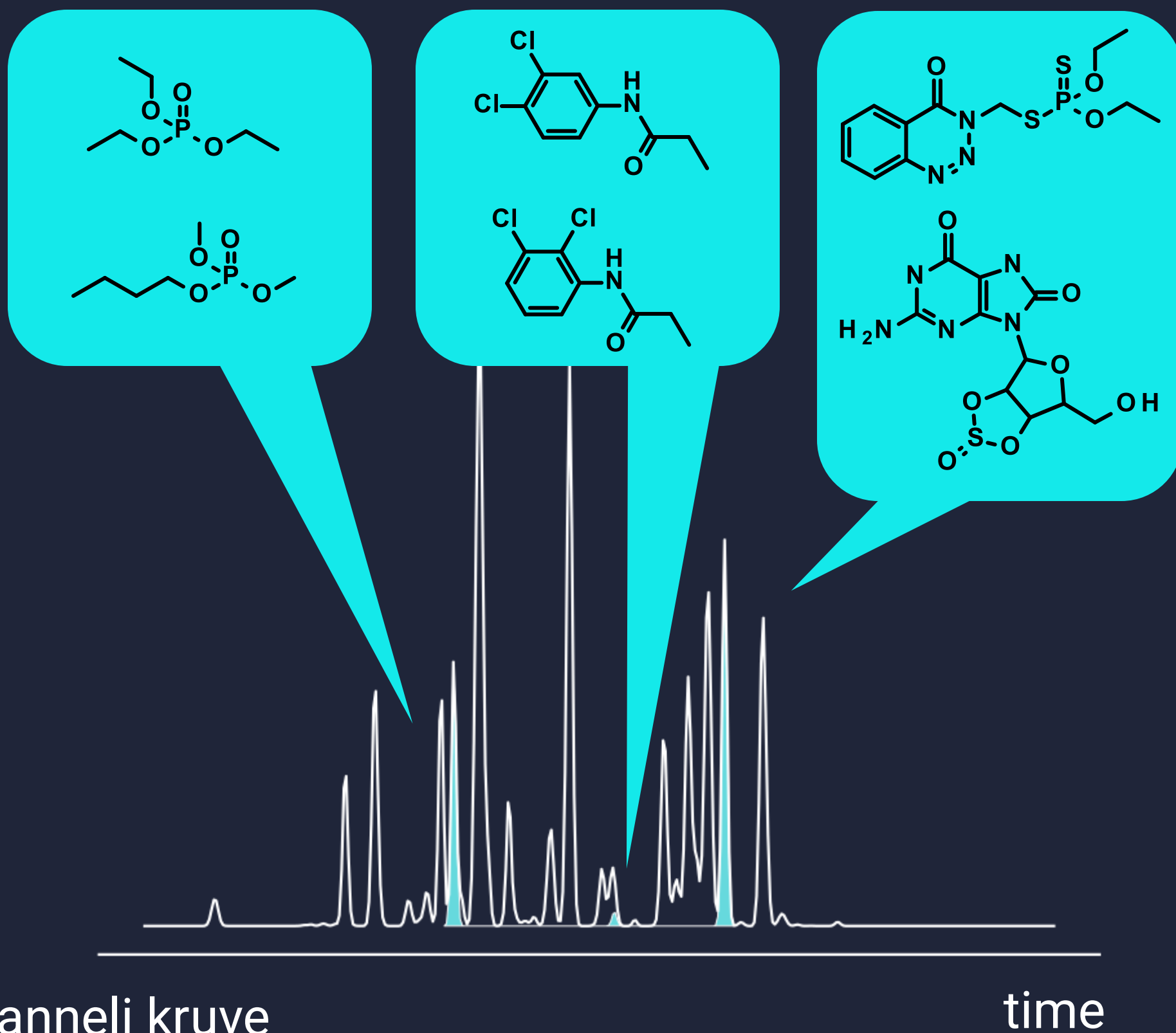
anneli.kruve@su.se

anneli kruve

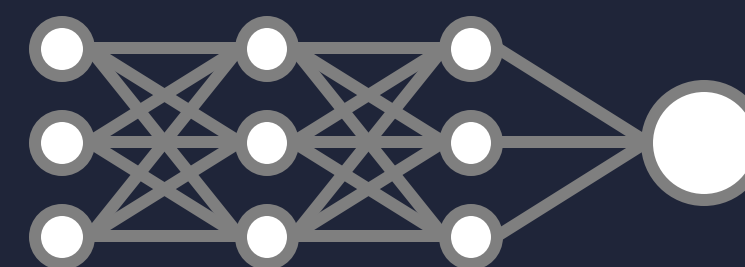
time

semi-supervised learning for RT

Sandberg et al. in preparation



$\log P$
 pK_a
 $C=O$

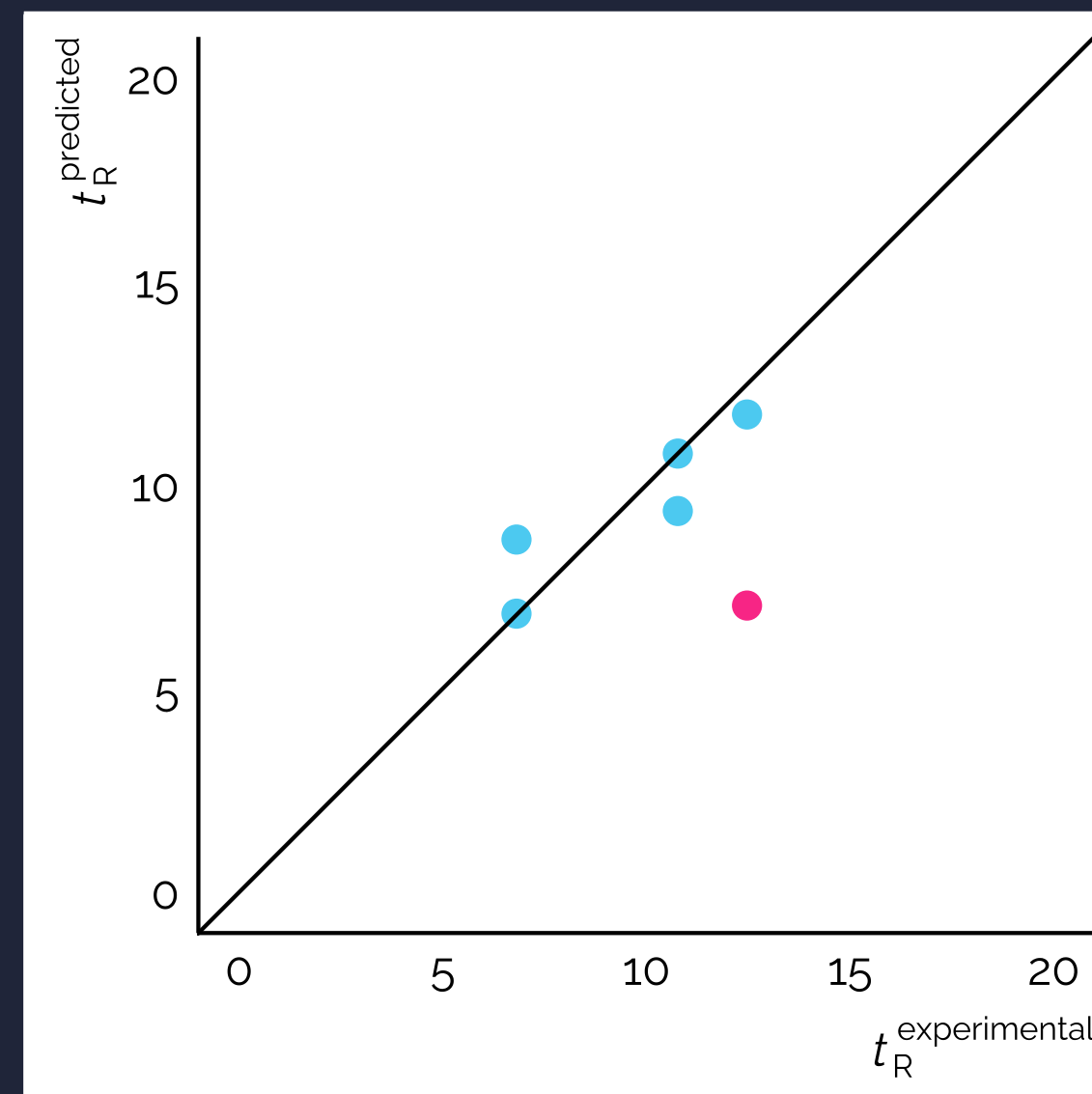
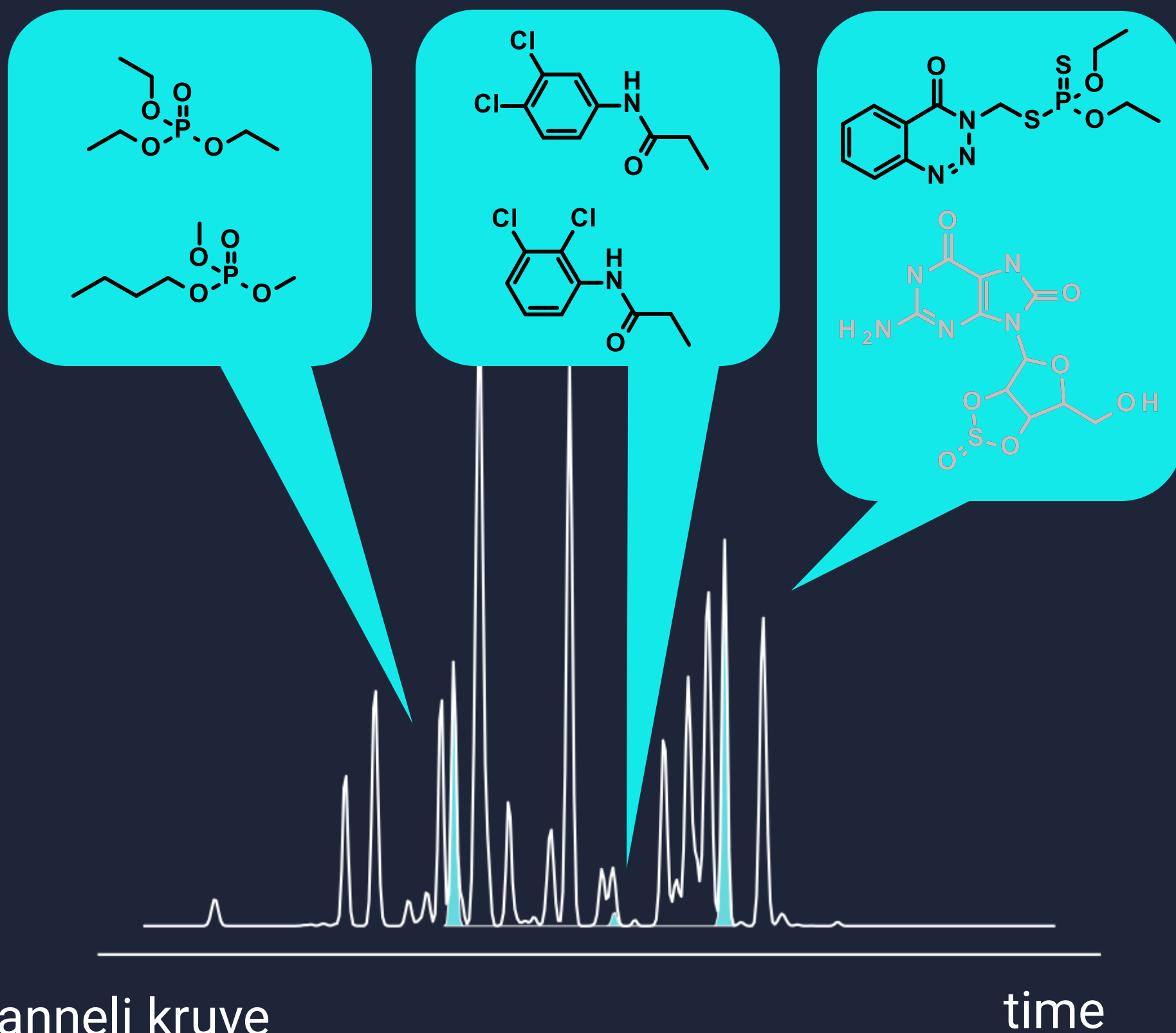


t_R

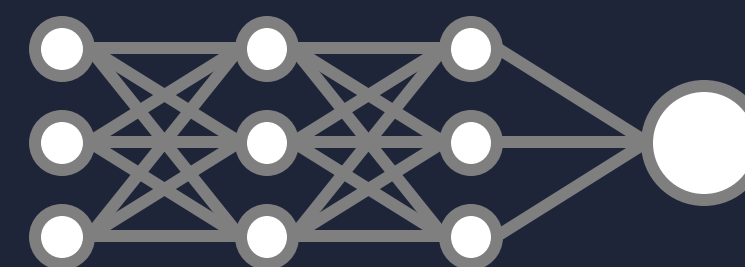
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semi-supervised learning for RT

Sandberg et al. in preparation



$\log P$
 pK_a
 $C=O$

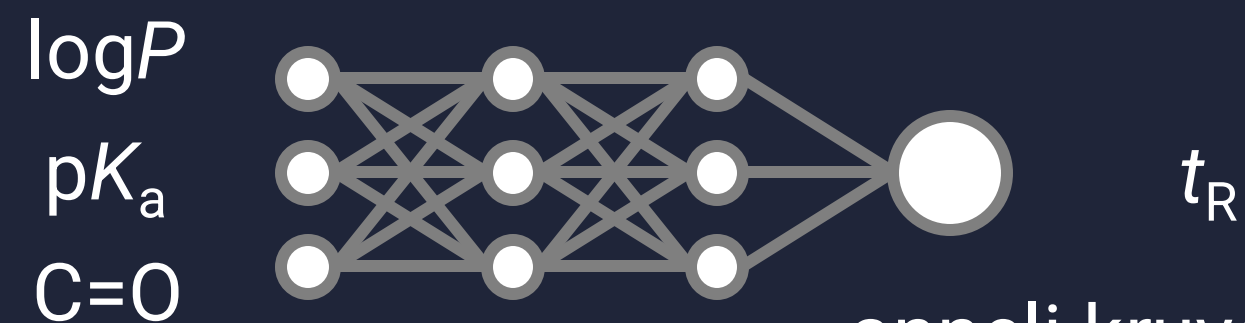
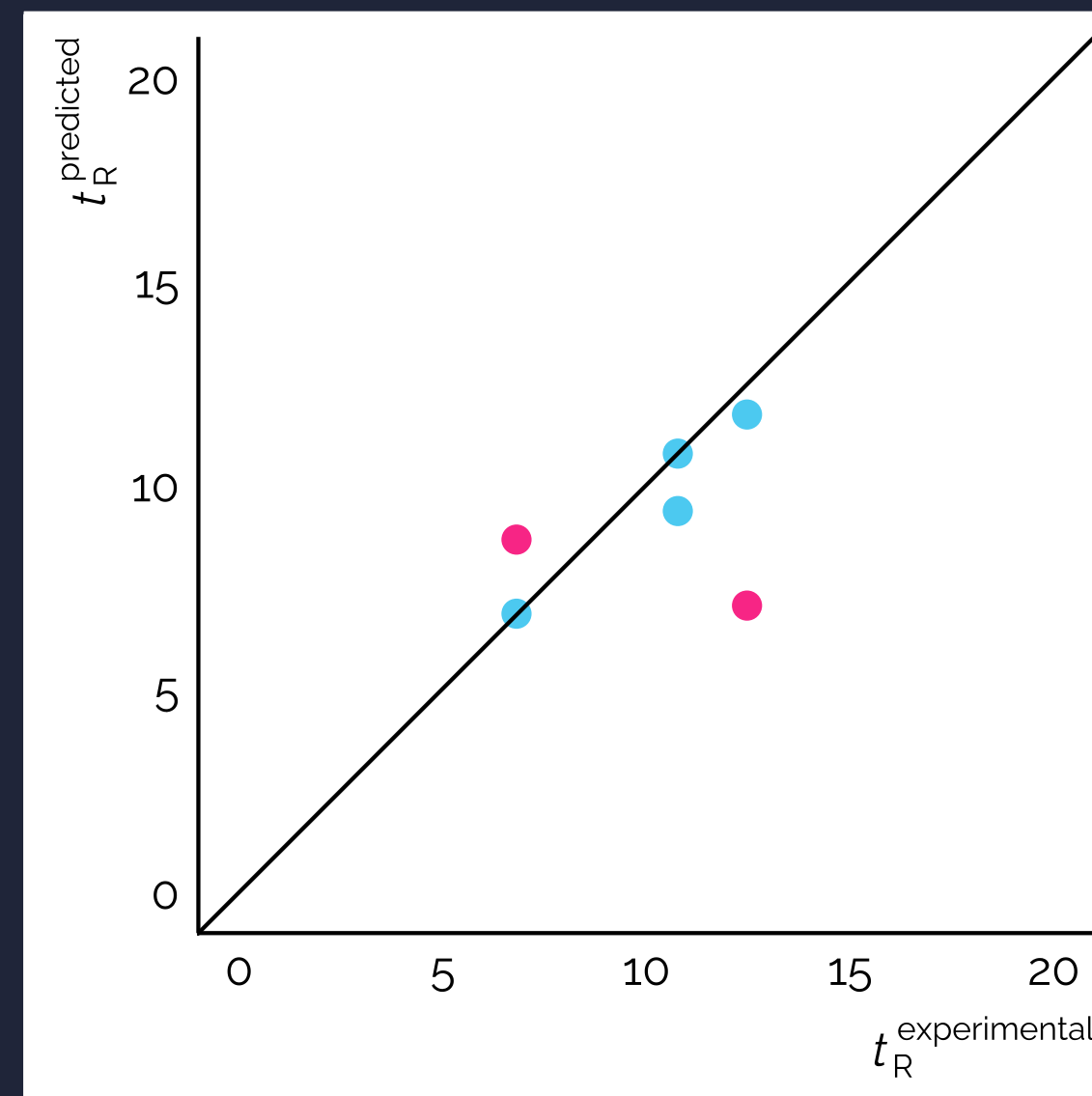
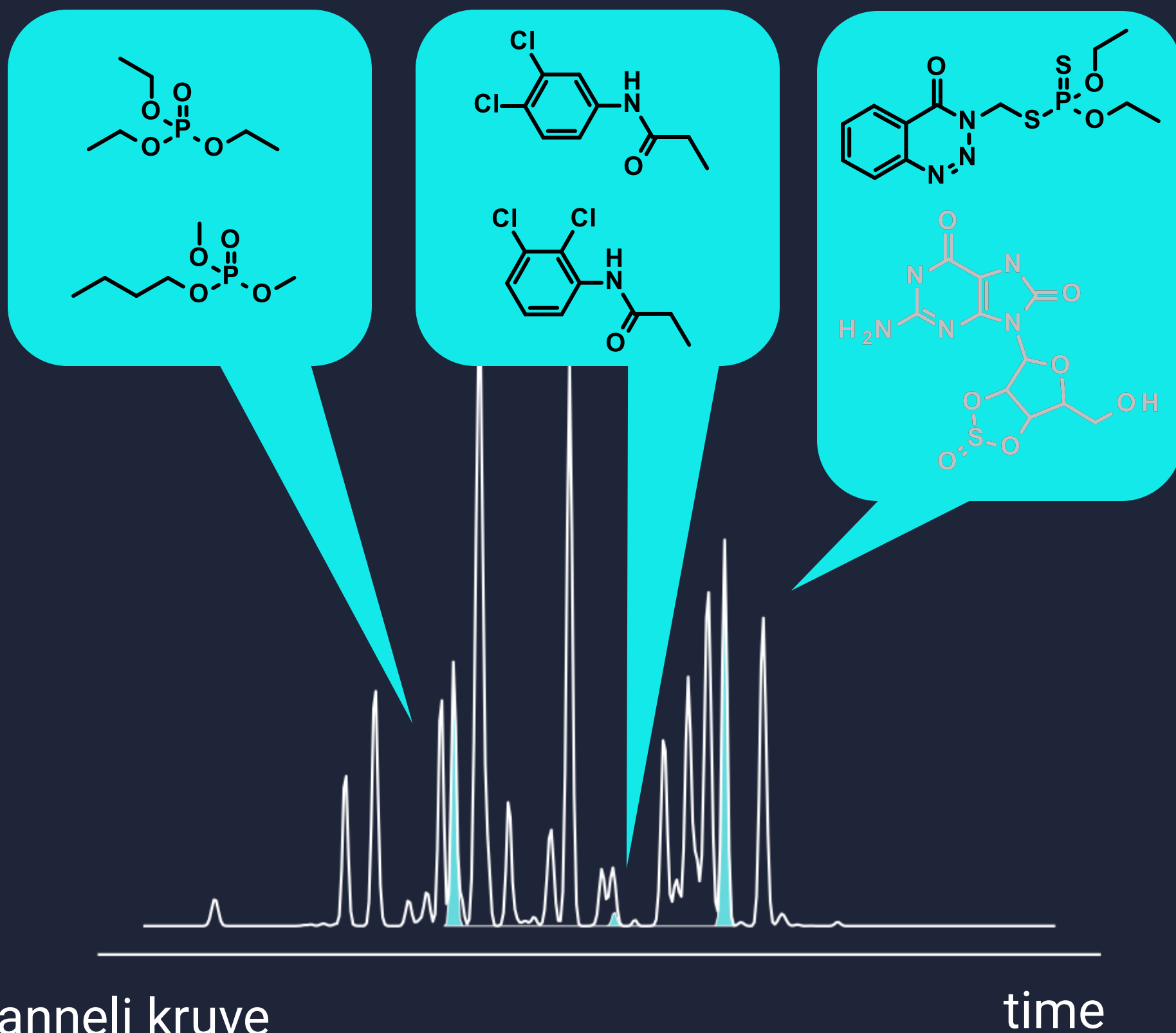


t_R

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semi-supervised learning for RT

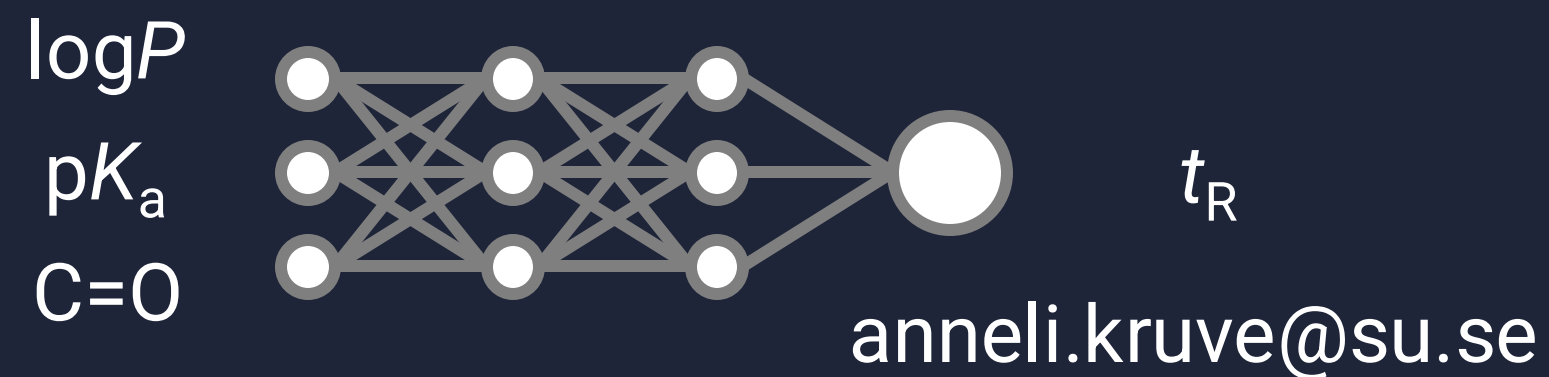
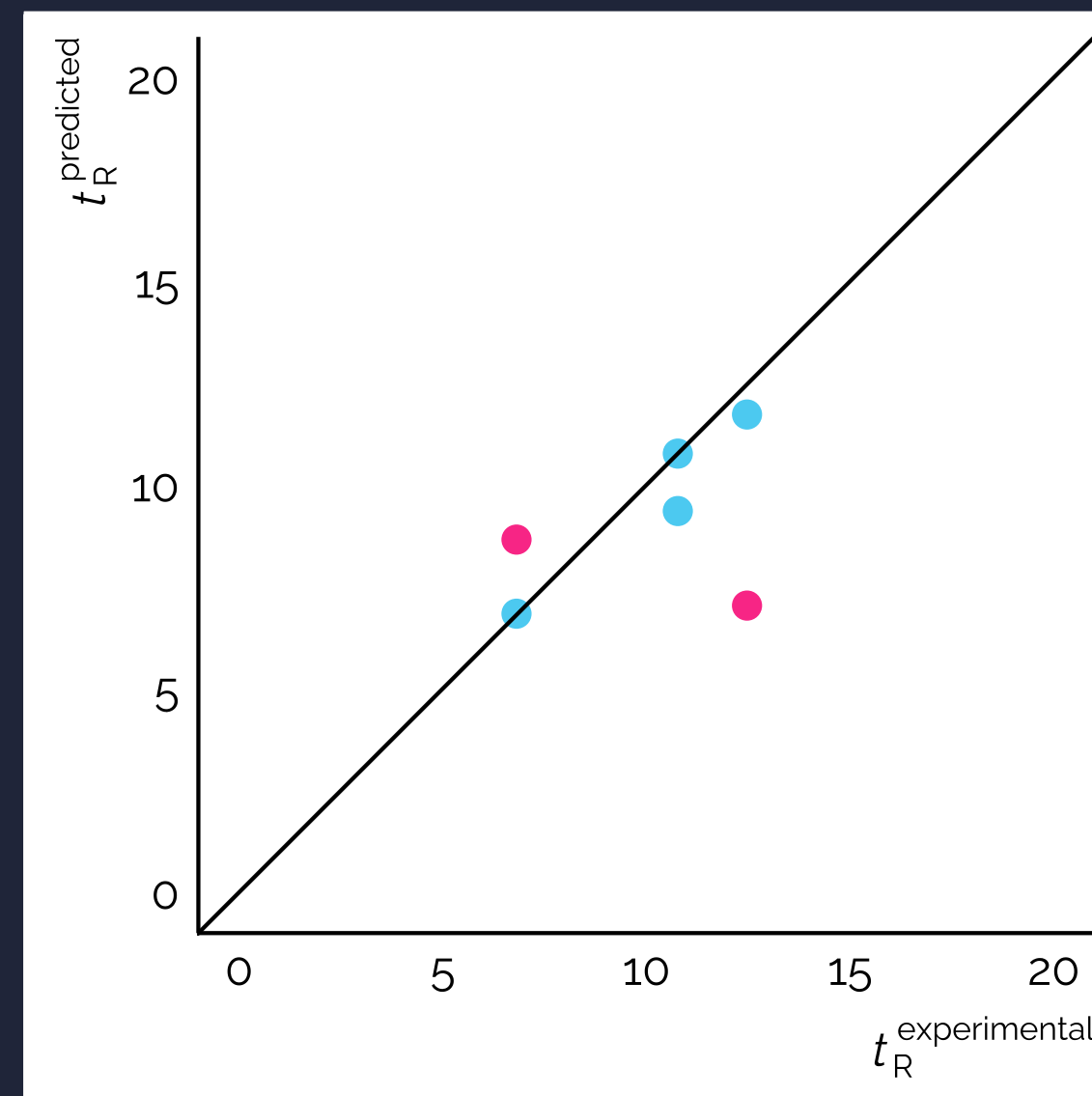
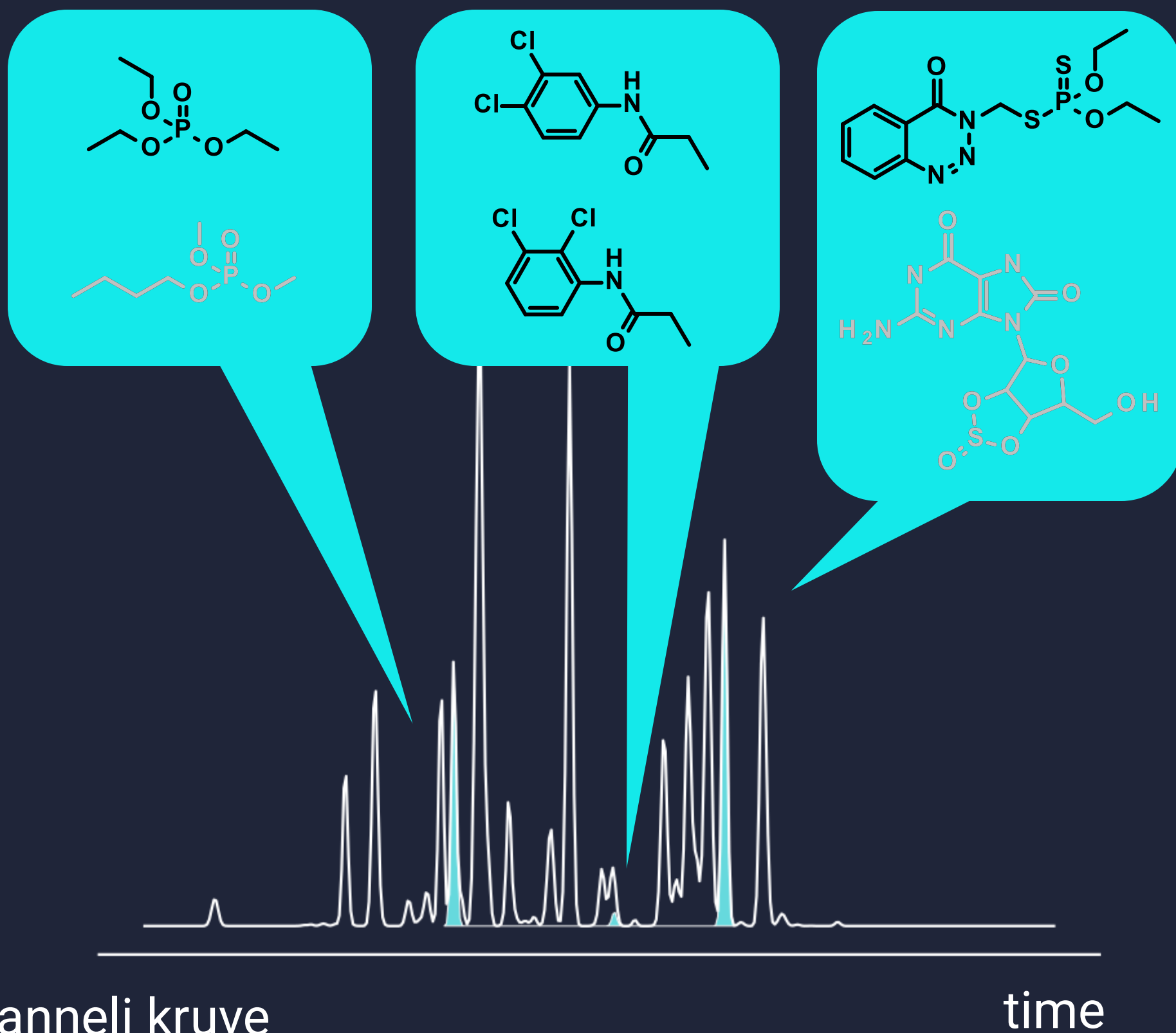
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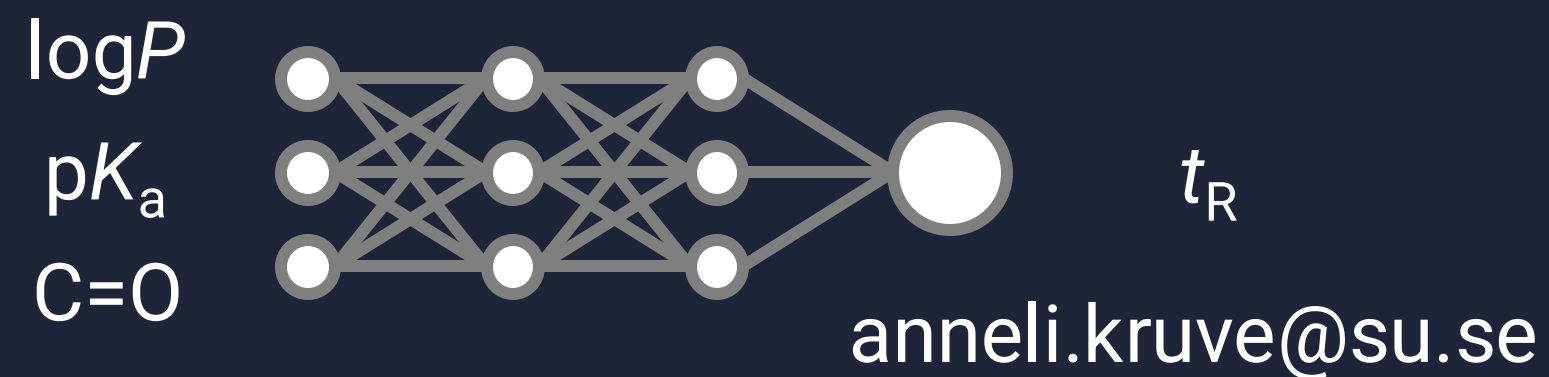
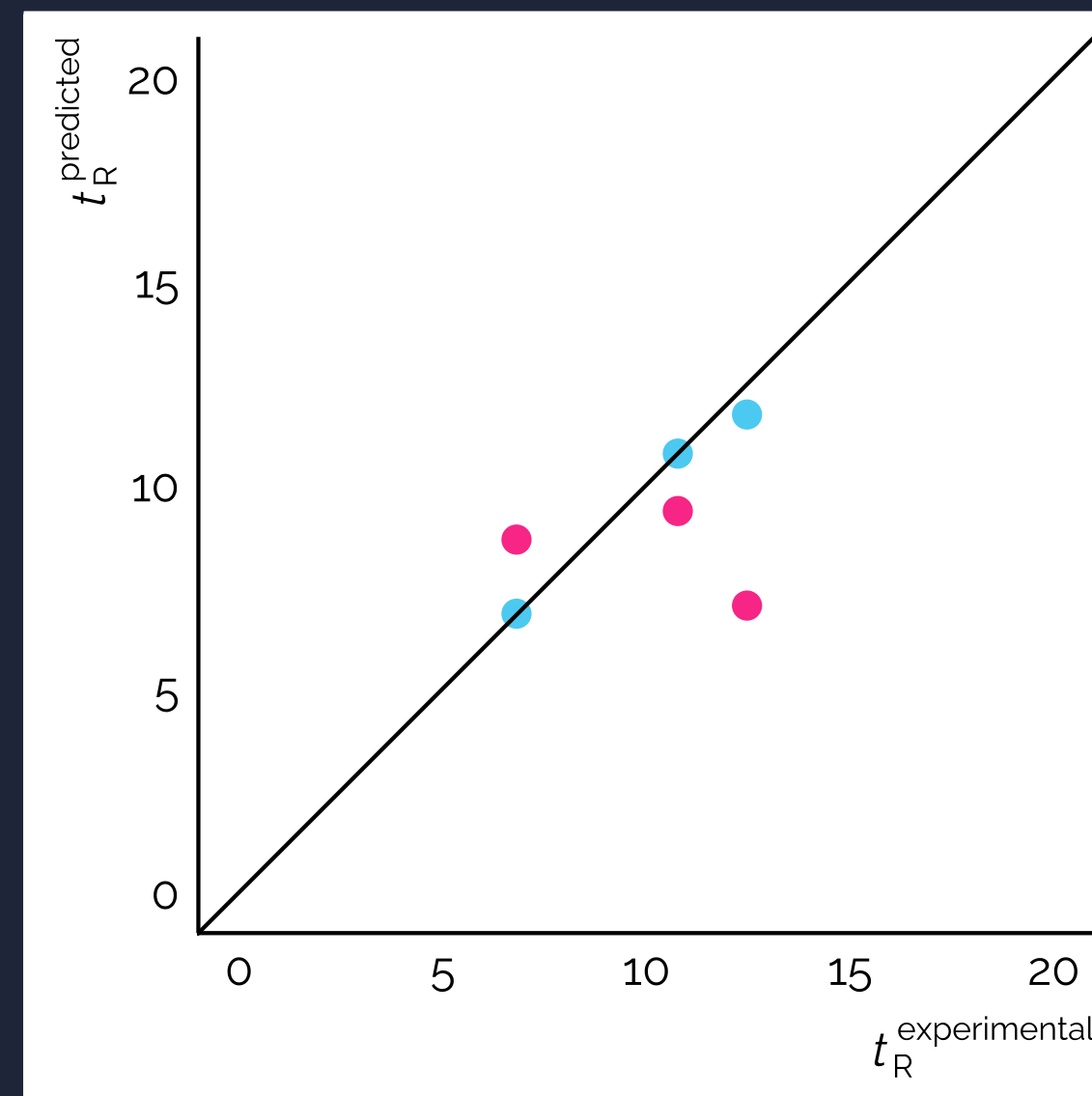
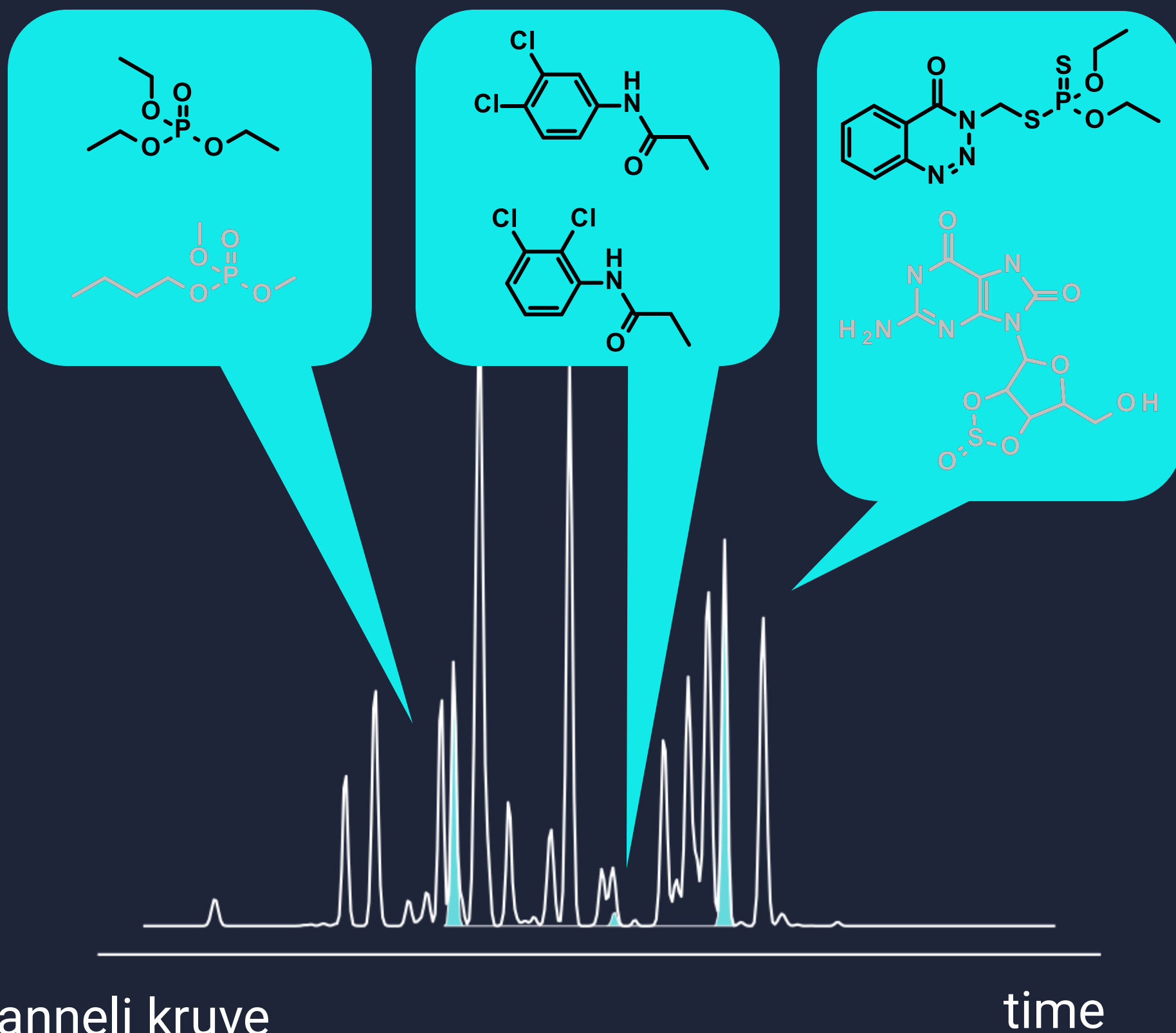
semi-supervised learning for RT

Sandberg et al. in preparation



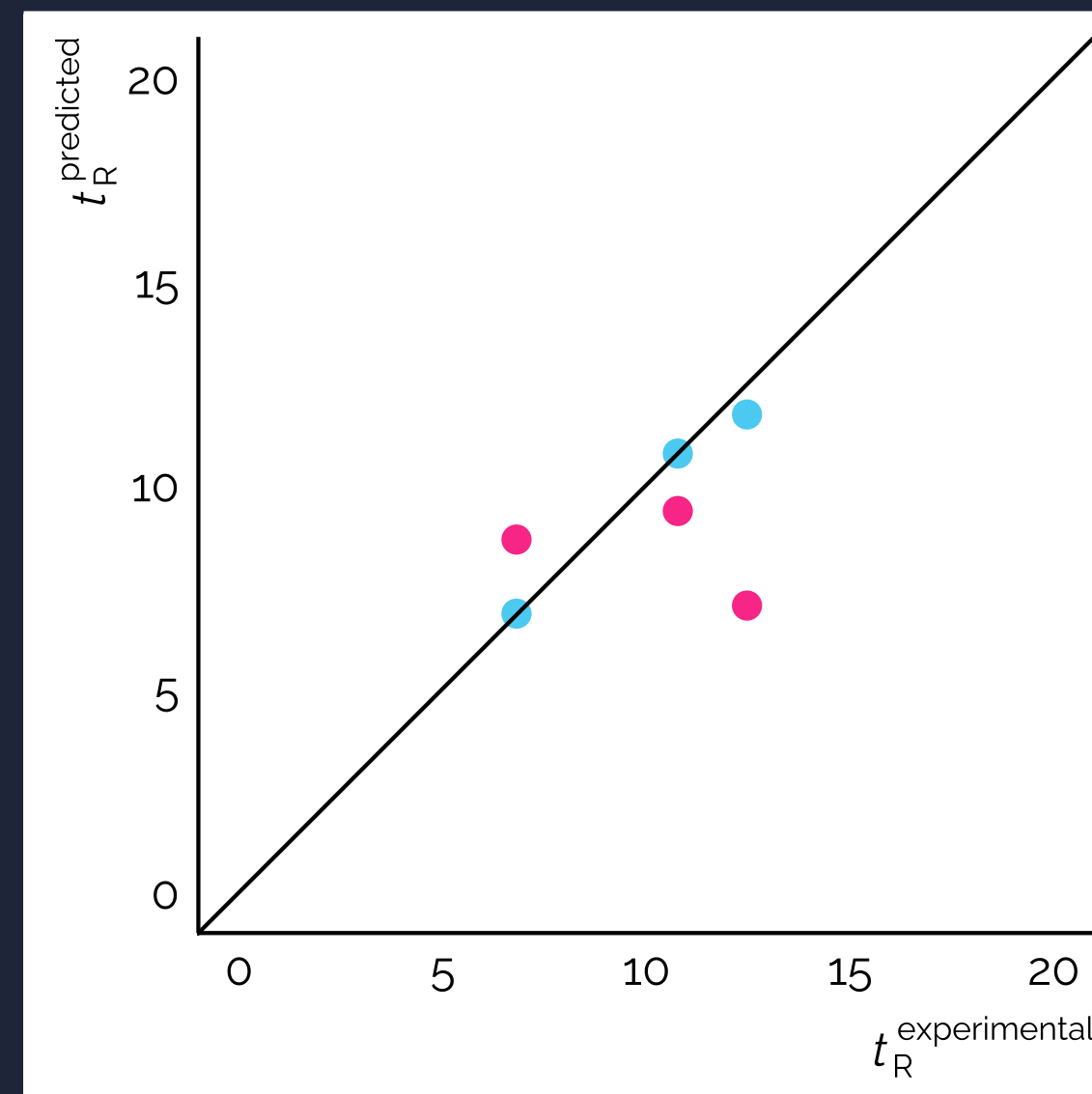
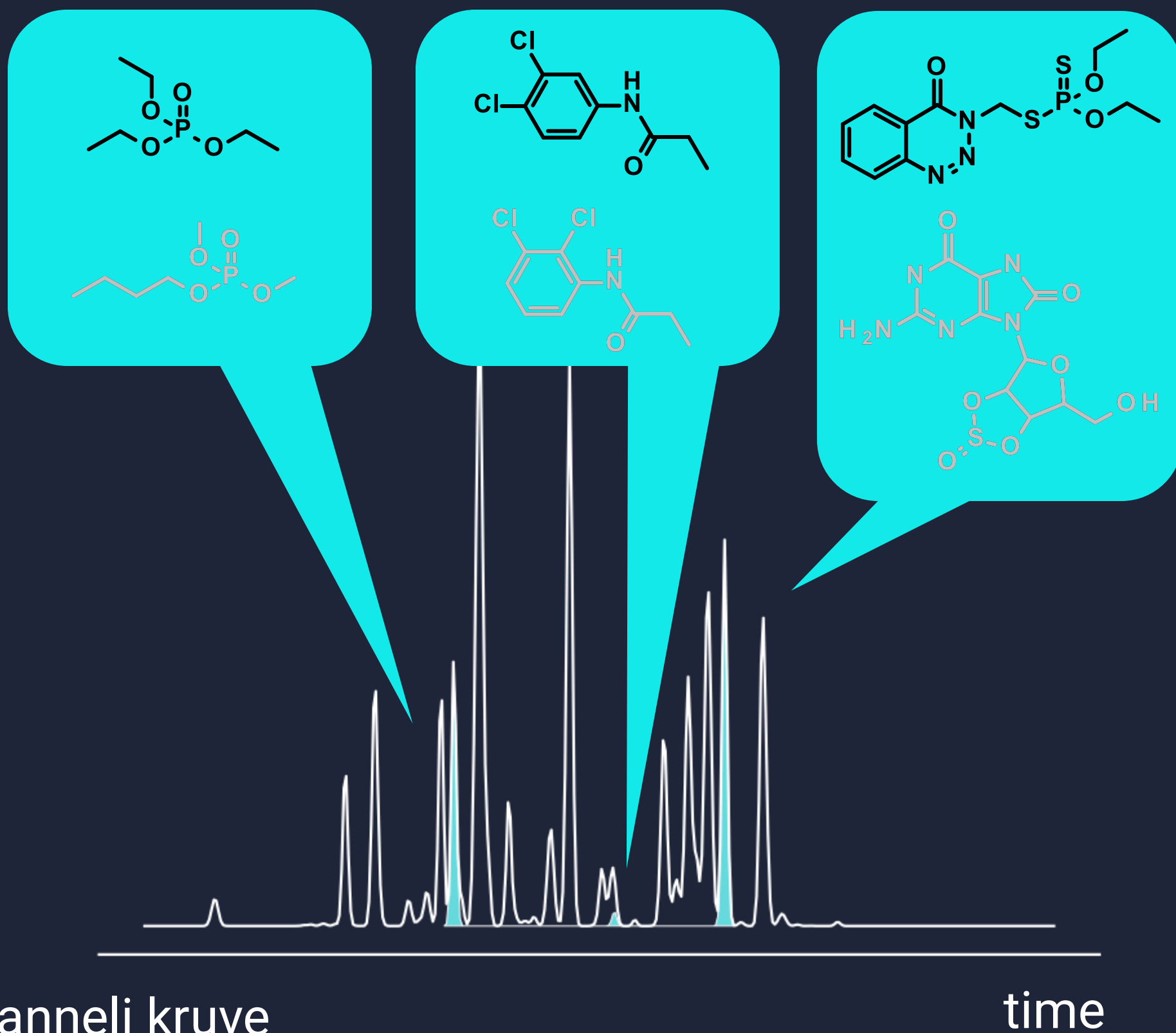
semi-supervised learning for RT

Sandberg et al. in preparation

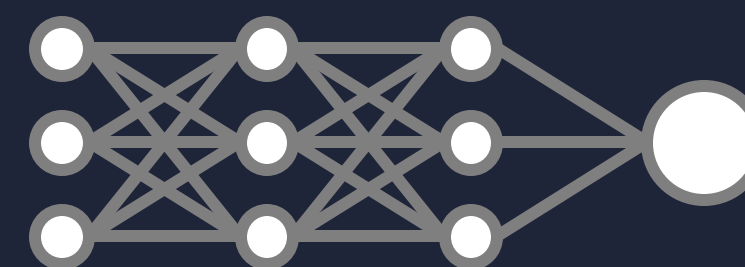


semi-supervised learning for RT

Sandberg et al. in preparation



$\log P$
 pK_a
 $C=O$



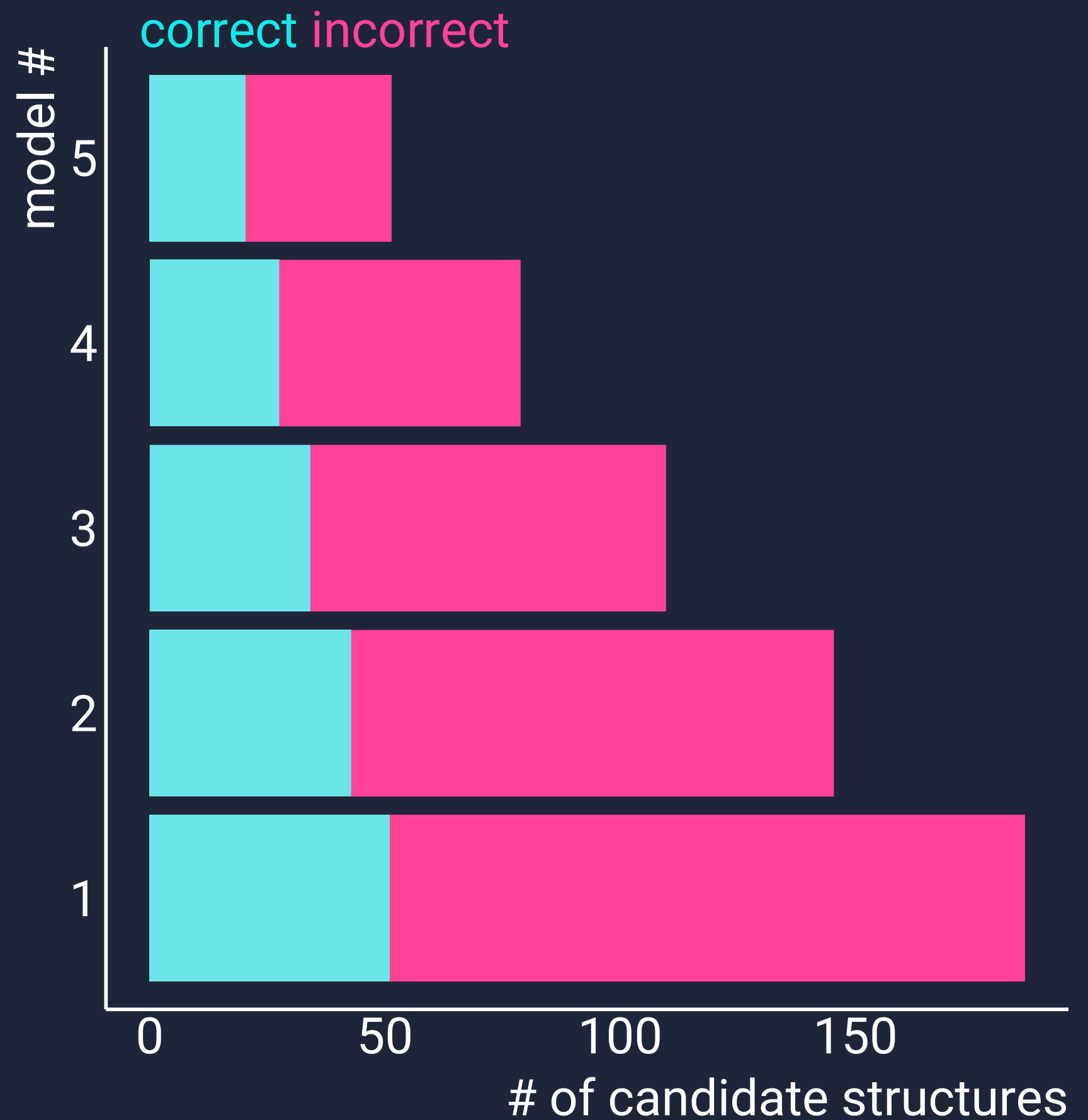
t_R

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candidate structure

evaluation

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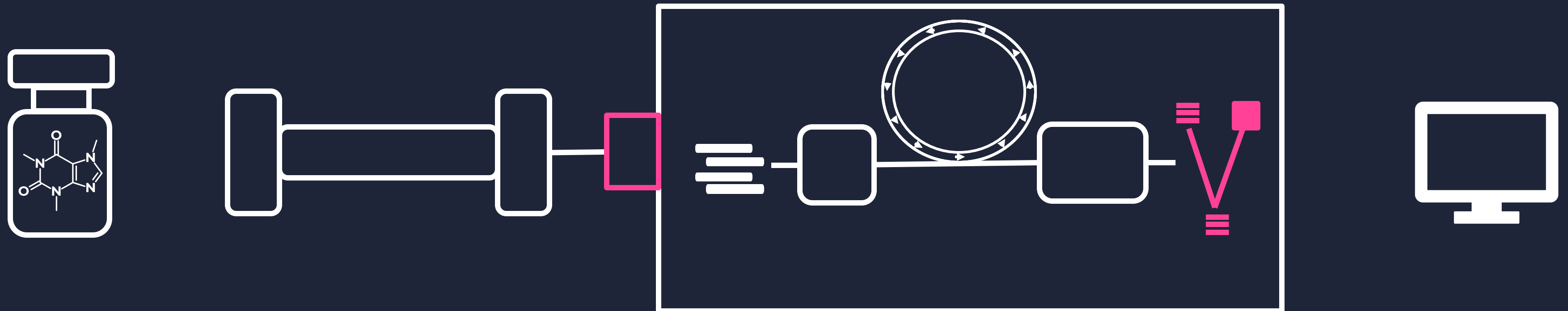


ionization

adduct type & ionization polarity

ionization

Costalunga et al. Anal Chim Acta 2022



adduct type

Costalunga et al. Anal Chim Acta 2022

trans-ferulic acid

in-source
fragment
177.055

[M+H]⁺
195.066

o-dimethyl phthalate

in-source
fragment
163.040

m-dimethyl phthalate

[M+H]⁺
195.066

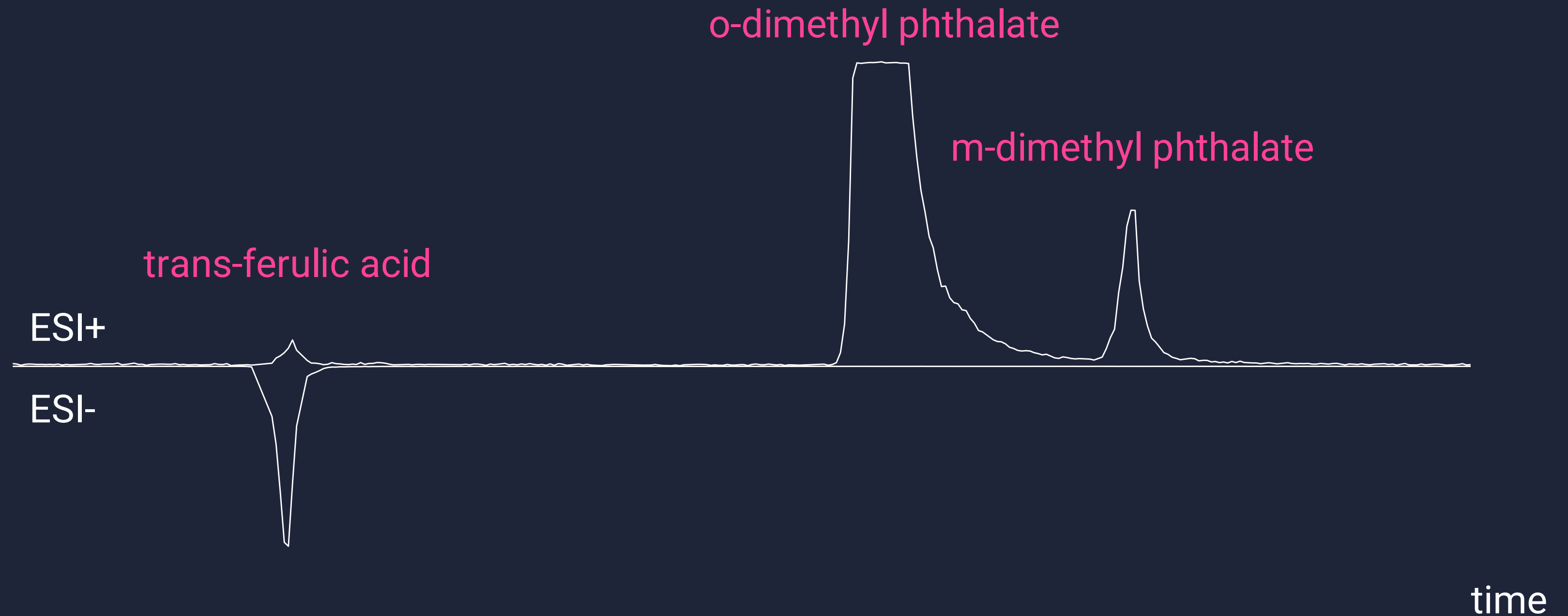
in-source
fragment
163.040

[M+Na]⁺
217.048

m/z

ionization mode

Hupatz, Rahu, et al. Anal Bioanal Chem 2024

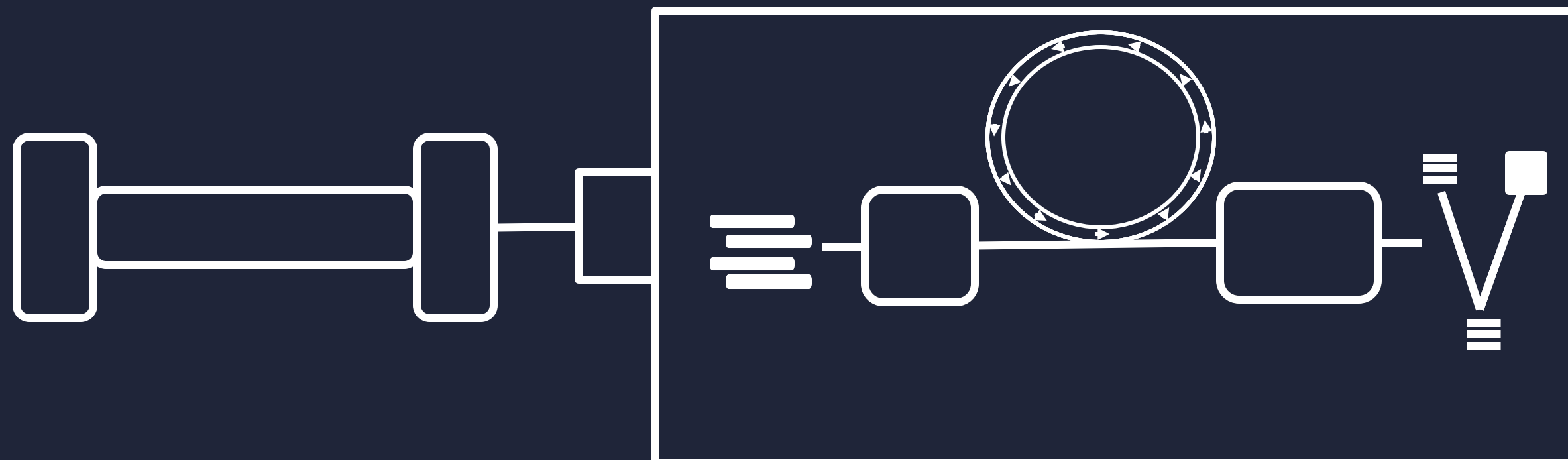
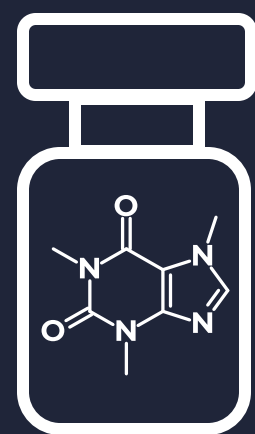


combining

separation & MS²

orthogonal separation

Akhlaqi et al. Anal Bioanal Chem 2023



14
isomeric
chemicals

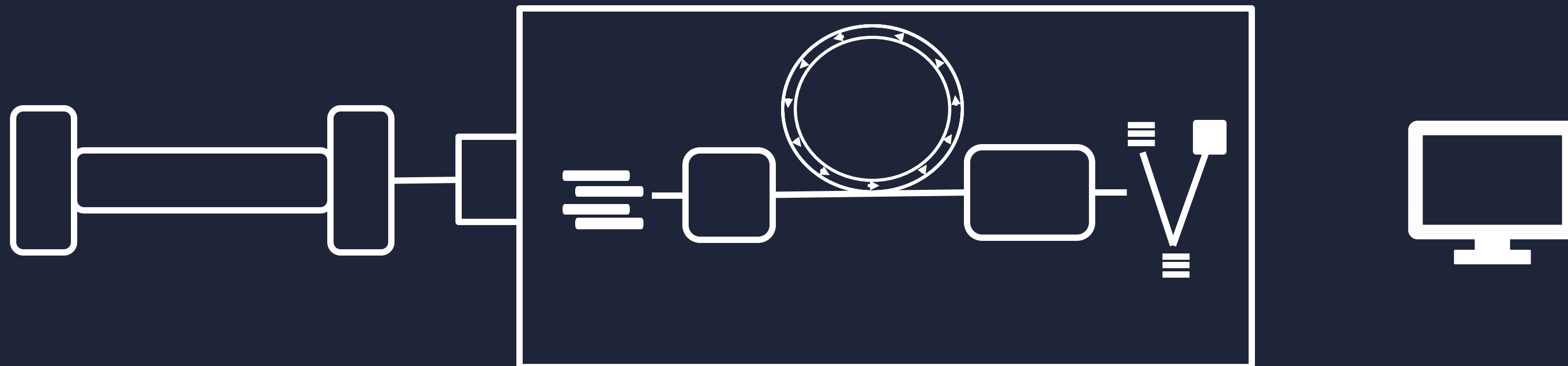
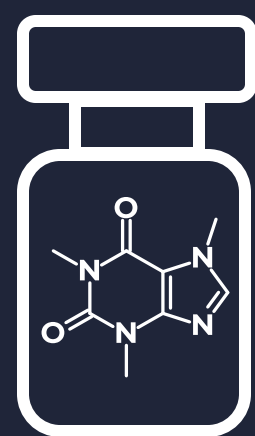
C₁₈ RP

Cyclic IMS
&
MS² with ToF

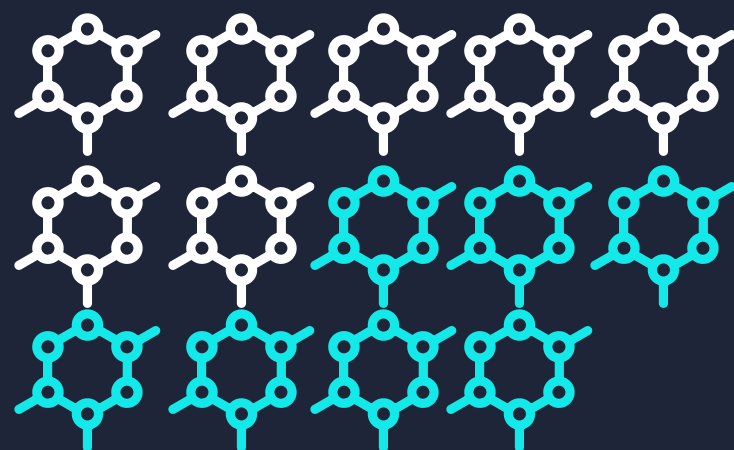
SIRIUS+
CSI:FingerID
&
CFM-ID

orthogonal separation

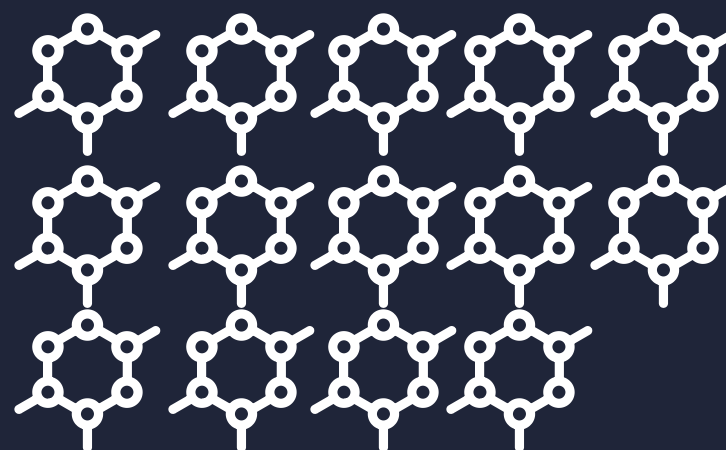
Akhlaqi et al. Anal Bioanal Chem 2023



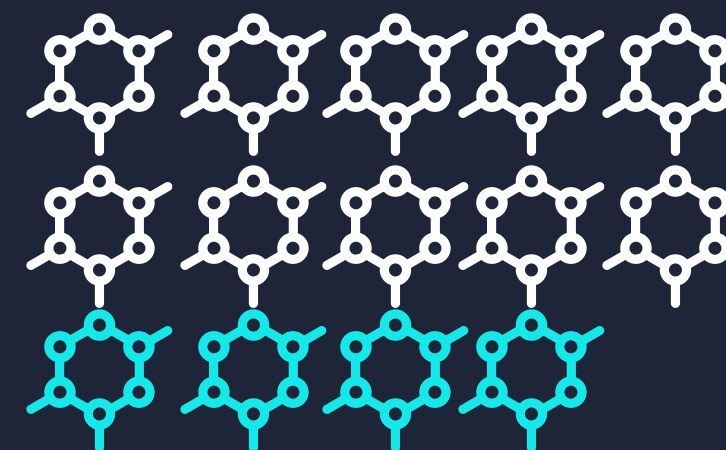
t_R



CCS



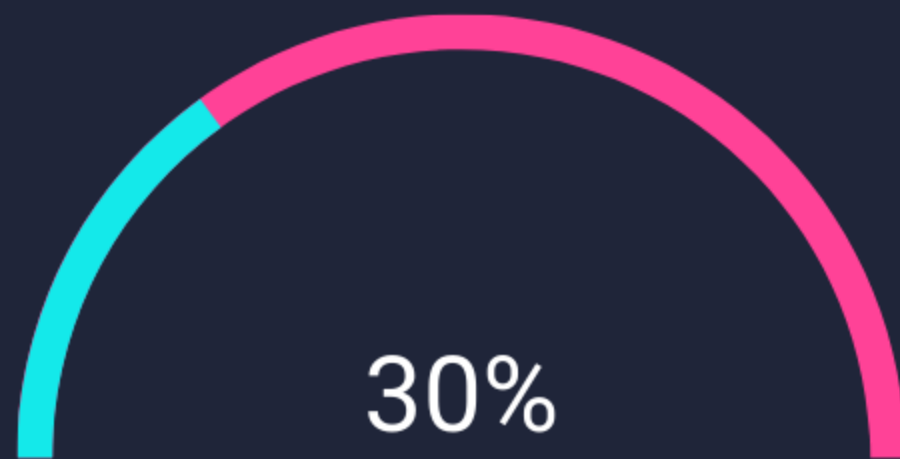
MS²



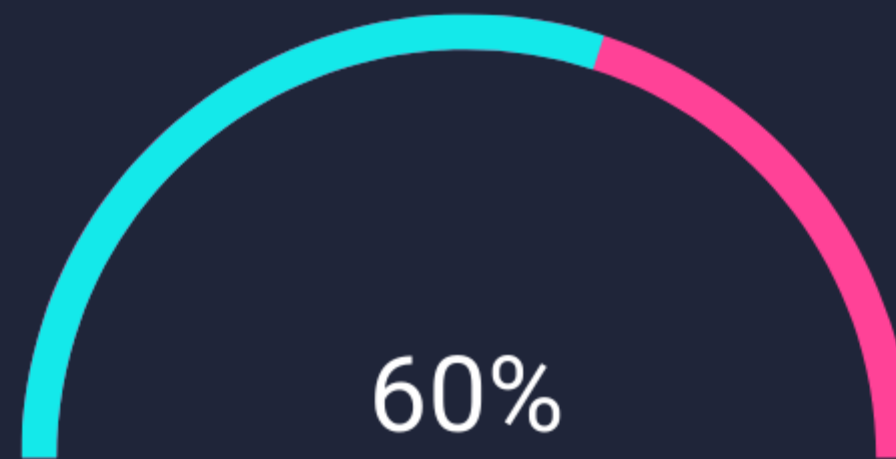
machine learning

for prioritization and identification in nontarget screening

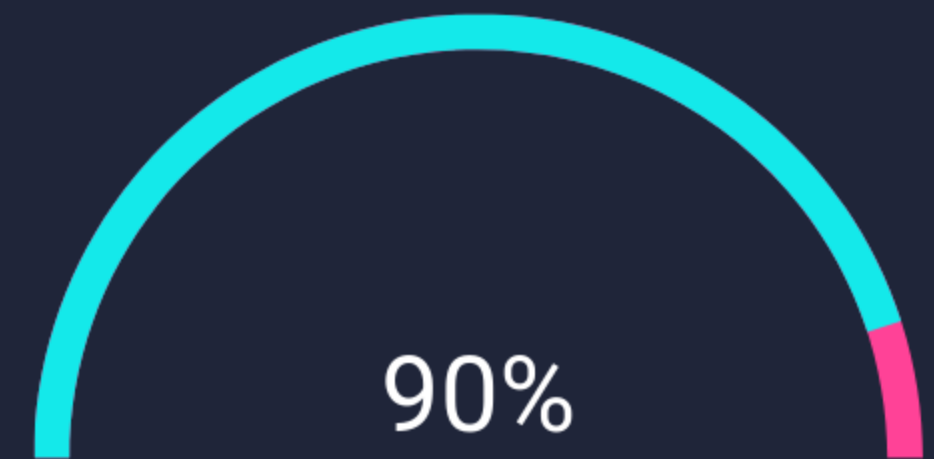
TOXICITY



QUANTIFICATION



RETENTION TIME



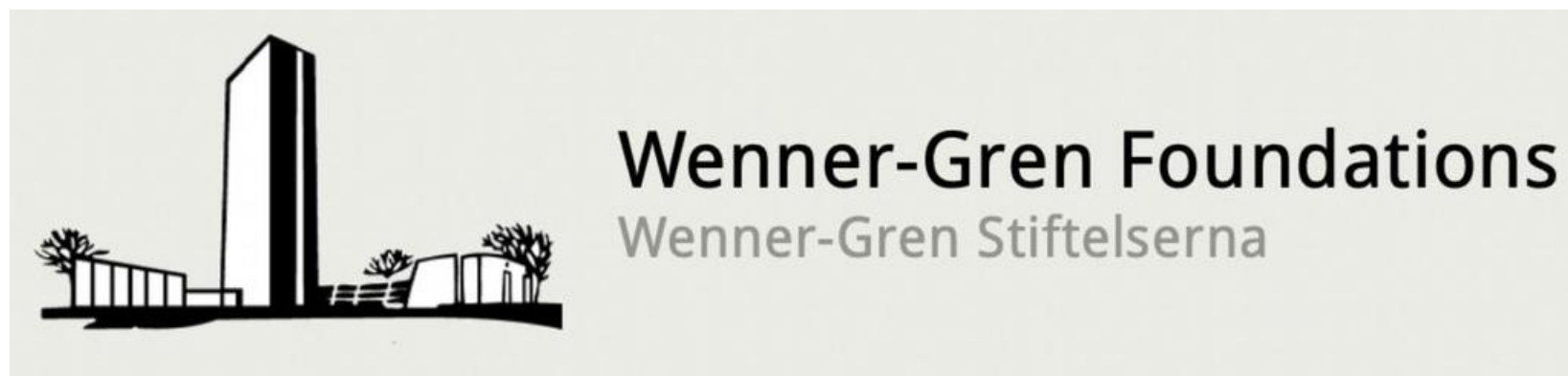


Swedish
Research
Council

FORMAS 

The logo for FORMAS includes the word "FORMAS" in a large, bold, black, sans-serif font. To the right of the text is a small icon consisting of five colored dots (yellow, red, blue, green, and orange) arranged in a slightly irregular grid.

Wallenberg Initiative
Materials Science
for Sustainability



CARL TRYGGERS
STIFTELSE
FÖR VETENSKAPLIG FORSKNING

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*Exploring the
research space...*

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