



detection and identification of high-risk chemicals with LC/HMRS and machine learning

anneli kruve

SVERIGE

Ökat hot mot svenskt dricksvatten – brott mer än var femte dag

Uppdaterad 2024-12-02 Publicerad 2024-12-01



Swedish drinking water

an offence more than every fifth day

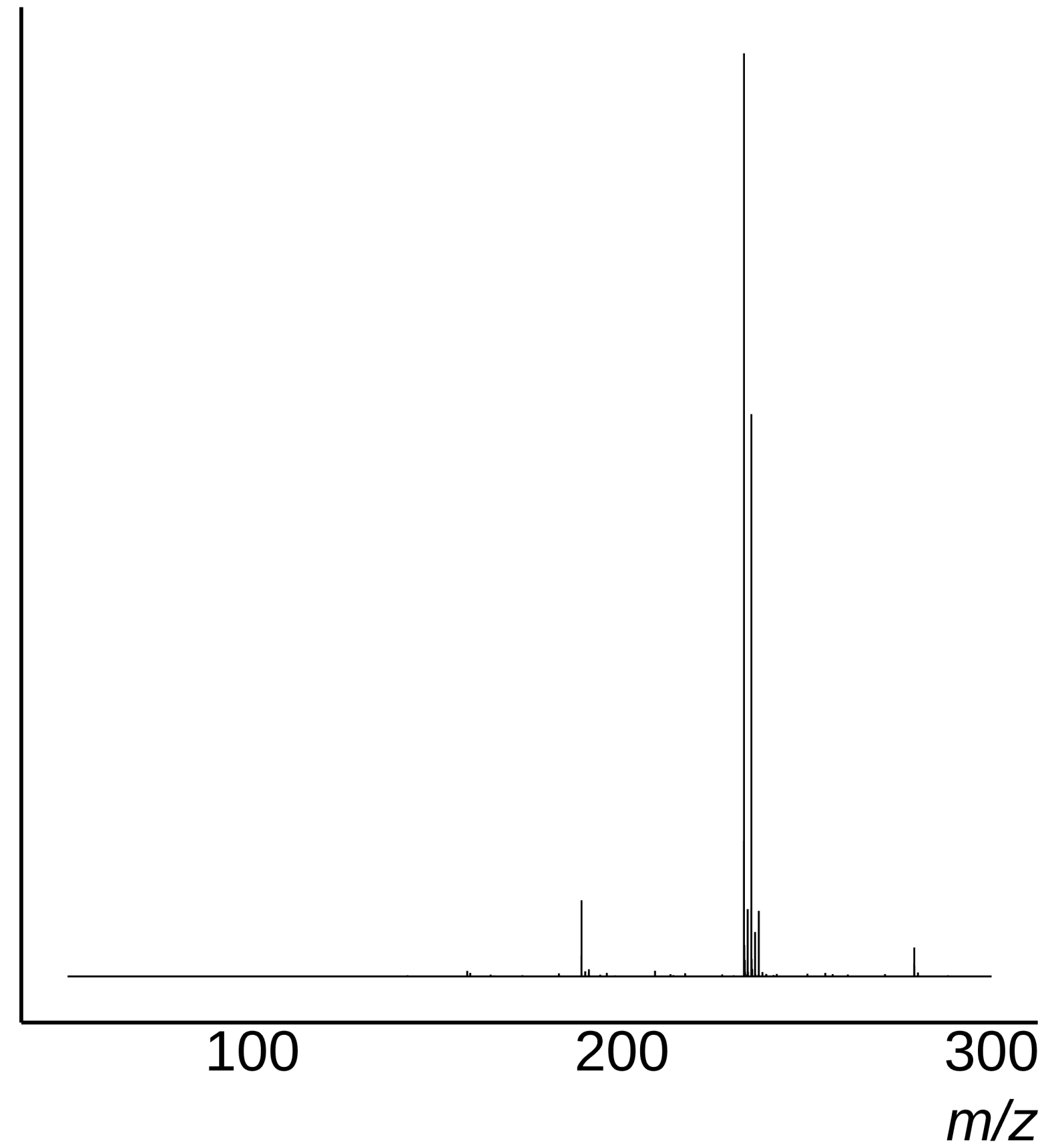
is chemical risk introduced?

analysis

- LC/HRMS
- targeted screening
- nontargeted screening

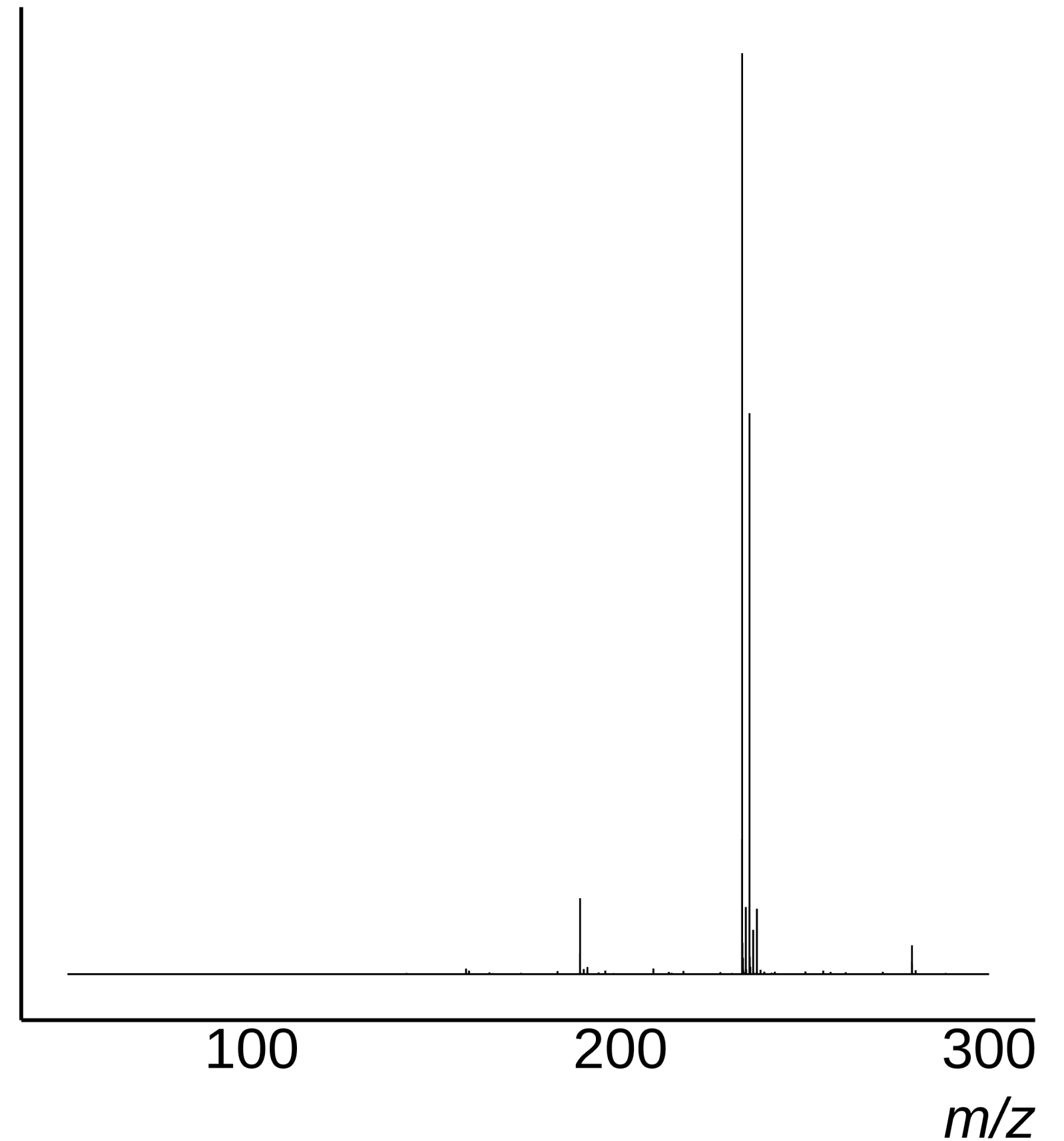


mass spectra



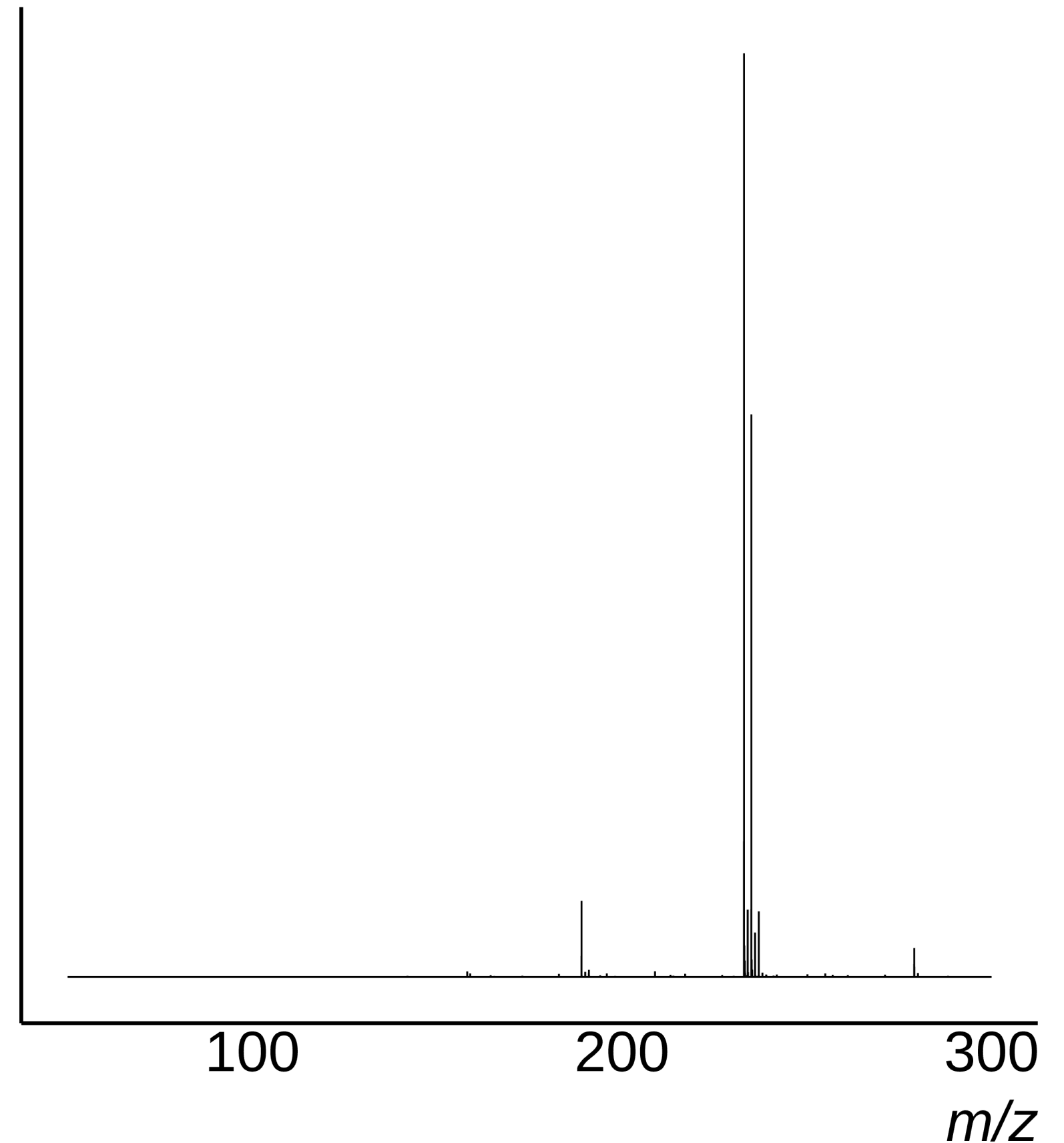
mass spectra

- library matching
MassBank, MoNA, GNPS, NIST, ...



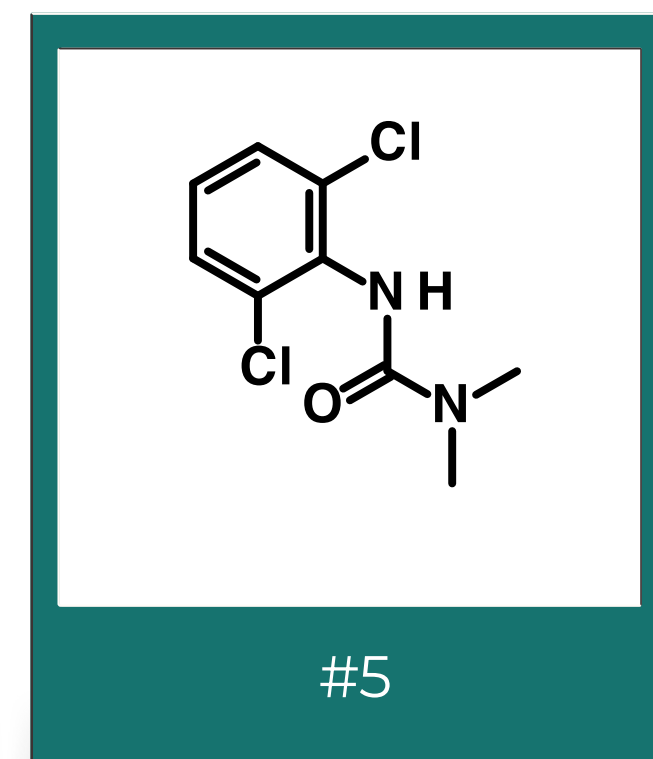
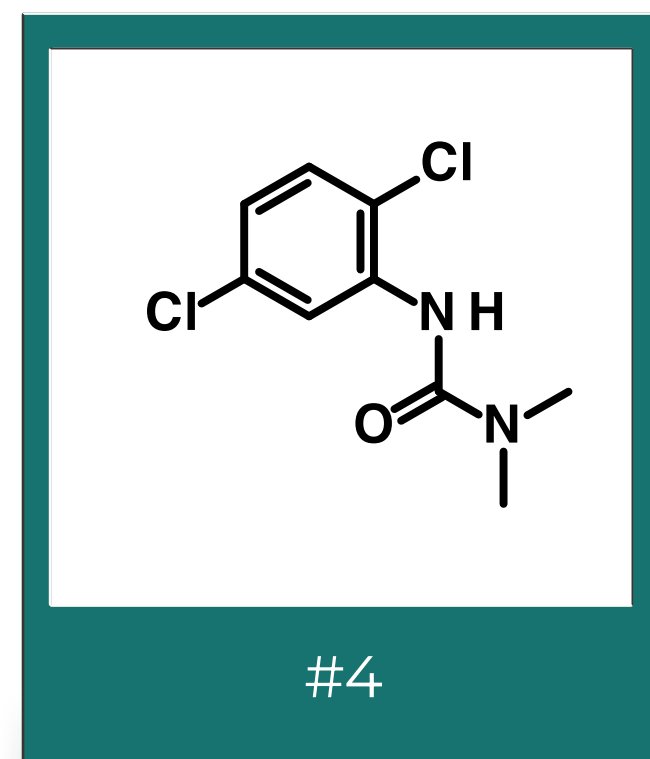
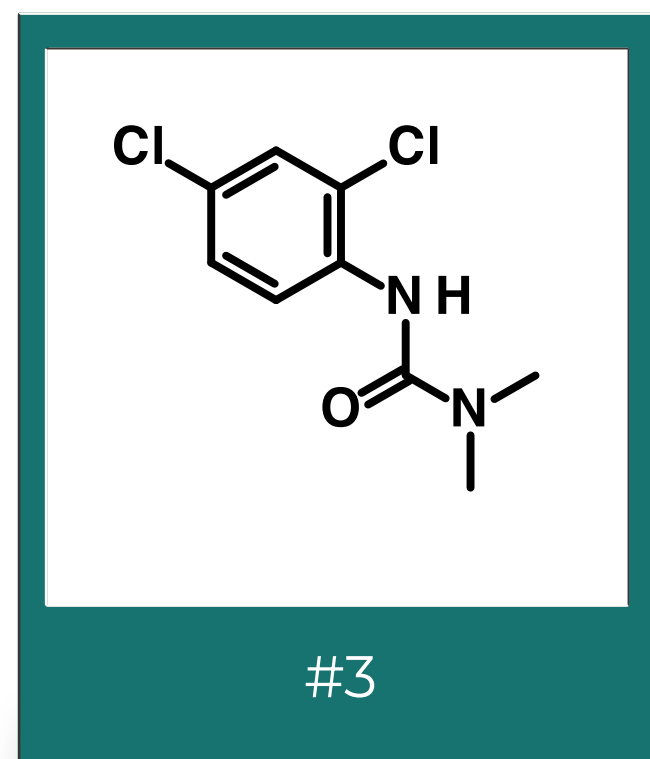
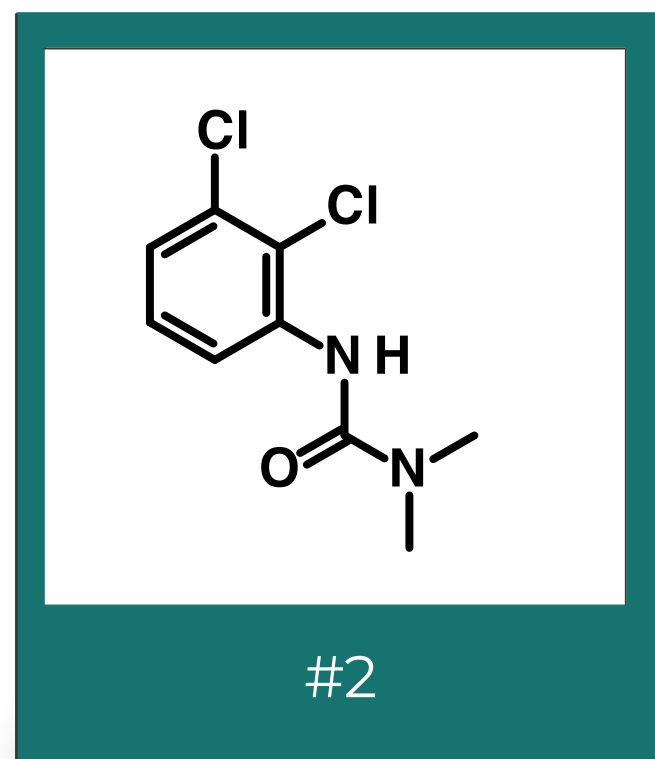
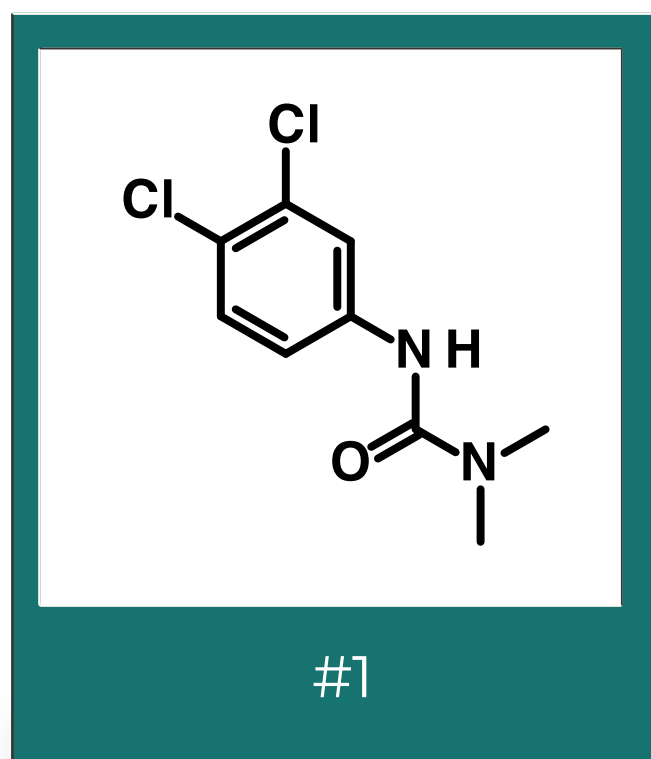
mass spectra

- library matching
MassBank, MoNA, GNPS, NIST, ...
- *in silico* interpretation
SIRIUS, MetFrag, CFM-ID, ...



candidate structures

structural/positional isomers

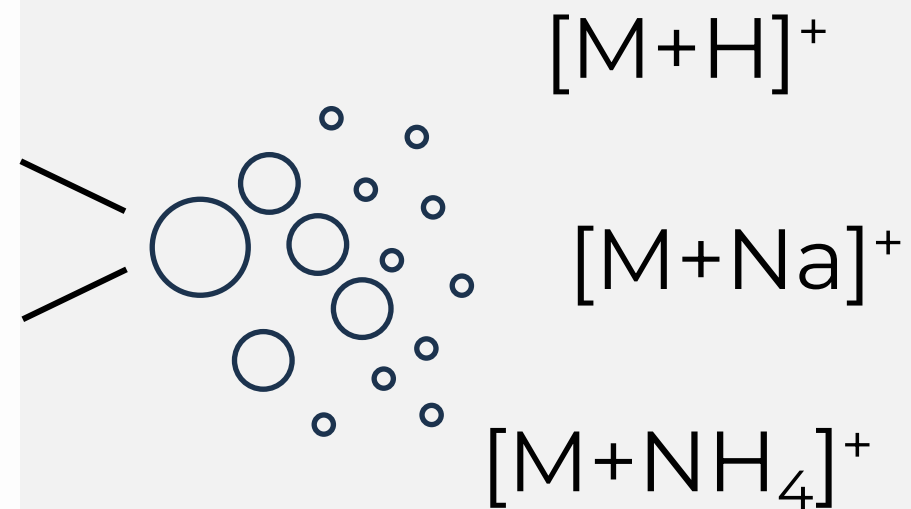


analytical information

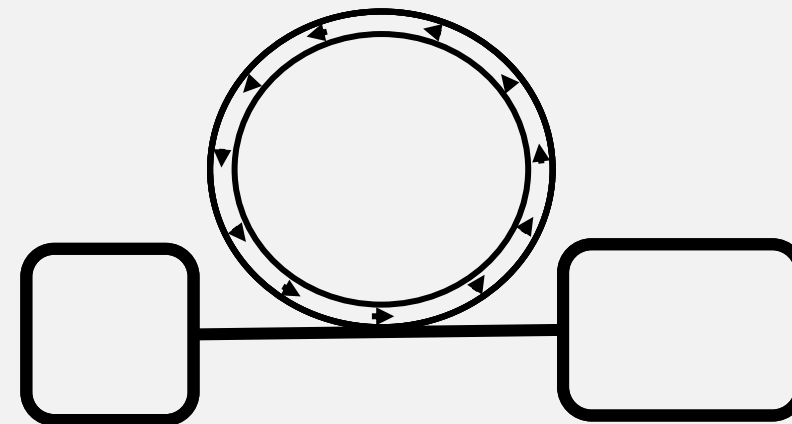
chromatography



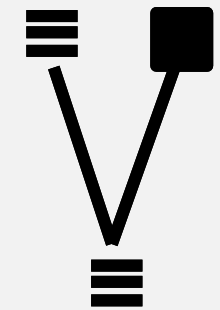
ionization



ion mobility



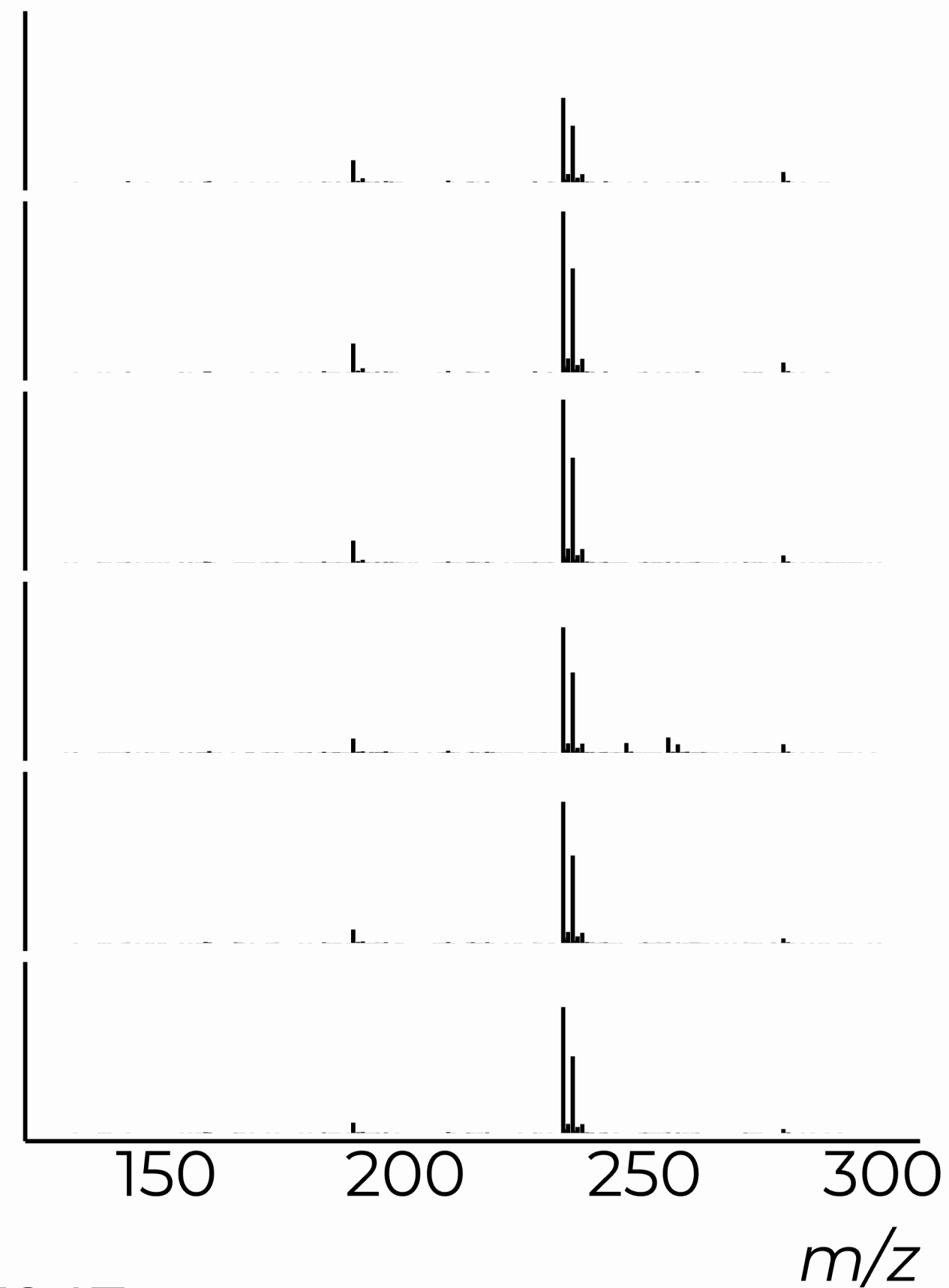
fragmentation



Akhlaqi et al. Anal. Bioanal. Chem. 415 (2023) 5247
Hupatz, Rahu, et al. Anal. Bioanal. Chem. 417 (2025) 473

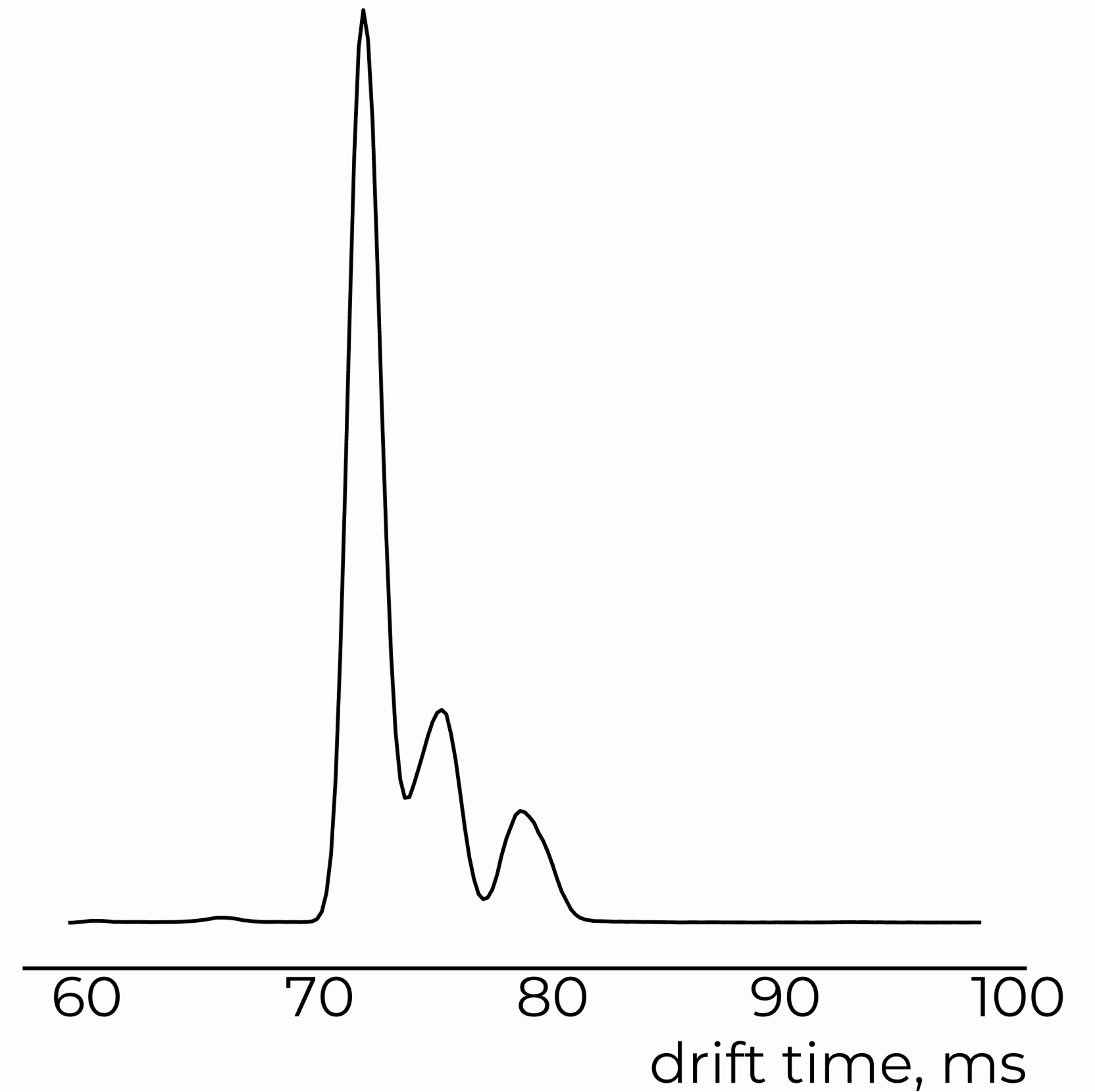
confirmation

- MS²
indistinguishable



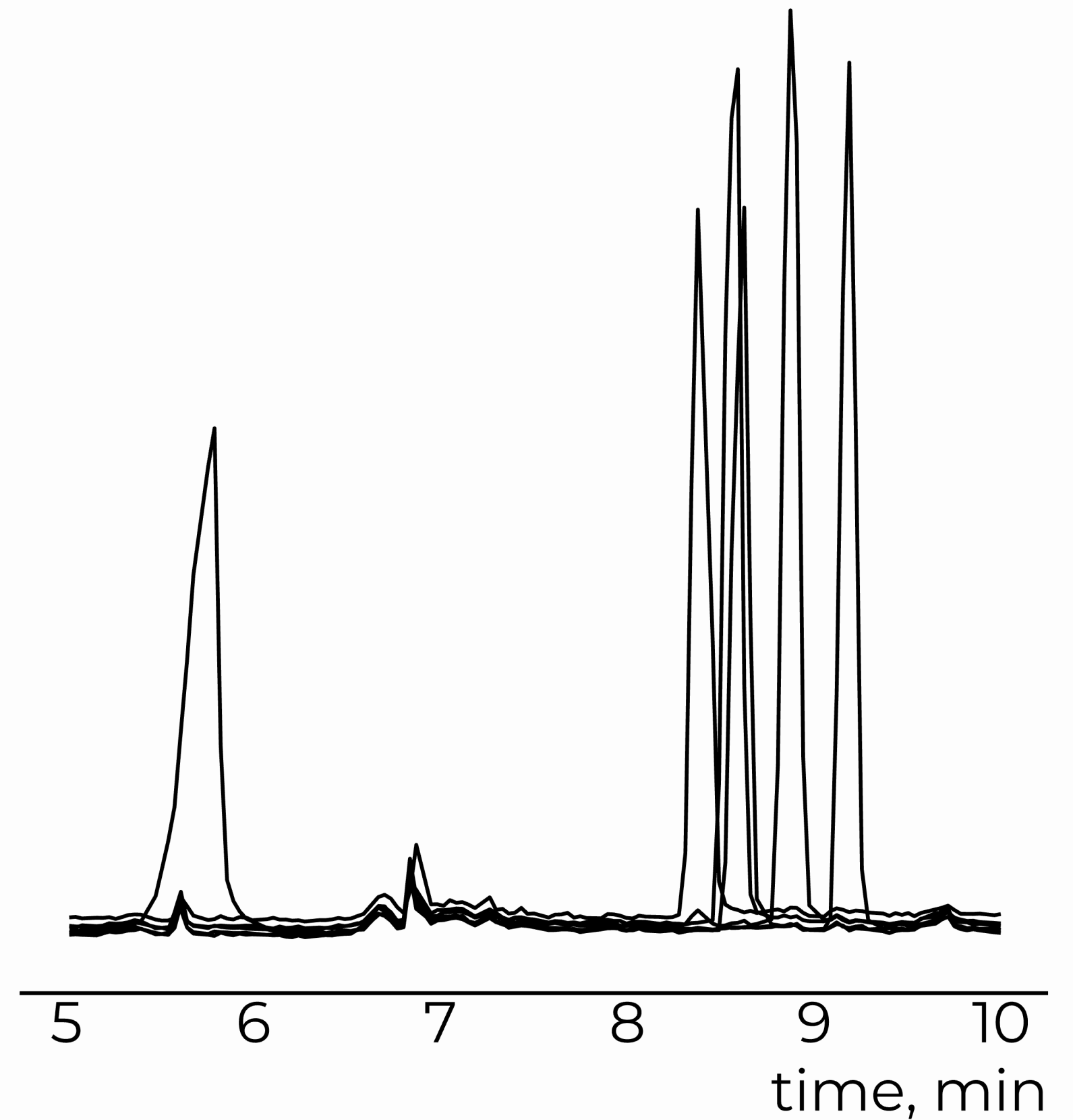
confirmation

- MS²
indistinguishable
- IMS
indistinguishable



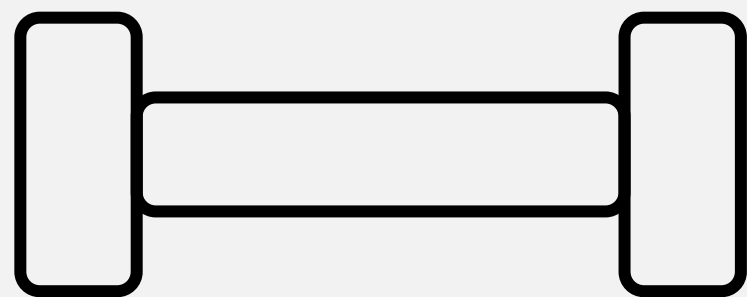
confirmation

- MS²
indistinguishable
- IMS
indistinguishable
- LC
50% distinguishable

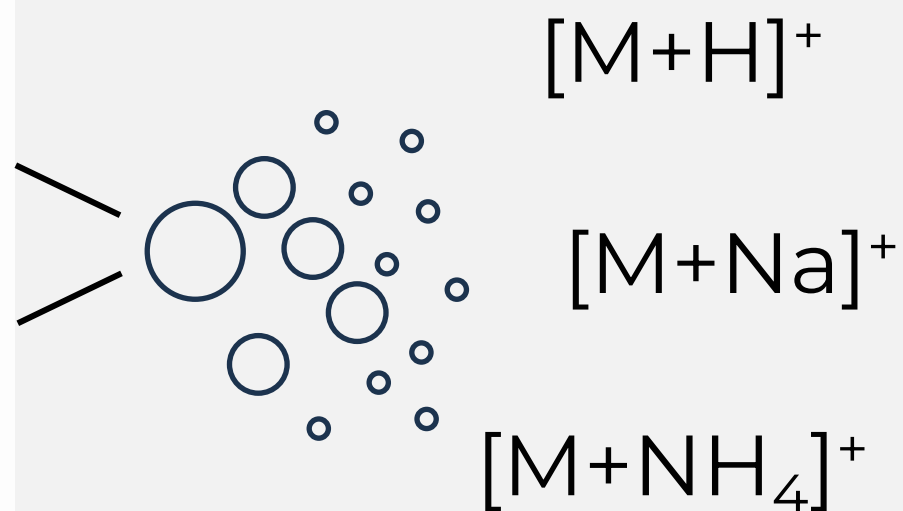


analytical information

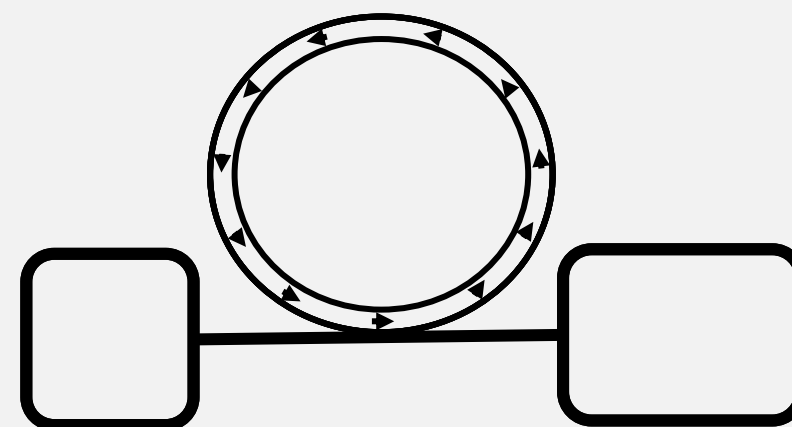
chromatography



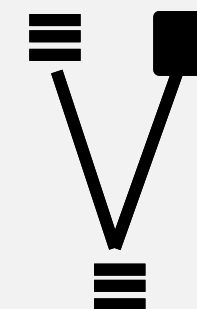
ionization



ion mobility



fragmentation



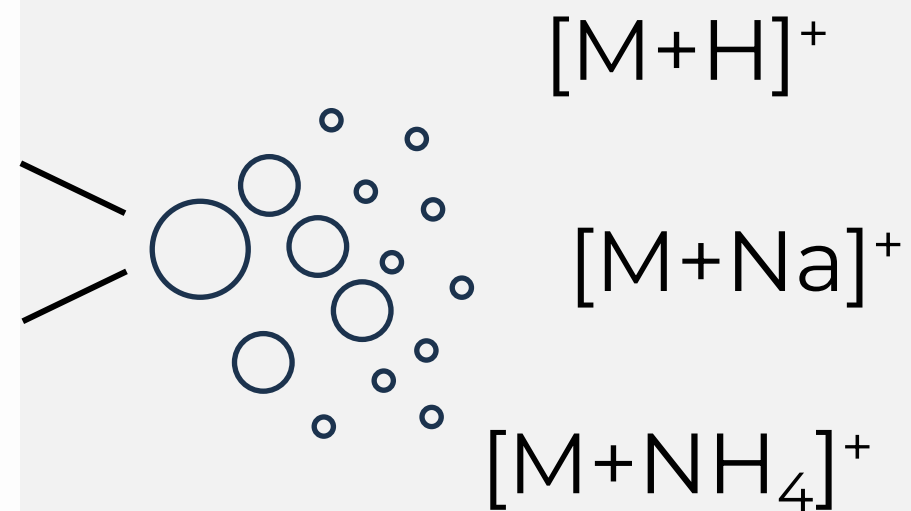
analytical information

chromatography



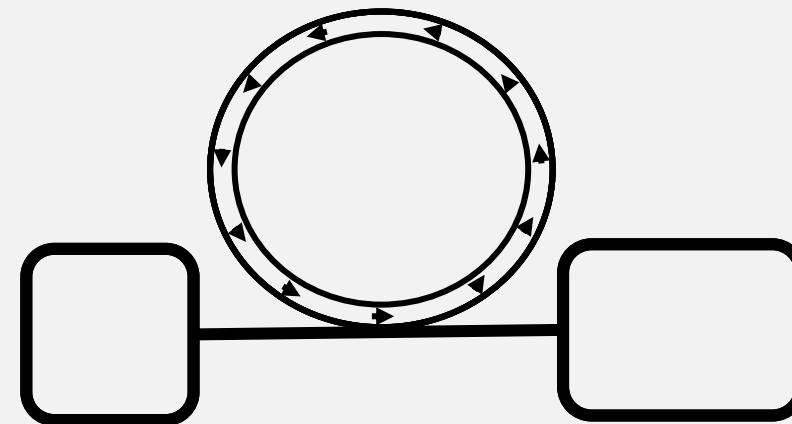
~3 of 5 cases

ionization



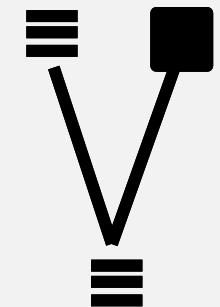
0 of 5 cases

ion mobility



~1 of 5 cases

fragmentation



~2 of 5 cases



#1a

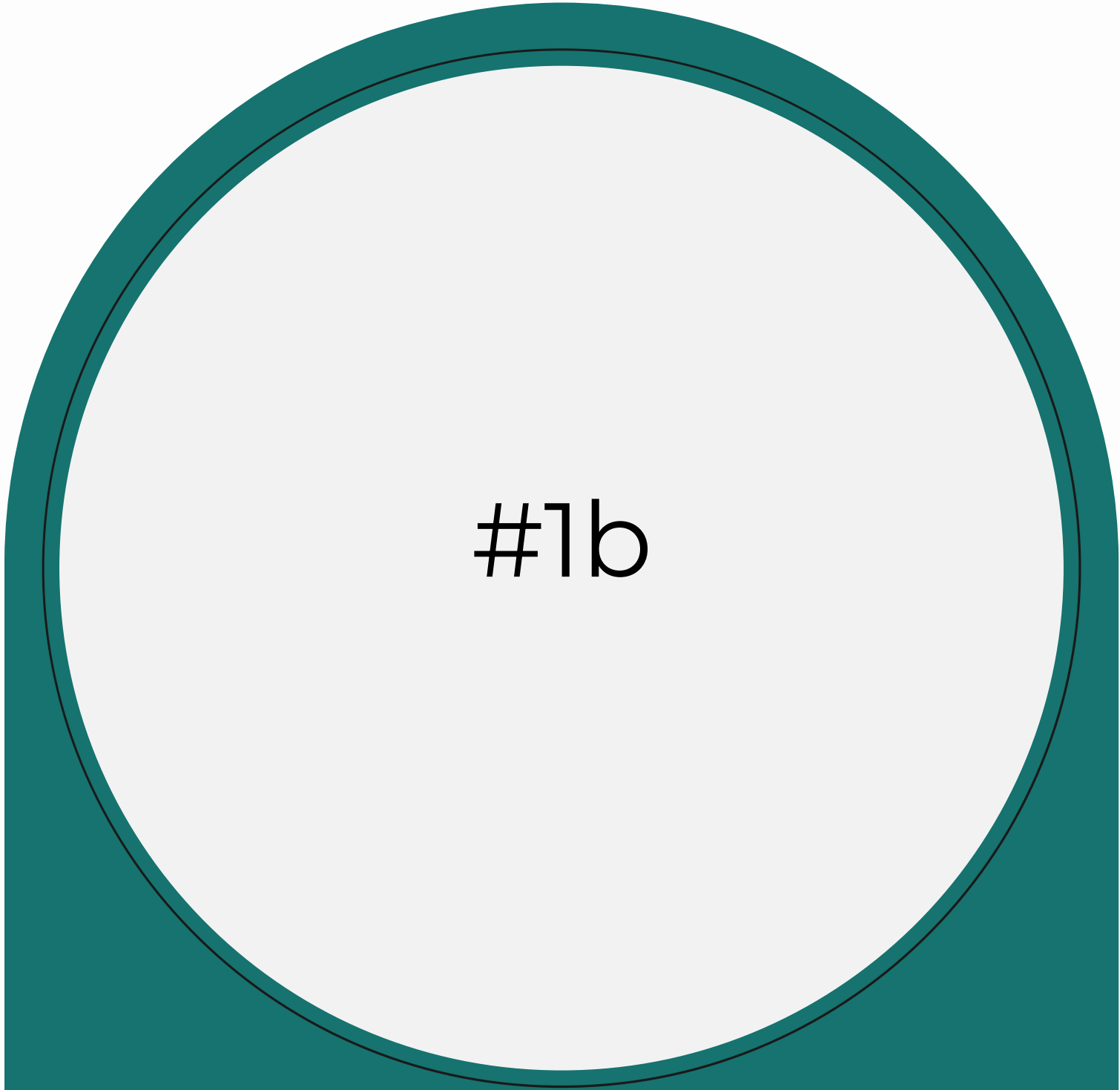
take-home
message

wrong candidate structure

≠

false positive

experimentally inseparable
isomeric candidate
structures are common



#1b

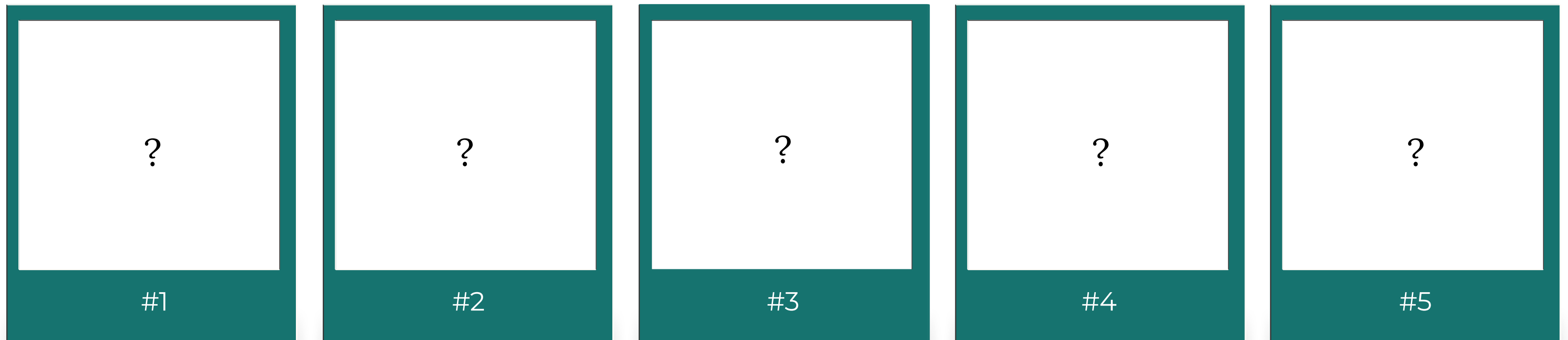
take-home
message

selectivity of analytics
matters

in silico tools cannot
distinguish chemicals that
are not separable
analytically

candidate structures

no (likely) candidate structures

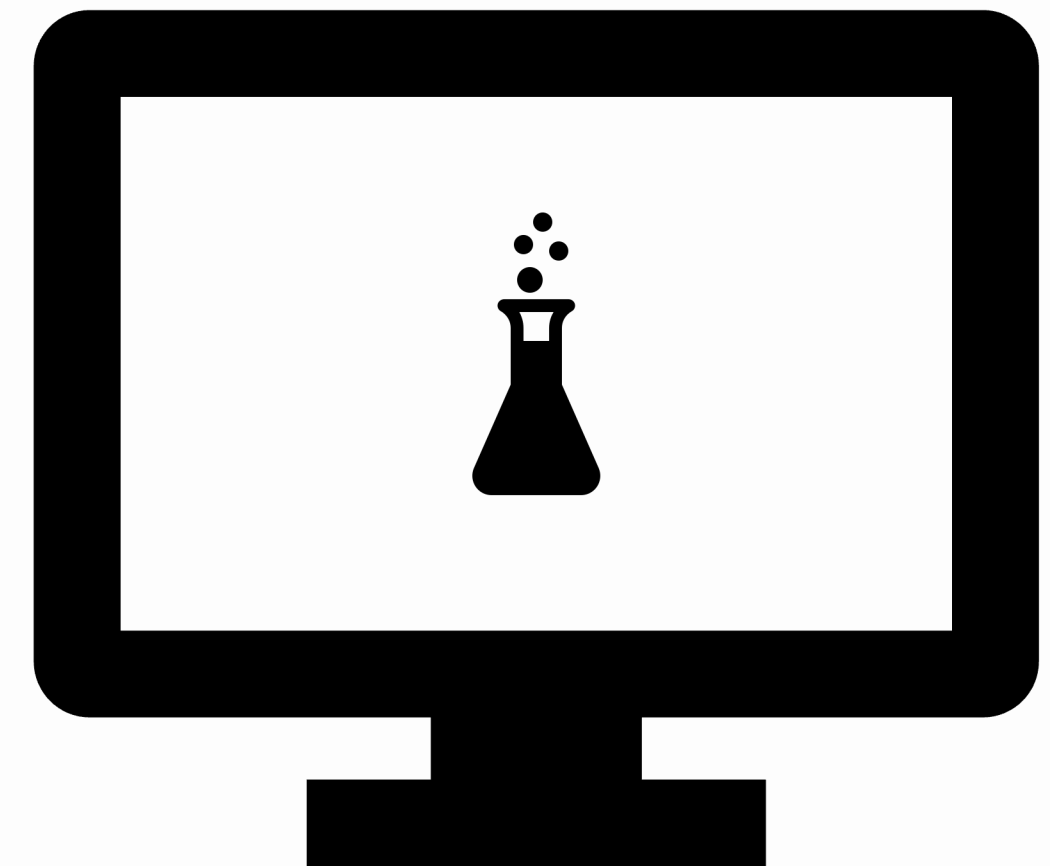


generate structures

- expert knowledge
time consuming

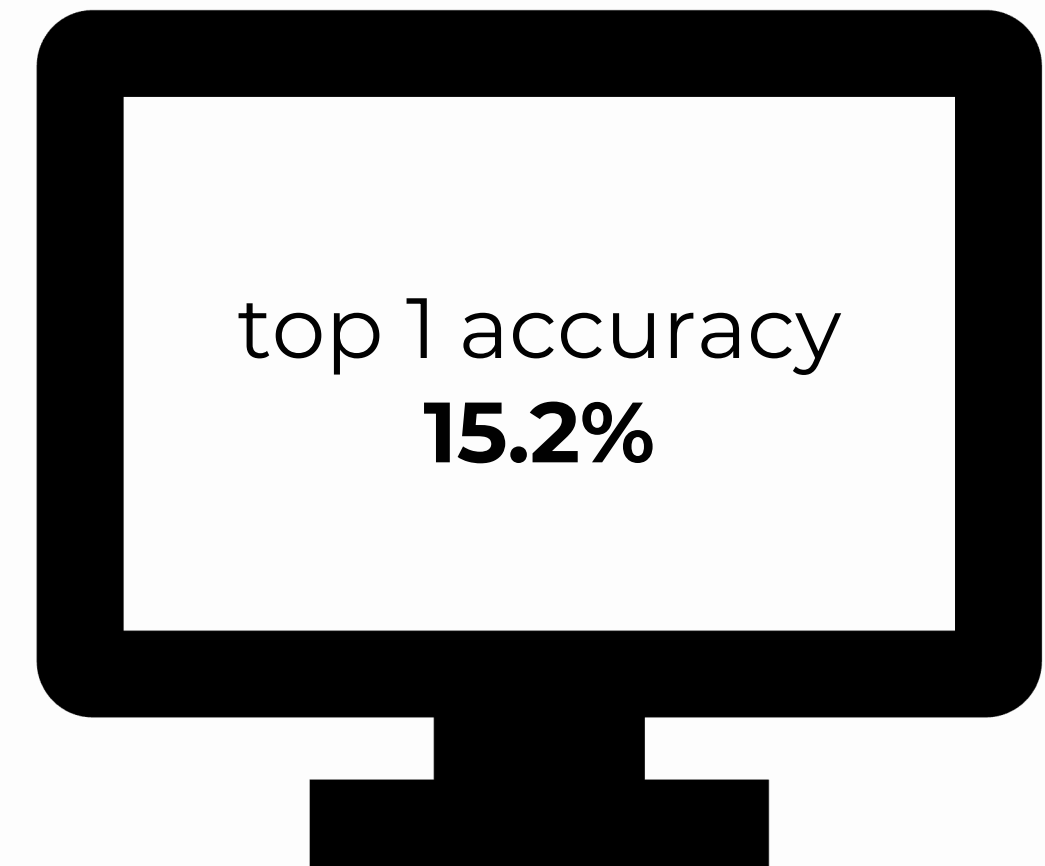
generate structures

- expert knowledge
time consuming
- in silico
MSNovelist



generate structures

- expert knowledge
time consuming
- in silico
MSNovelist





#2

take-home
message

evaluating unknowns
is hard

what is the ground truth?

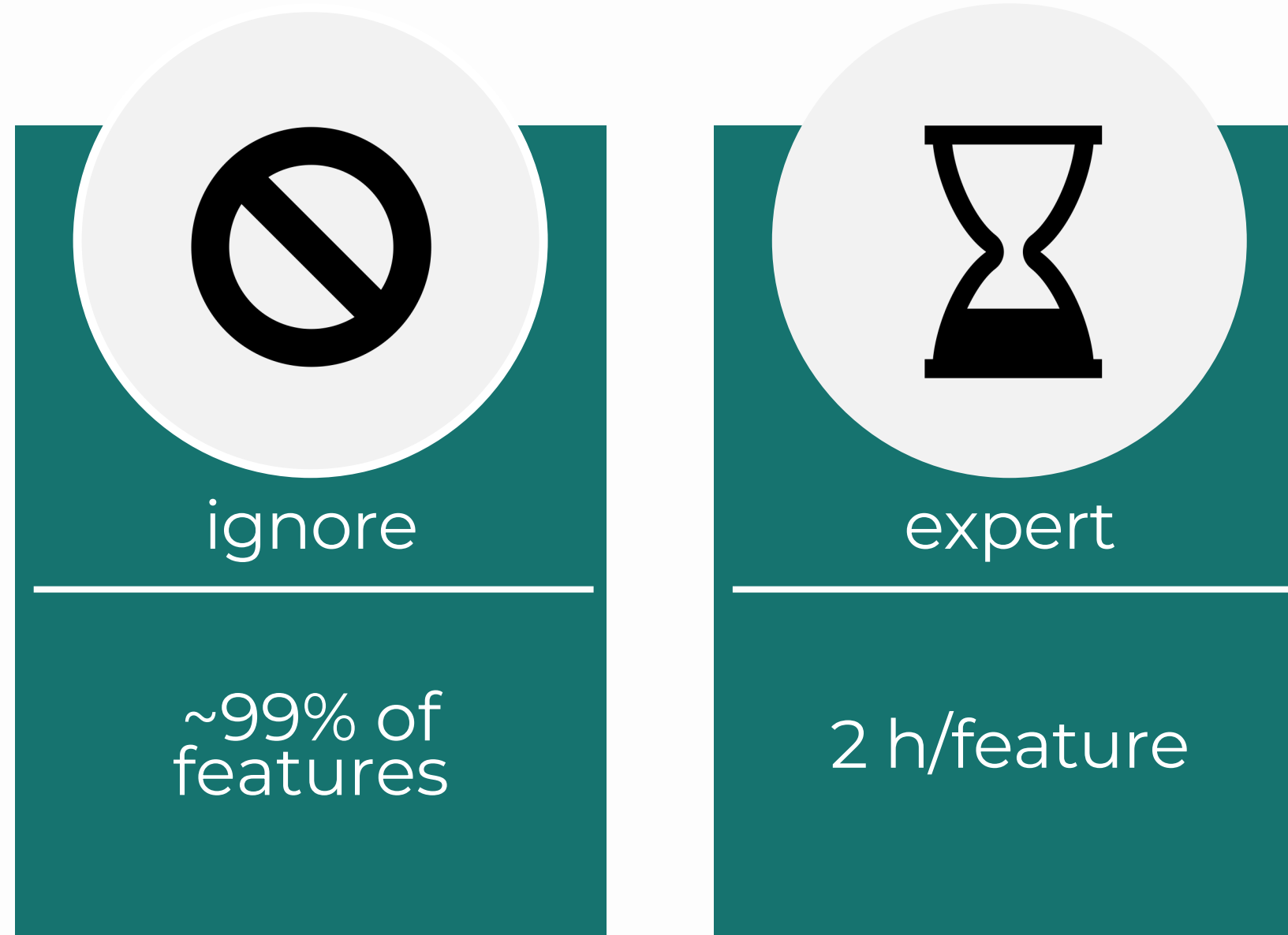
unidentified LC/HRMS features

how to handle them?



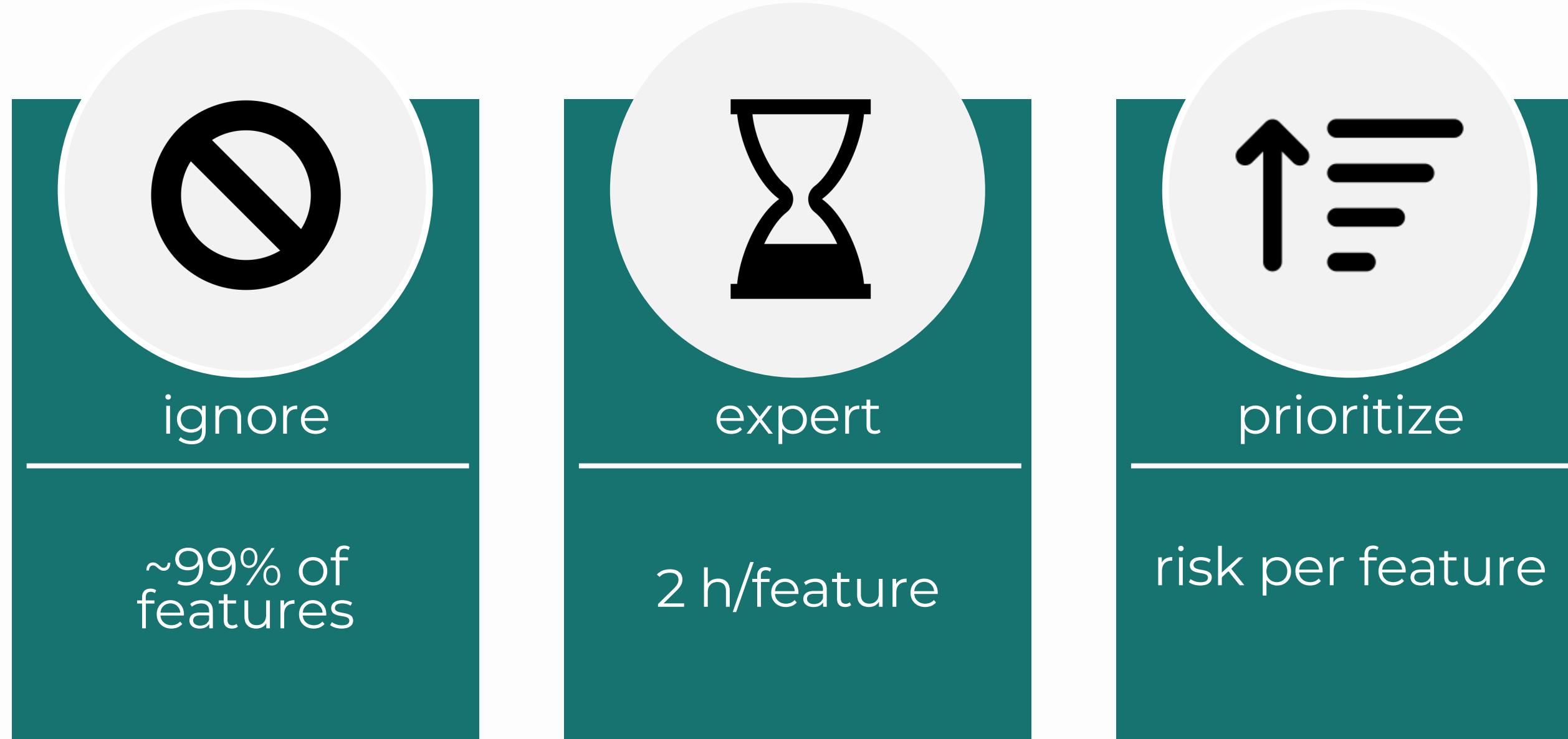
unidentified LC/HRMS features

how to handle them?



unidentified LC/HRMS features

how to handle them?



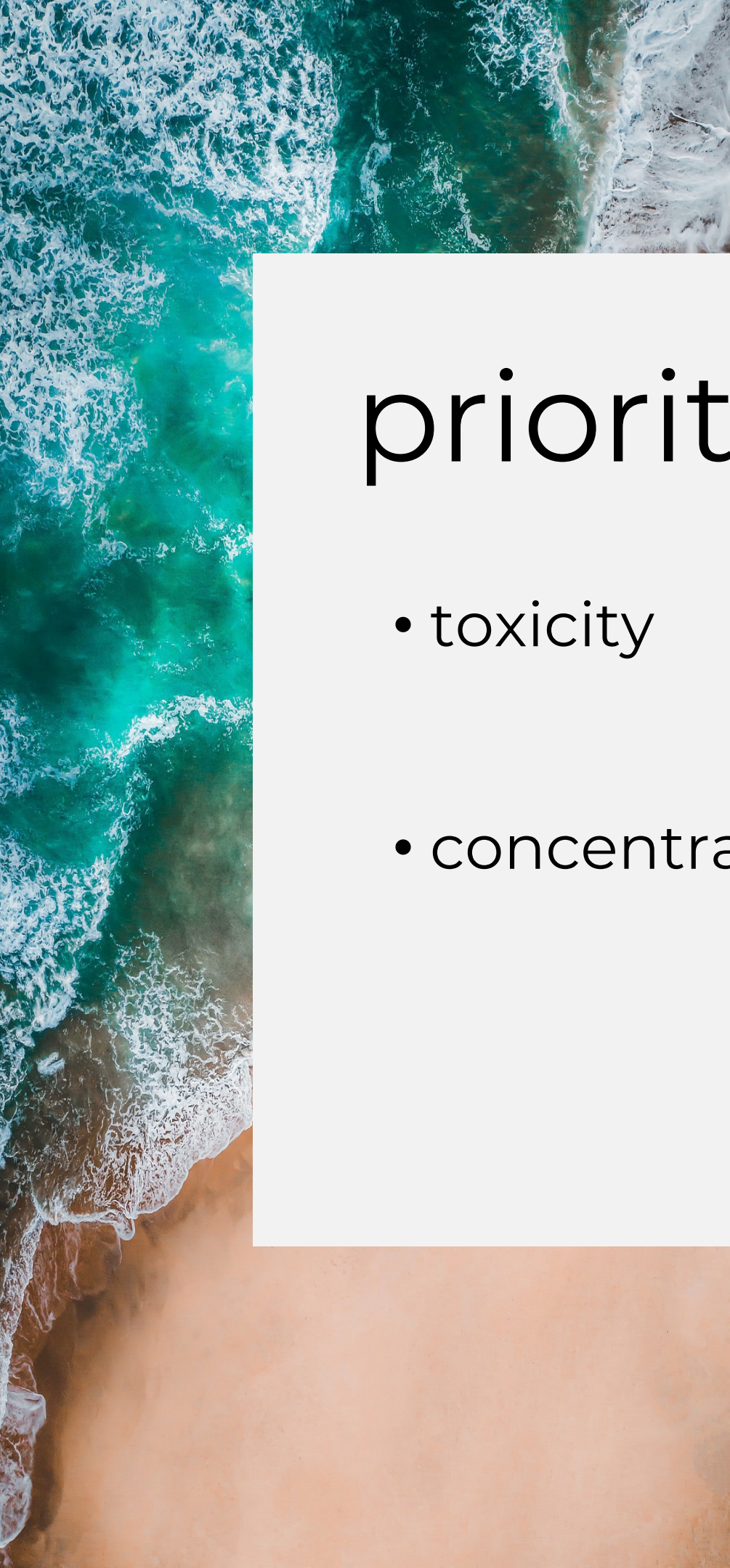
prioritization



prioritization

- toxicity

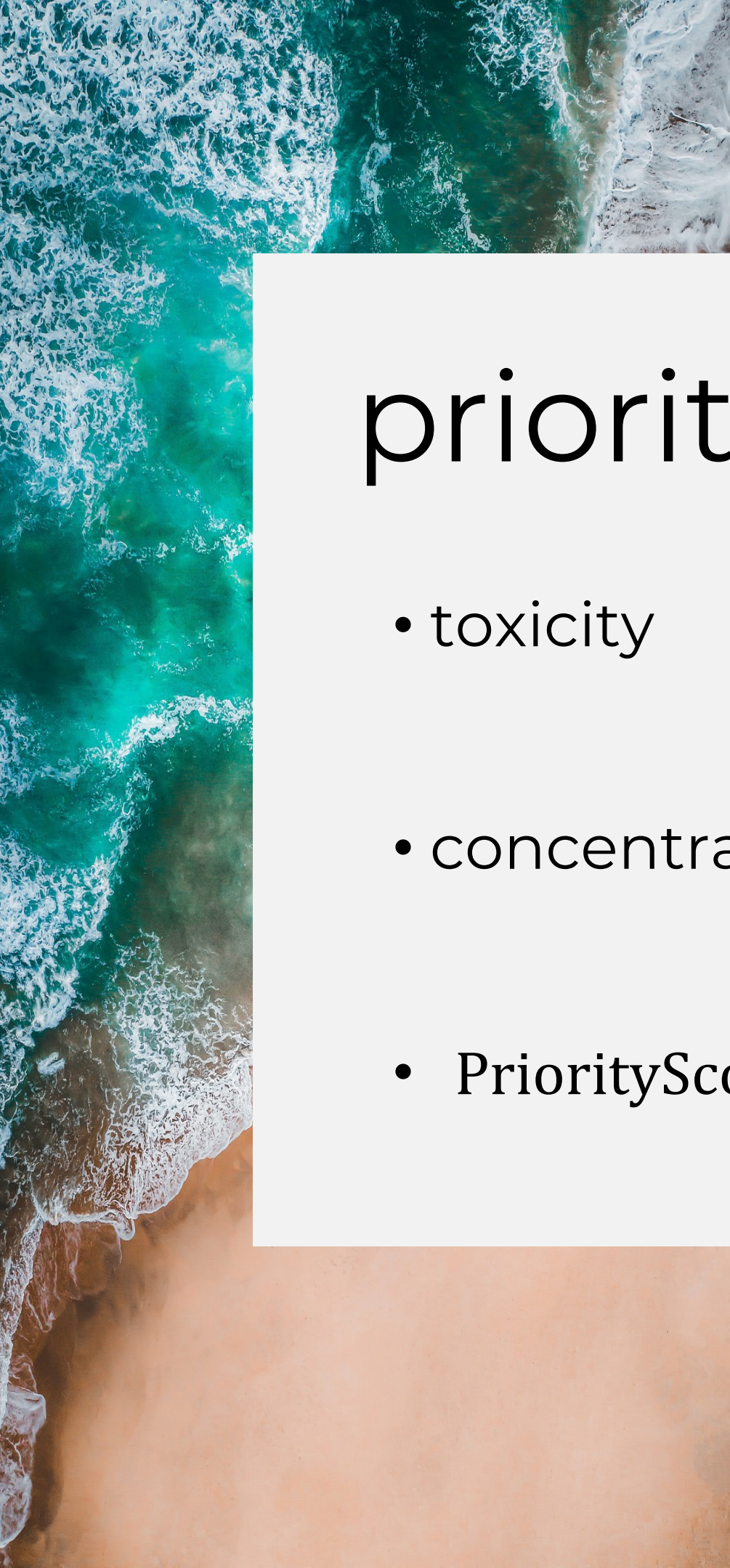




prioritization

- toxicity
- concentration





prioritization

- toxicity

- concentration

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





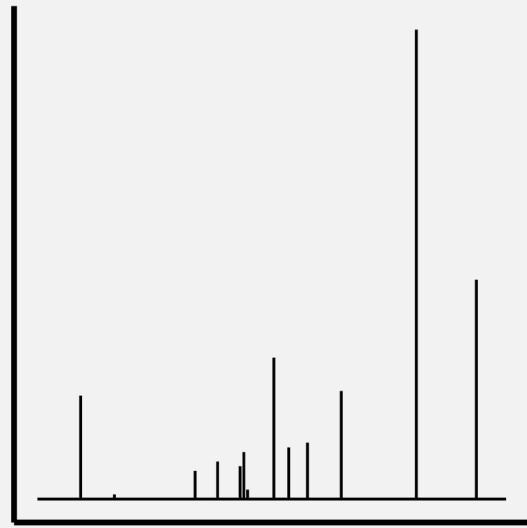
toxicity

- ecotoxicity
- endocrine disruptors



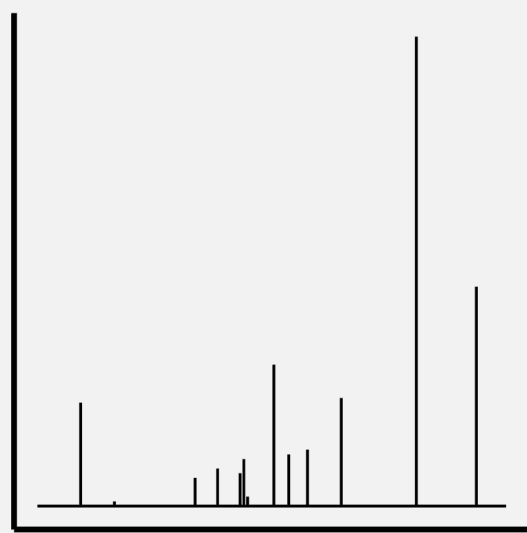
toxicity

MS² spectra

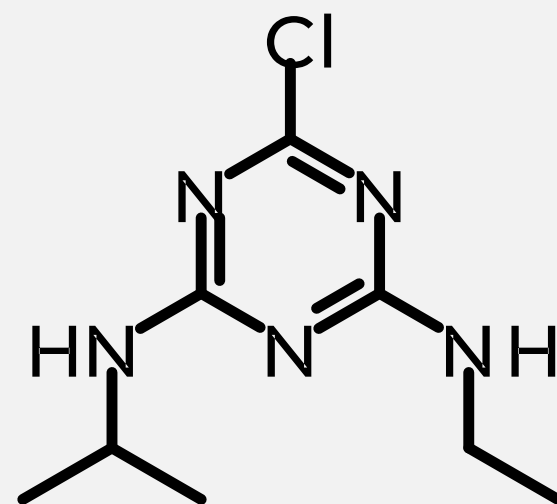


toxicity

MS² spectra

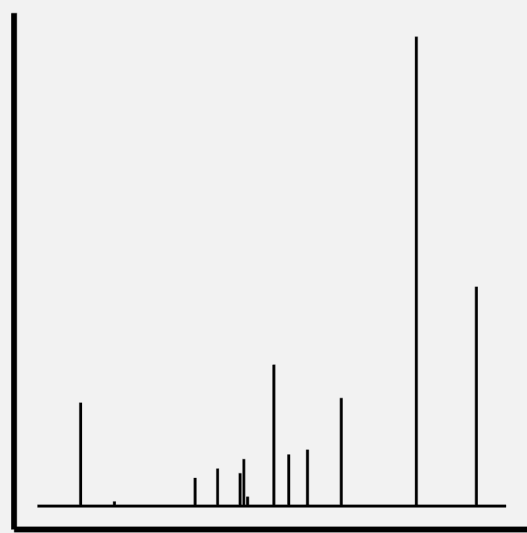


structure

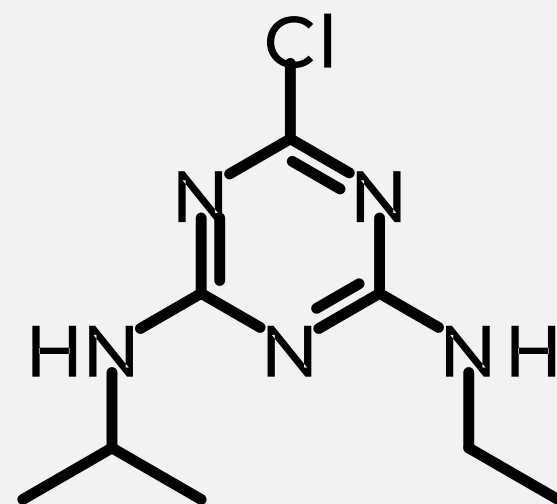


toxicity

MS² spectra



structure

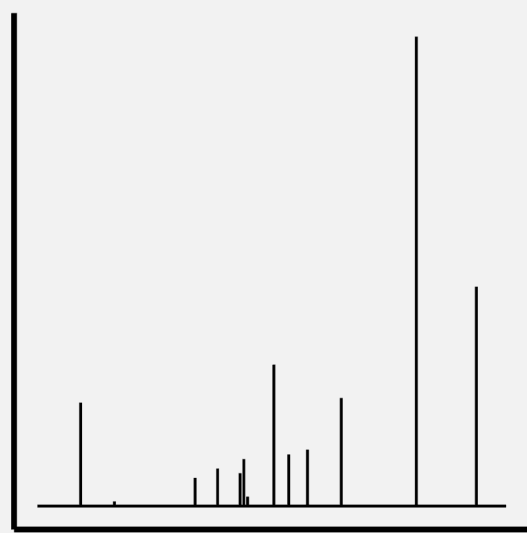


descriptors

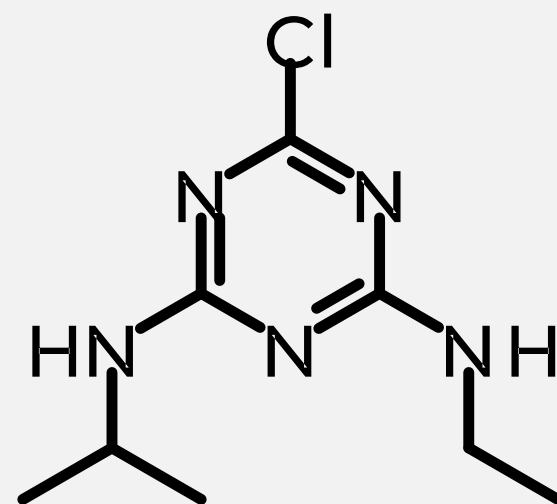


toxicity

MS² spectra



structure



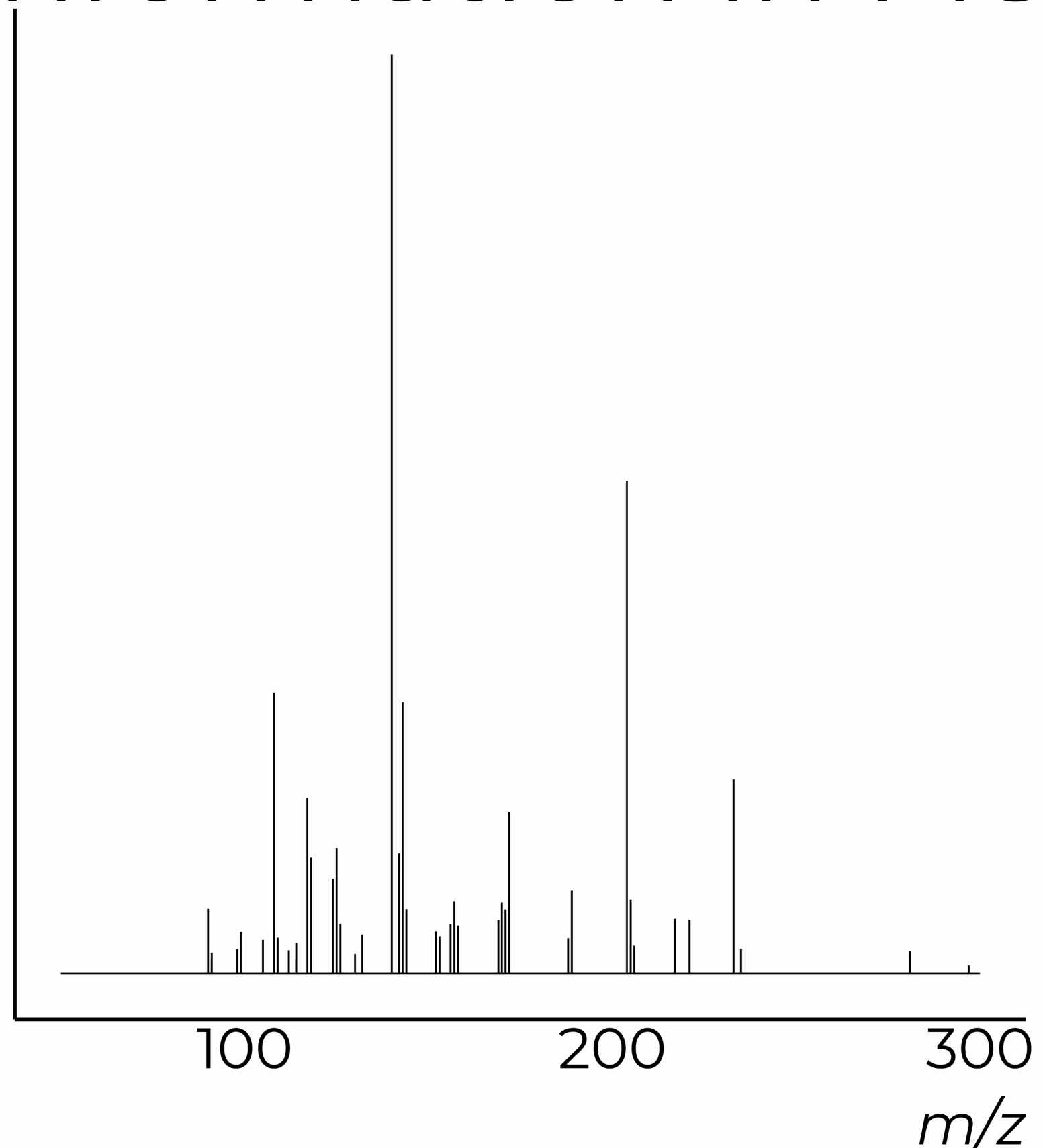
descriptors



predictions

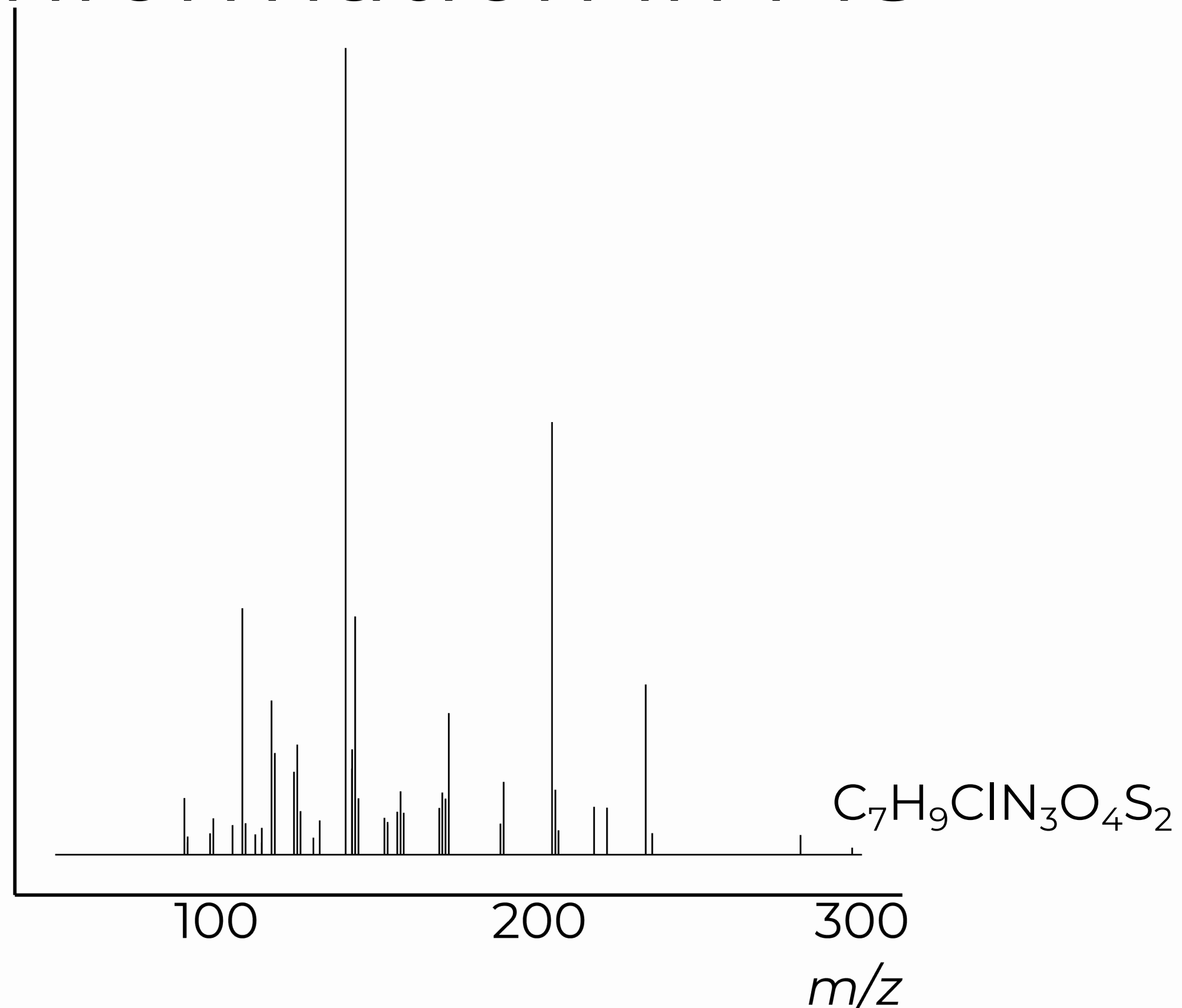


information in MS²



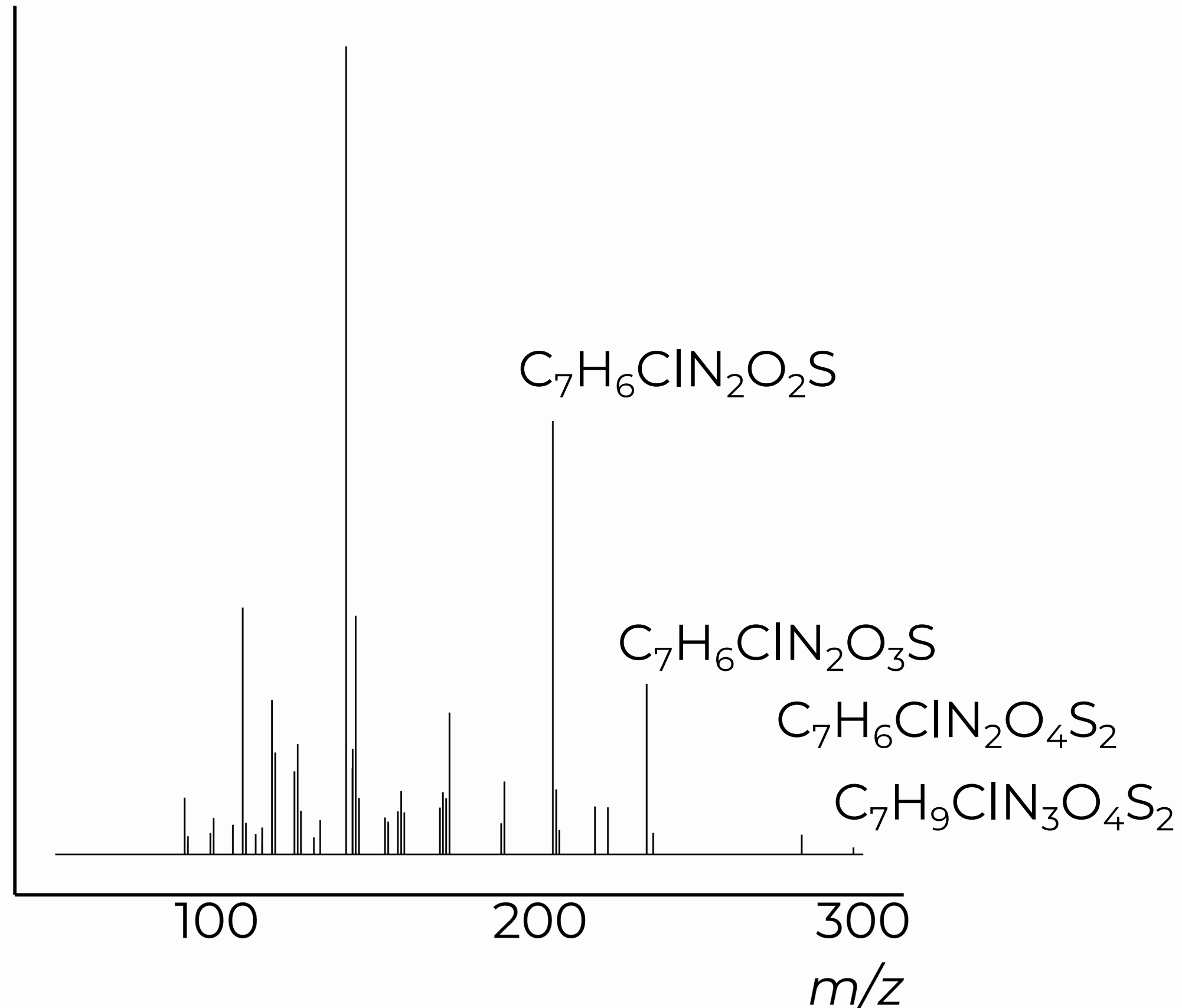
Dührkop et al. Nature Methods 16 (2019) 299.

information in MS²

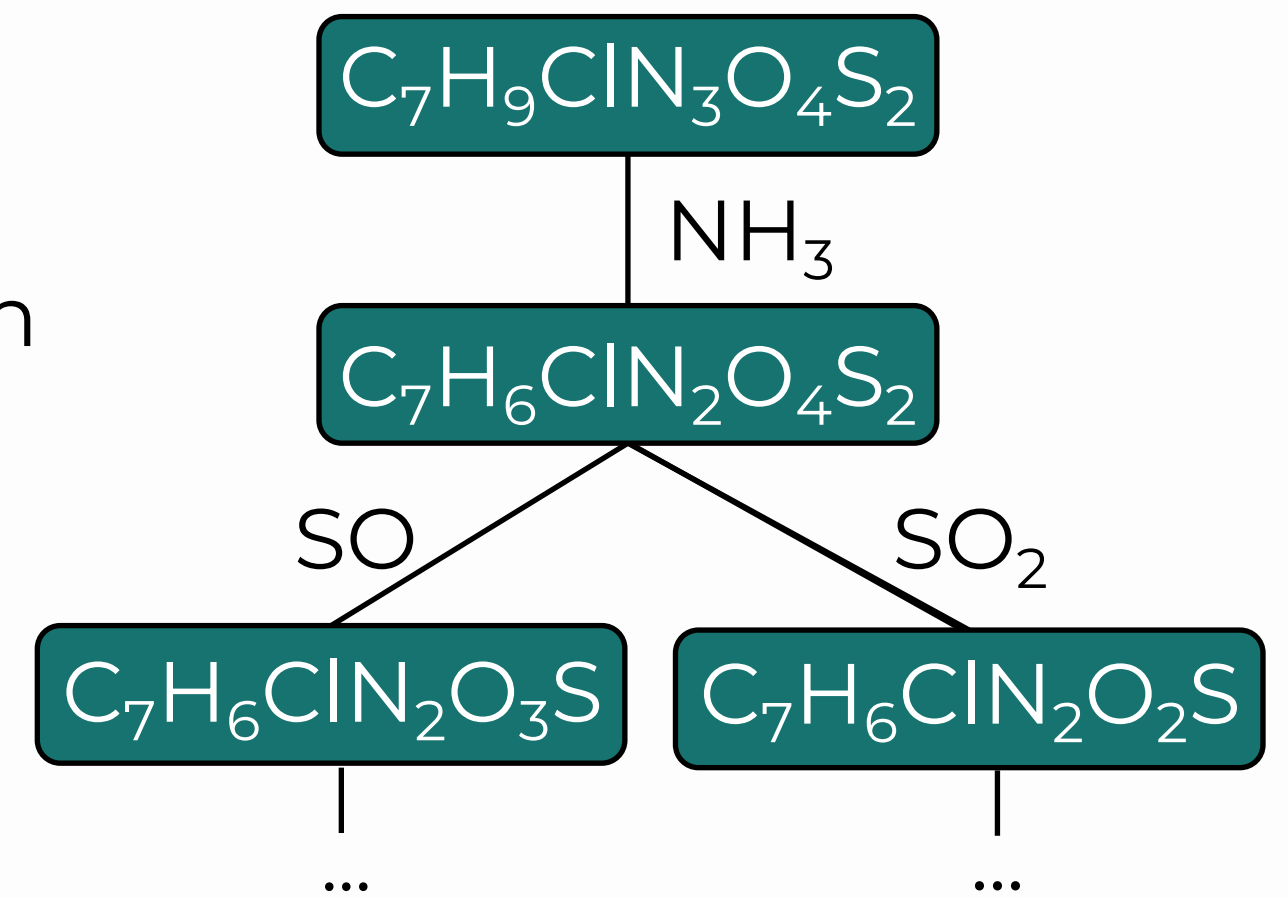
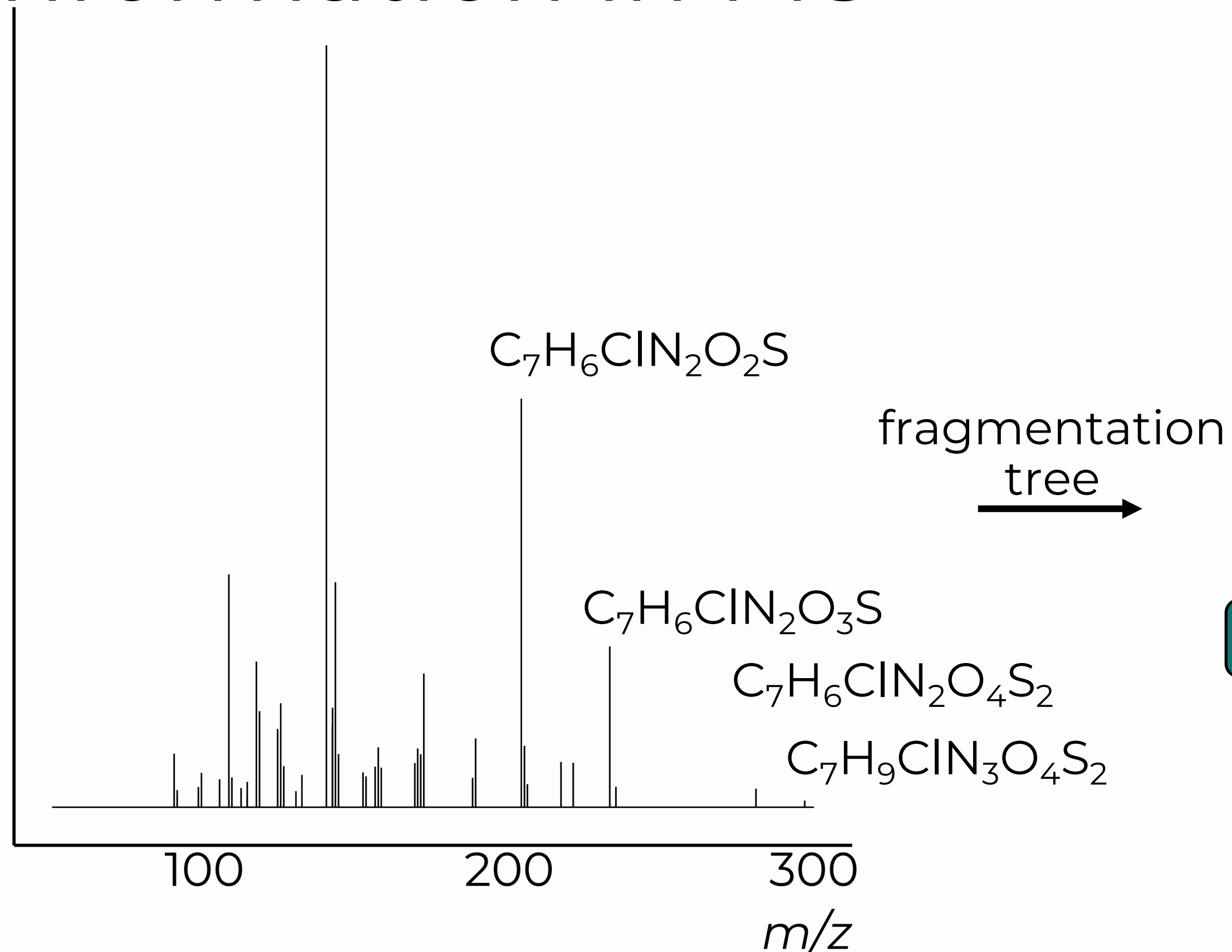


Dührkop et al. Nature Methods 16 (2019) 299.

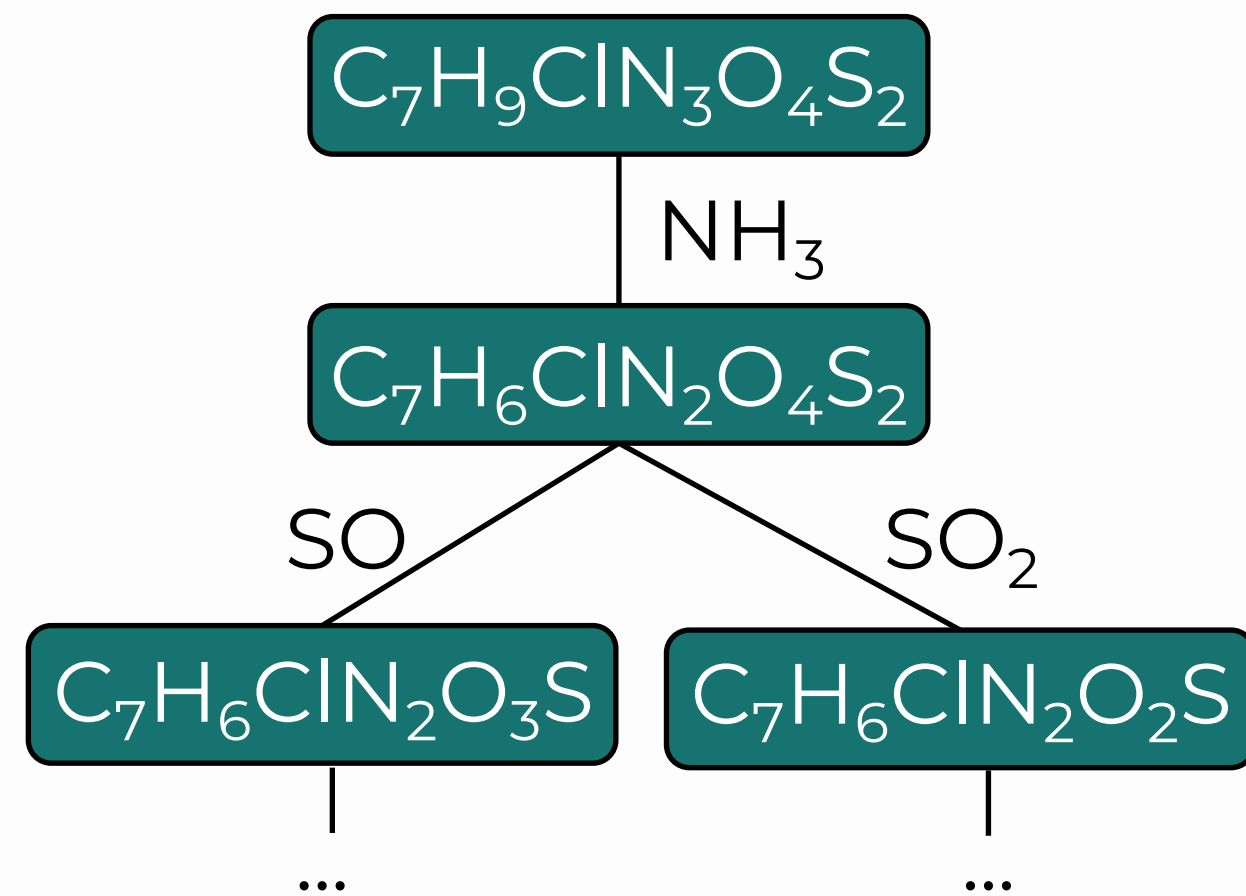
information in MS²



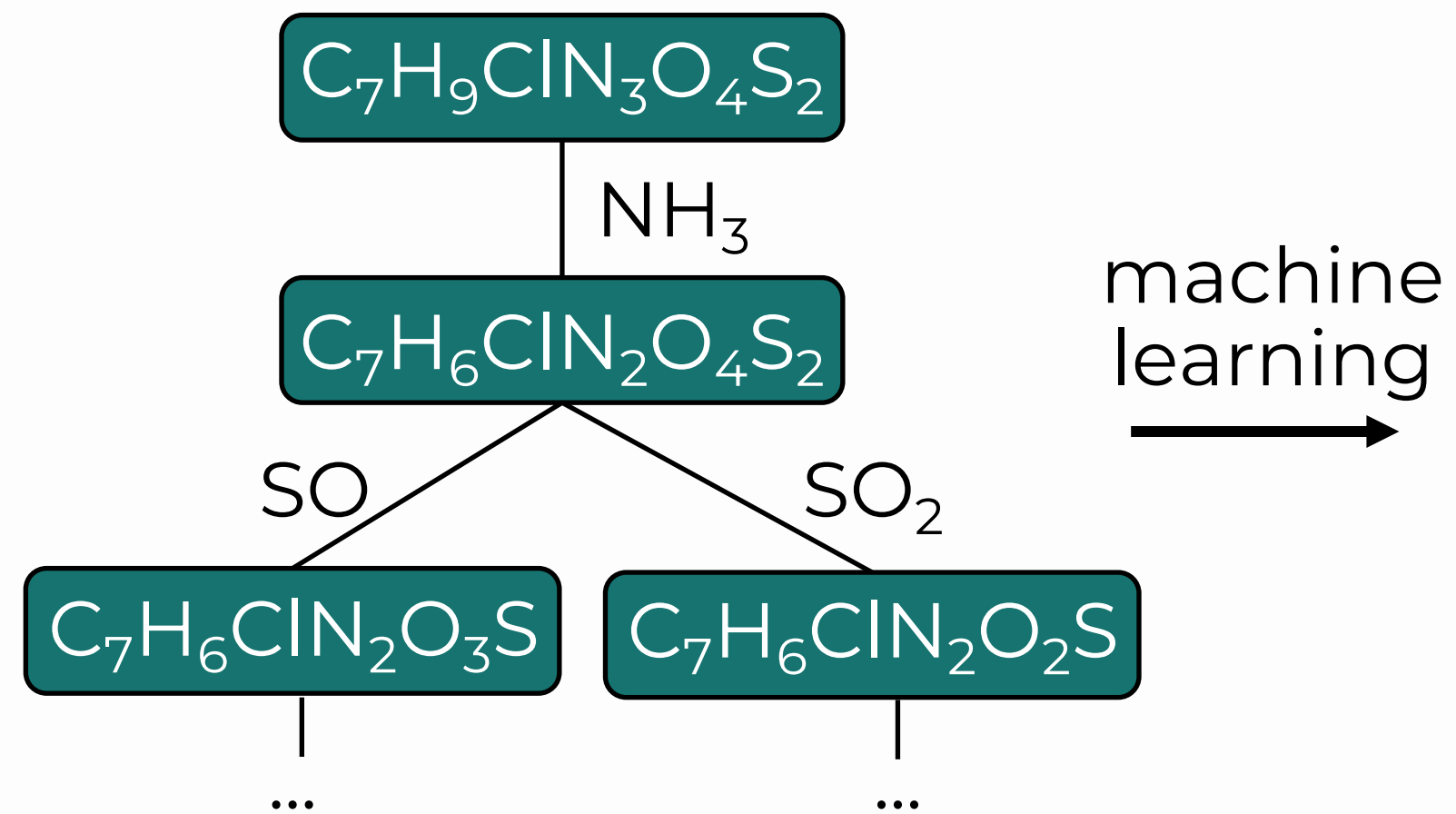
information in MS²



information in MS²



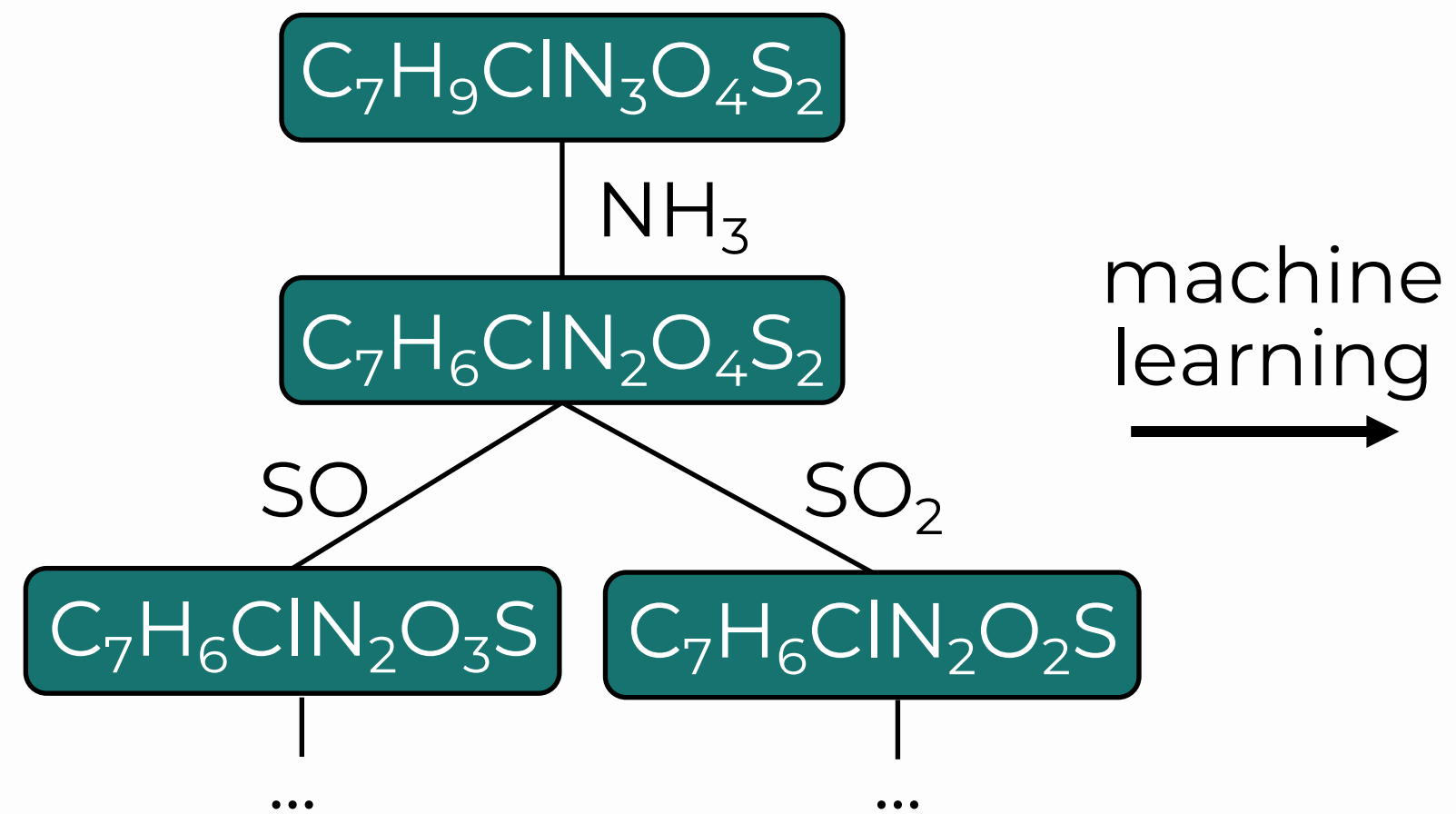
information in MS²



machine
learning
→

0.08	O-P
0.87	S=O
0.15	C-F
0.93	NH ₂
0.99	O

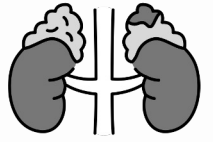
information in MS²



0.08	O-P
0.87	S=O
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0.99	O

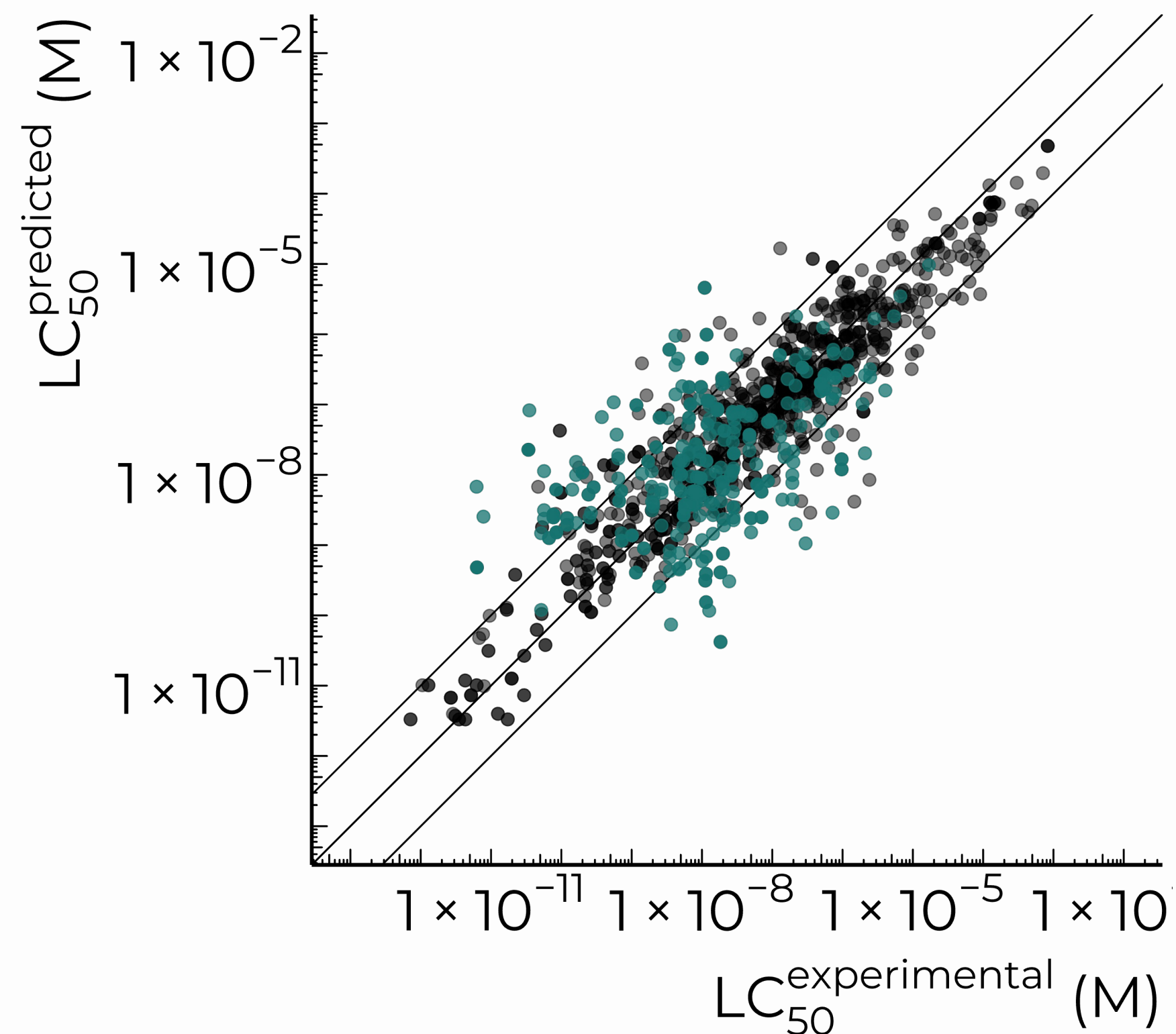
machine learning →


LC₅₀ fish


endocrine
disruption
potential

LC₅₀ predictions

- from structure
RMSE 0.78 log(M)
- from MS² data
RMSE 0.88 log(M)
- experimental
SD 0.44 log(M)



endocrine disruption potential

- Tox21 Data Challenge
- 12 endpoints
- active vs inactive

endocrine disruption potential

- Tox21 Data Challenge

- 12 endpoints

- active vs inactive

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

endocrine disruption potential

- Tox21 Data Challenge

- 12 endpoints

- active vs inactive

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

endocrine disruption potential

- Tox21 Data Challenge

- 12 endpoints

- active vs inactive

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

FPR @ TPR = 0.9

endocrine disruption potential

- Tox21 Data Challenge
- 12 endpoints
- active vs inactive

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



#3

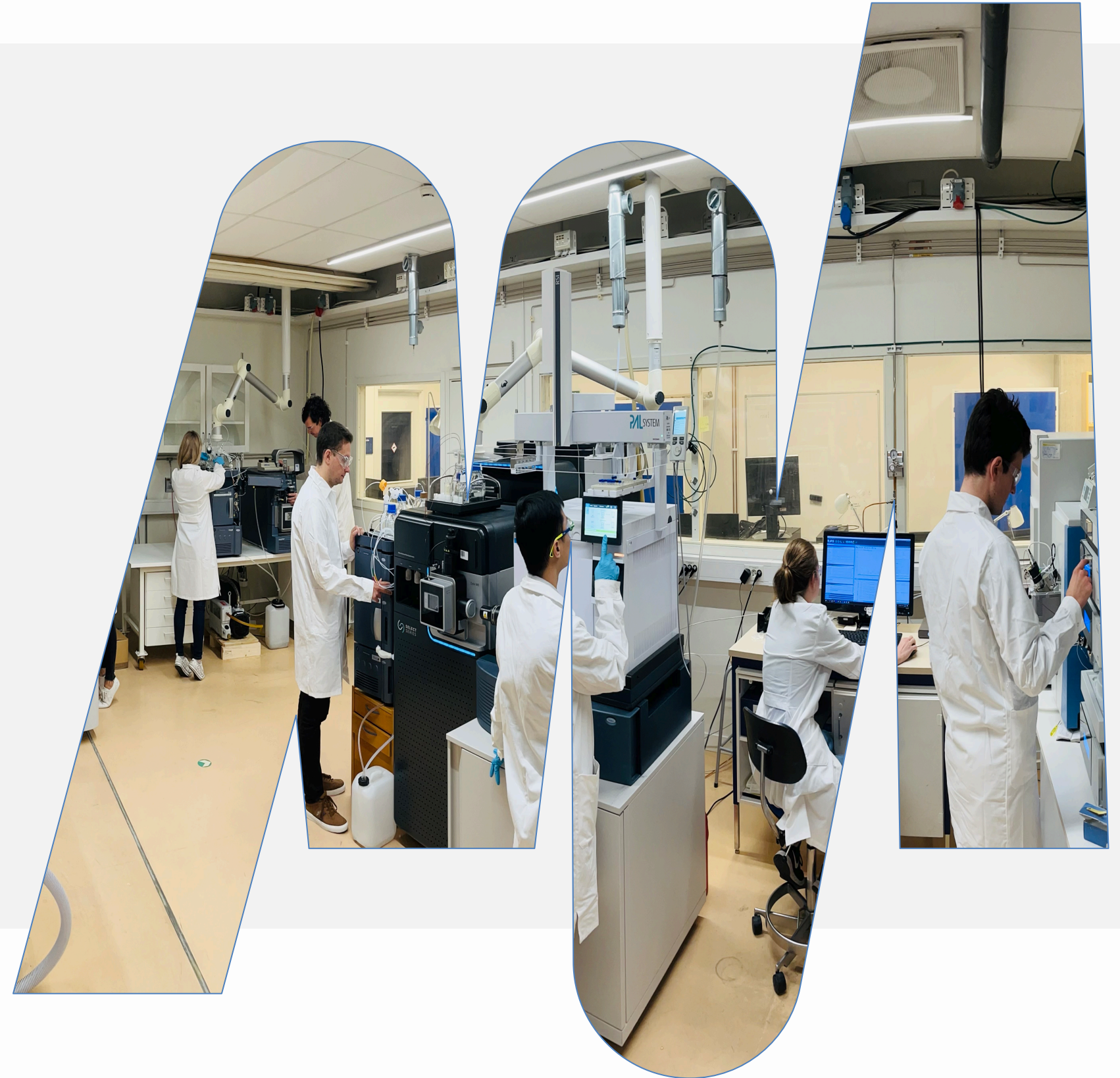
take-home
message

metrics matter

performance needs to be
measured with metric
relevant for the task

interlab

- organized by
SLV, FOI, SVA
- aim
detect and identify
potentially hazardous
chemicals in drinking water
under time stress



interlab

● full sample activity
BioCell Analytica

Machine Translated by Google



Table 1. The following BEQ values were measured in the current samples:

	Nrf2 activity	Anti�AR activity	AR activity	ER activity	AhR activity
	�g/L (tBHQ equivalents)	ng/L (OHF equivalents)	ng/L (DHT equivalents)	pg/L (E2 equivalents)	ng/L (TCDD equivalents)
Reference to sample 1	<LOD	<LOD	<LOD	<LOD	<LOD
Sample 1	<LOD	<LOD	79300	784	0.0814
Reference to sample 2	21.1	73.6	<LOD	21.7	<LOD
Sample 2	992	2670	<LOD*	<LOD*	<LOD*
Detection limit	8.34	43.8	0.122	12.5	0.0196
detection limit*			6.93	50.0	0.156

Table 2. Genotoxicity.

	Genotoxic?
Reference to sample 1	No
Sample 1	No
Reference to sample 2	No
Sample 2	Could not be determined*

*Due to extensive cytotoxicity, despite repeated analyses, it could not be determined whether sample 2 was genotoxic or not. The sample was tested down to the concentration REF 12.5, but even then was too cytotoxic to be able to determine if it was genotoxic.

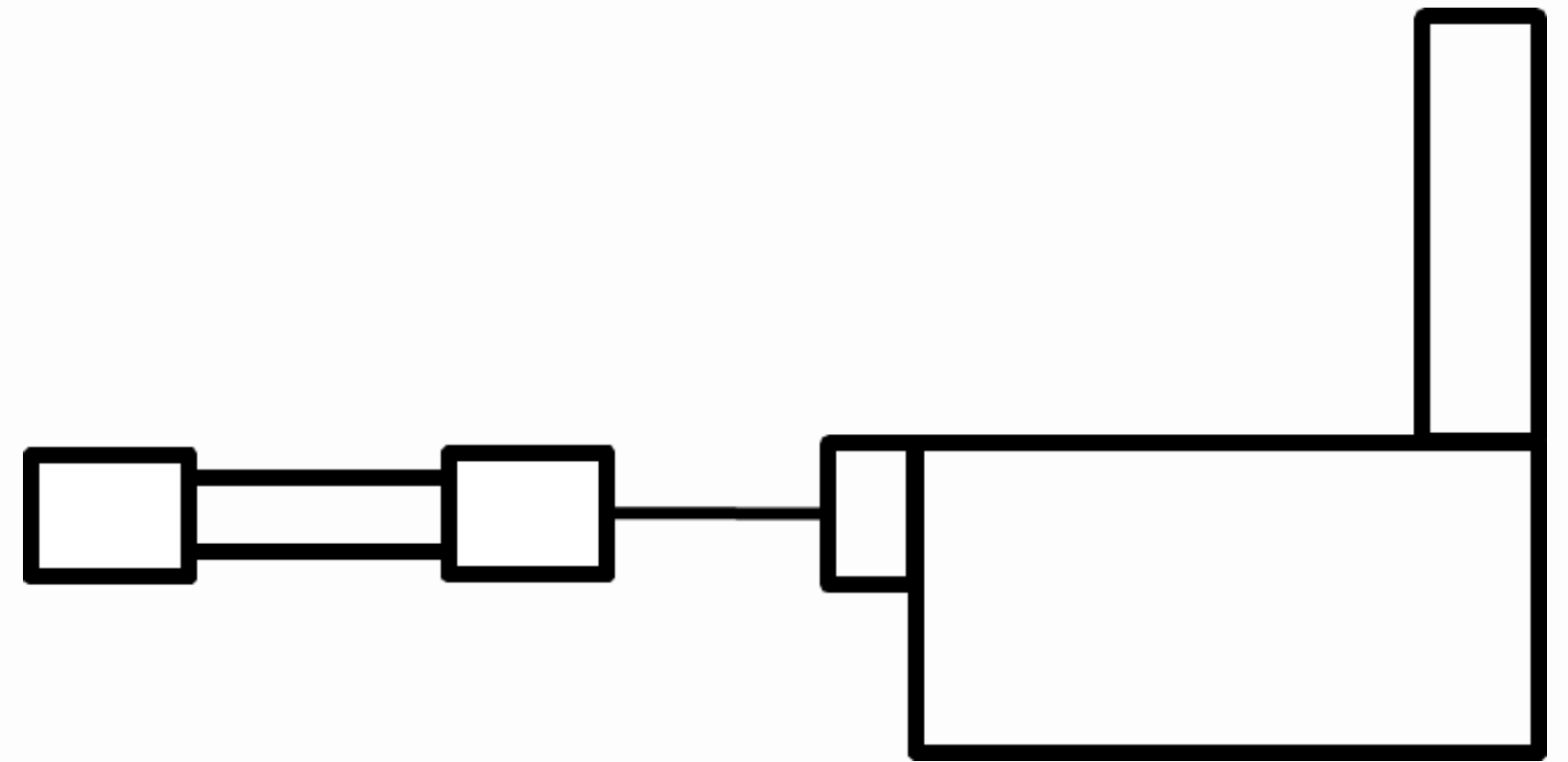
BioCell Analytica Uppsala AB
Ulls v g 29C, 756 51 Uppsala

biocellanalytica.se
kontakt@biocellanalytica.se

Sandberg, Rahu, et al. in preparation

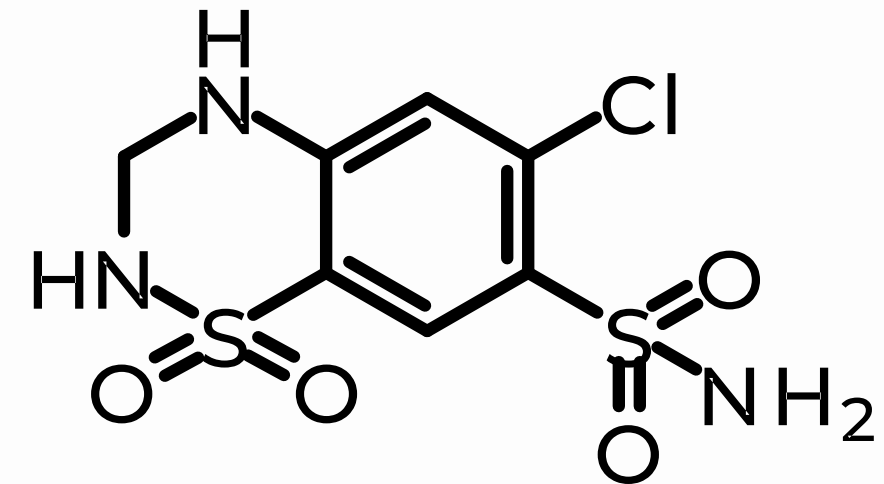
interlab

- full sample activity
BioCell Analytica
- LC/HRMS
4700 features



interlab

- full sample activity
BioCell Analytica
- LC/HRMS
4700 features
- predicted AhR activity
55 features



alternative approaches

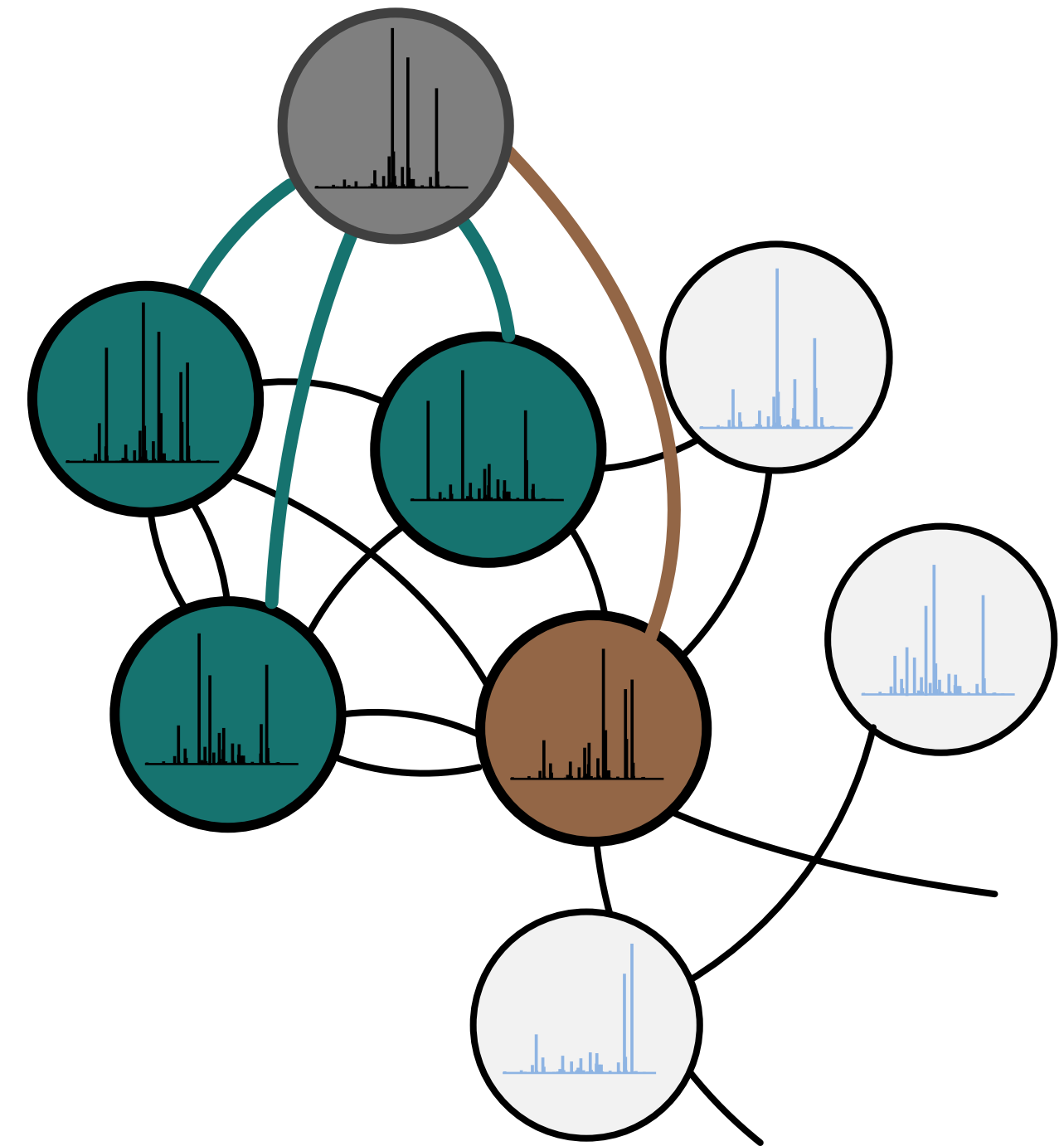
- molecular networking
- conformal predictions



molecular networking

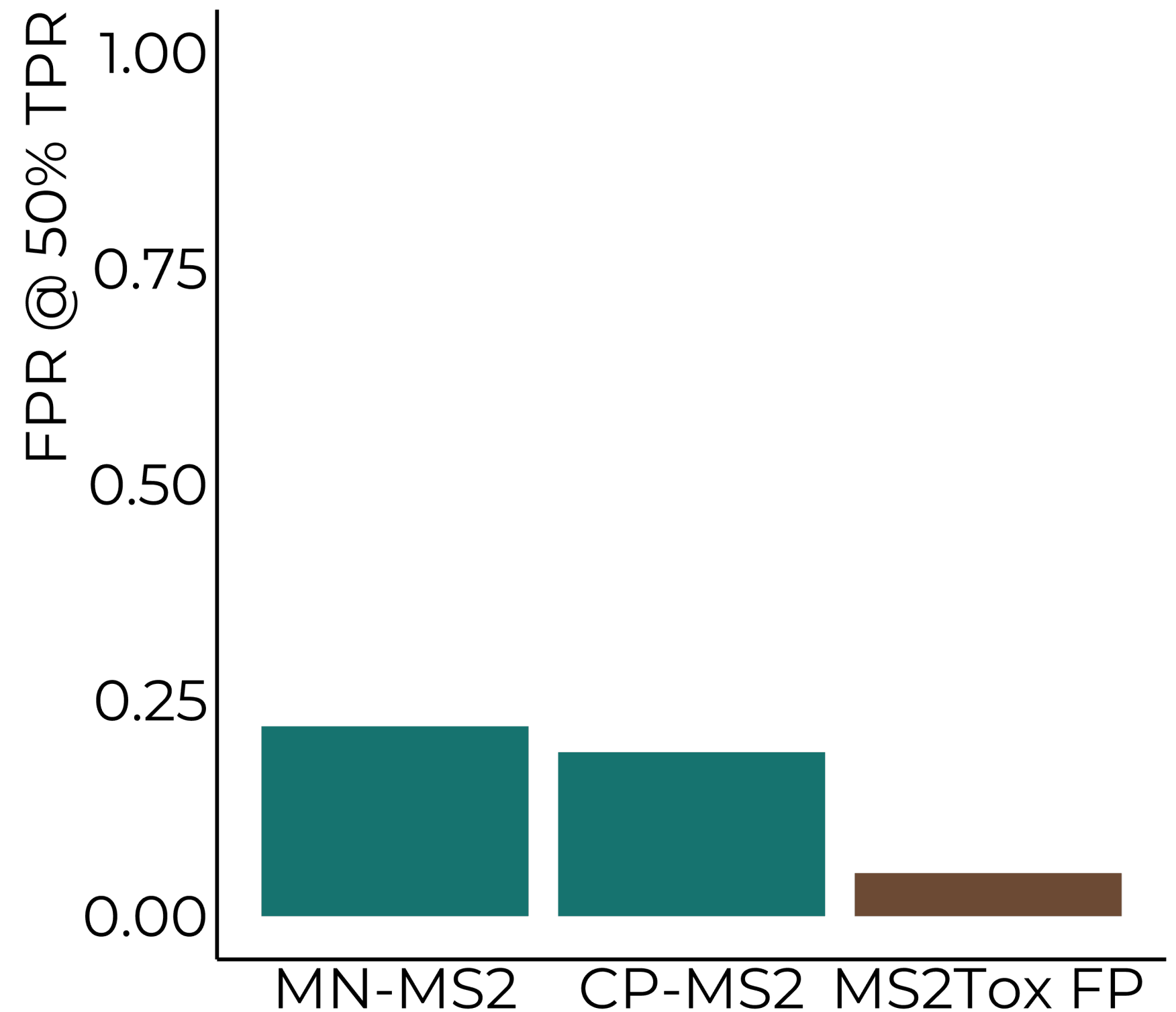
- similarity of MS² spectra
MS2DeepScore
- consensus vote
immediate neighbors

- Active
- Inactive
- Inconclusive
- Unknown



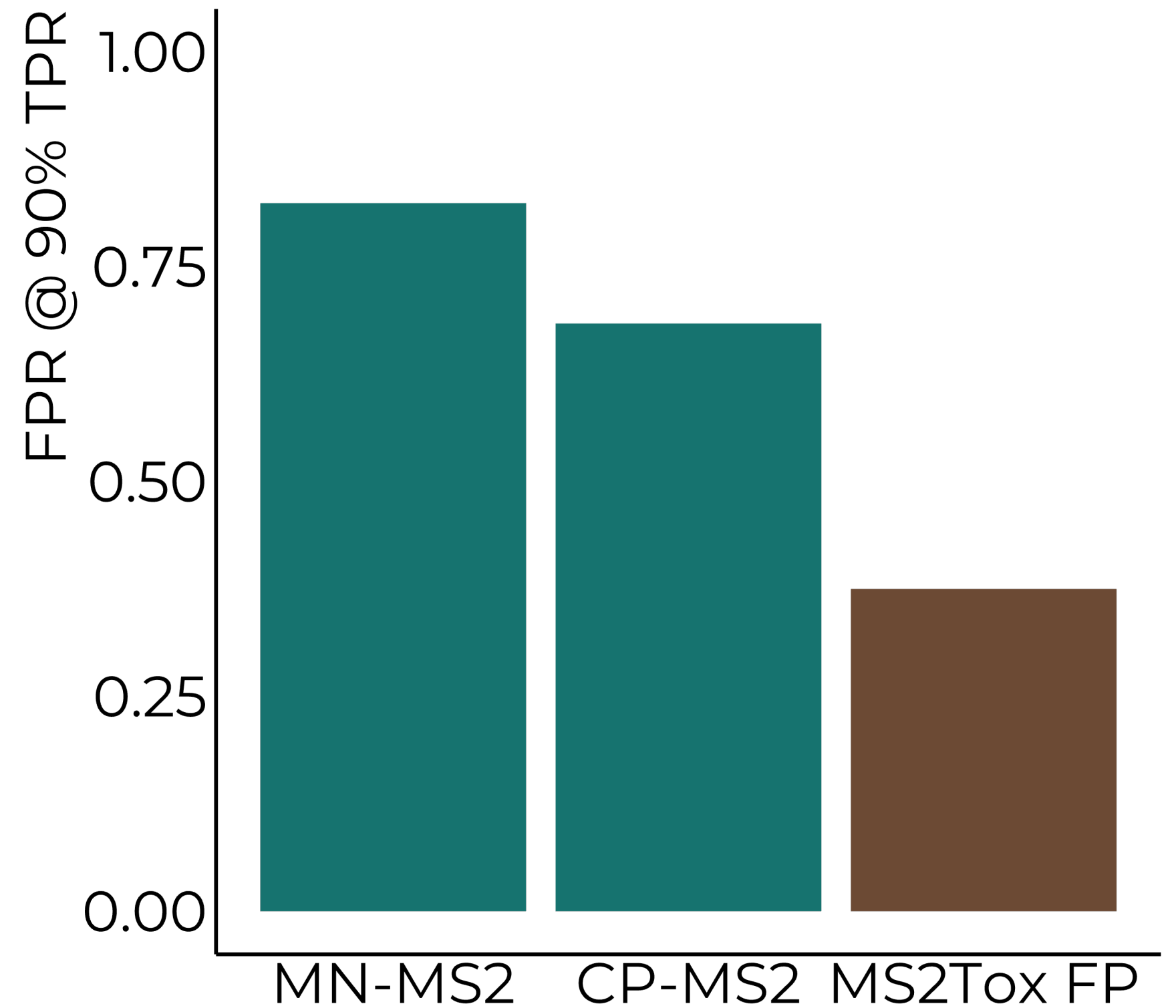
comparison of approaches

- unsupervised approach yields high FPR
- MS² similarity is insufficient

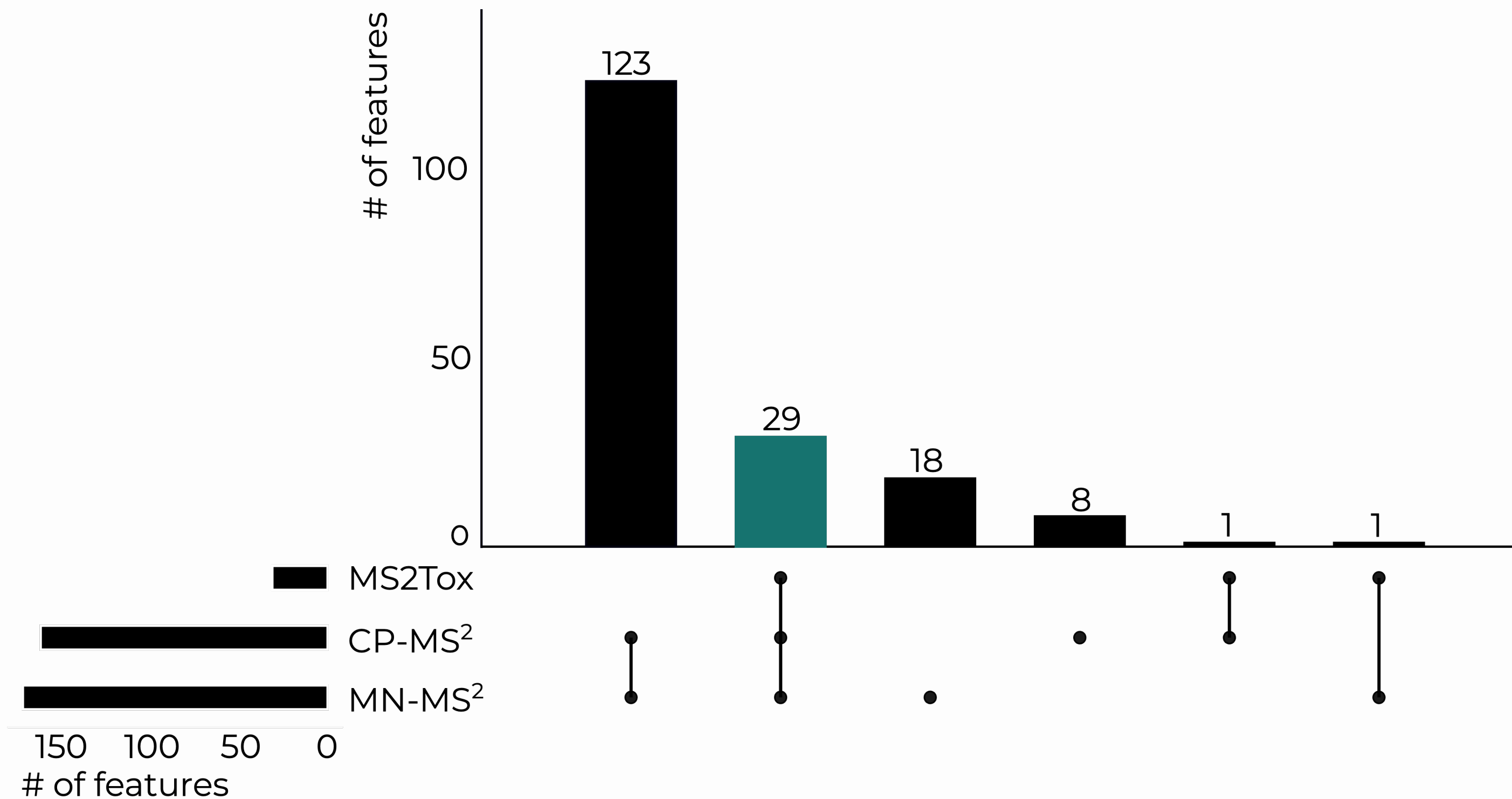


comparison of approaches

- unsupervised approach yields high FPR
- MS² similarity is insufficient

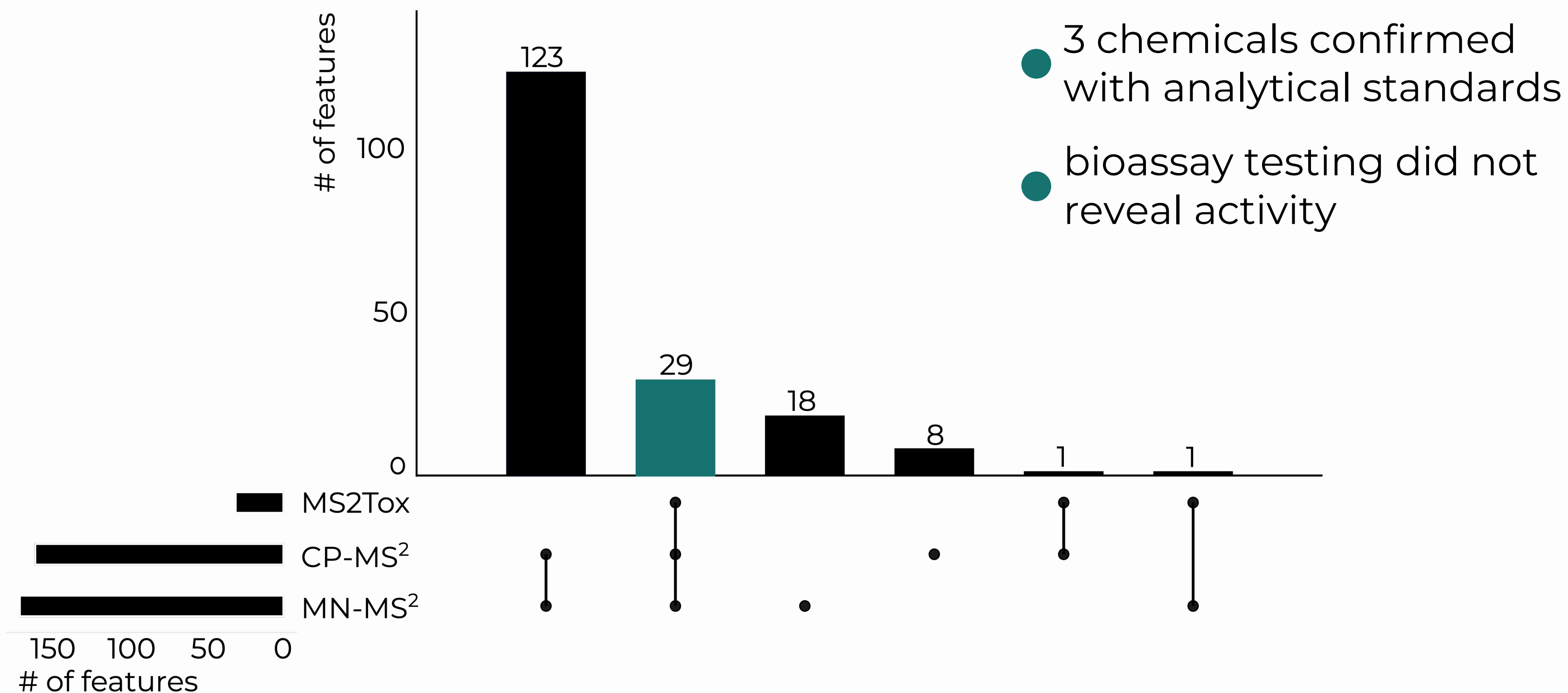


combining approaches



Kreutzer et al. in preparation

combining approaches

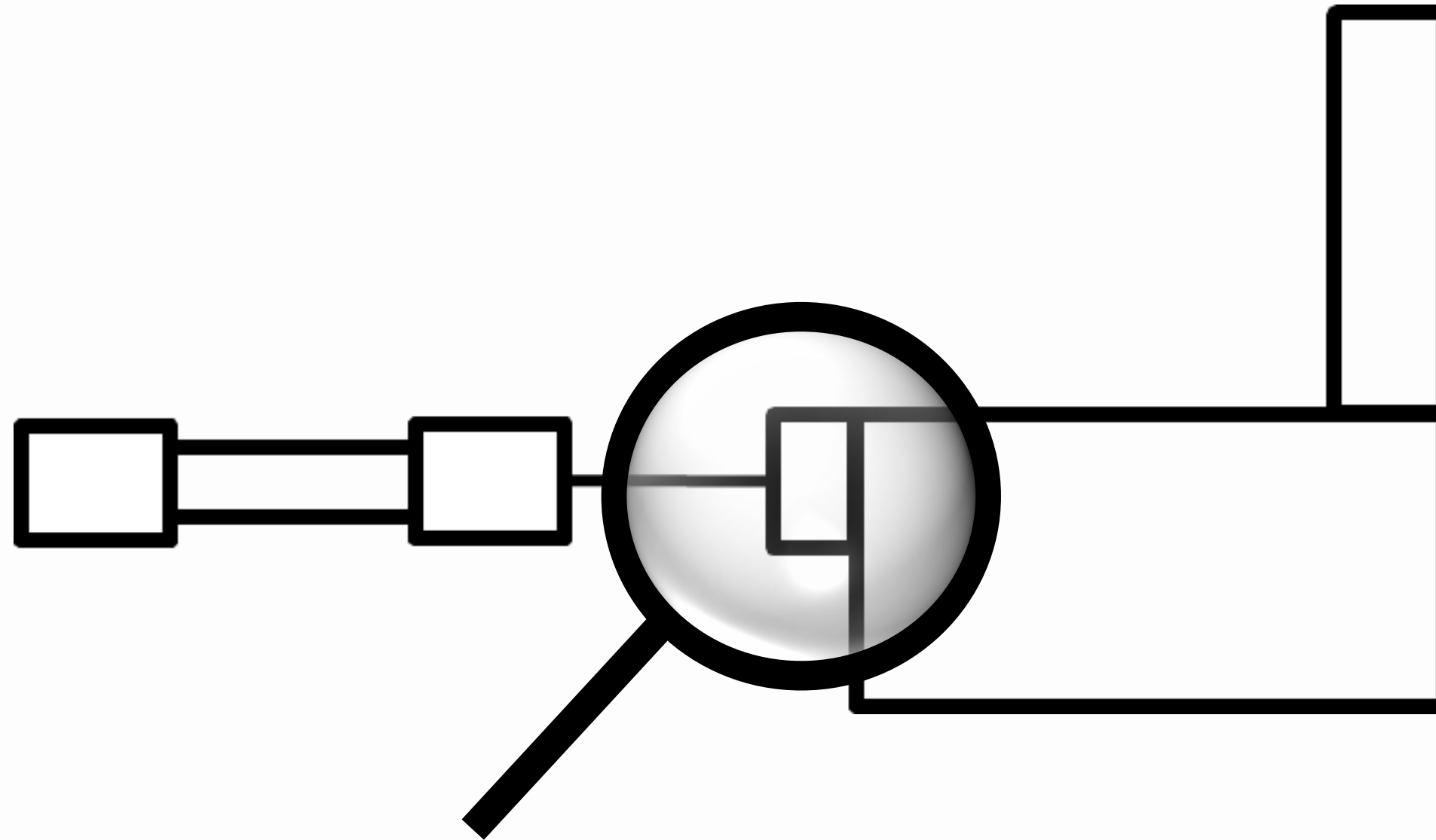


quantification

- lack of analytical standards
- different ionization efficiencies

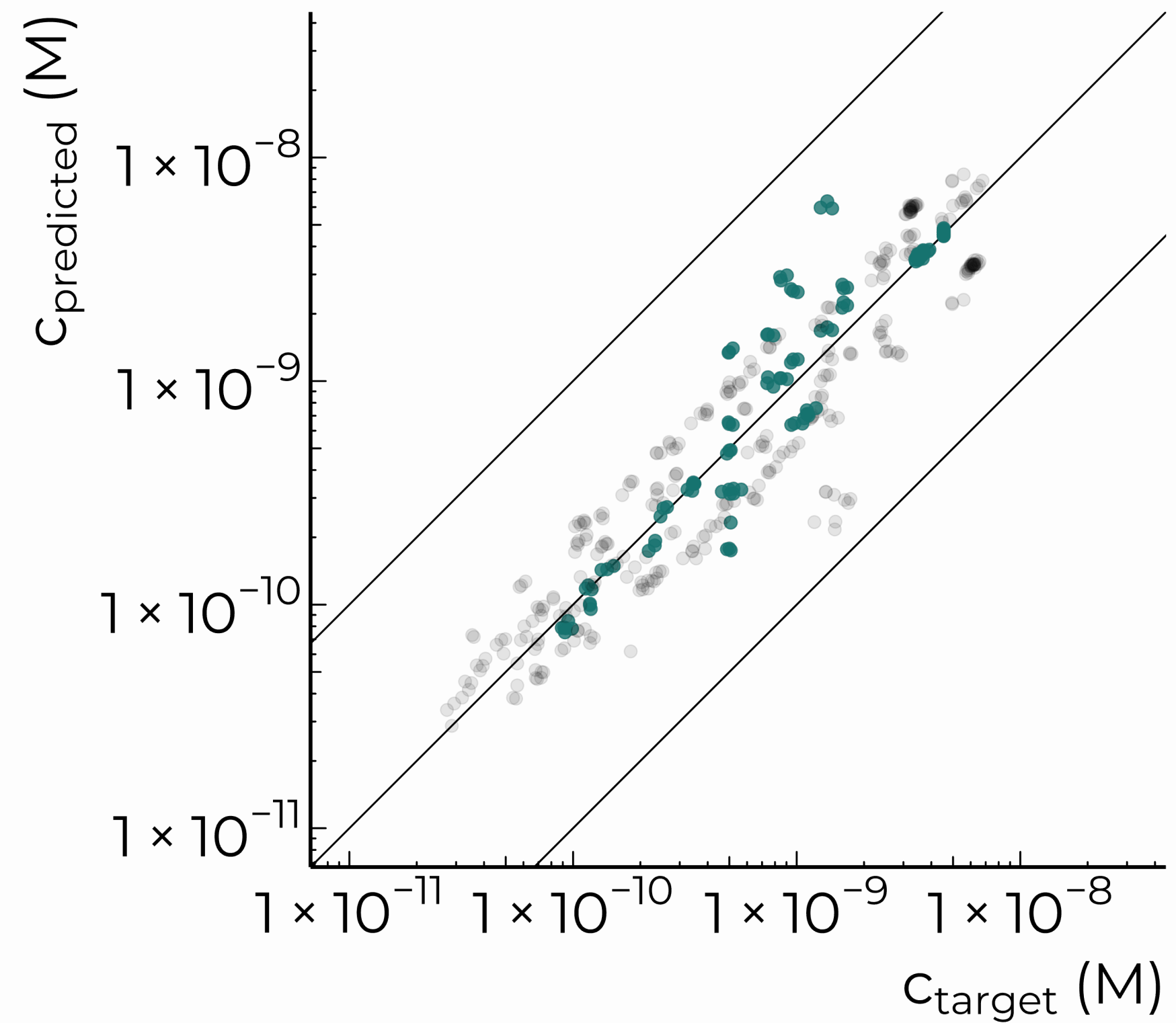


electrospray ionization



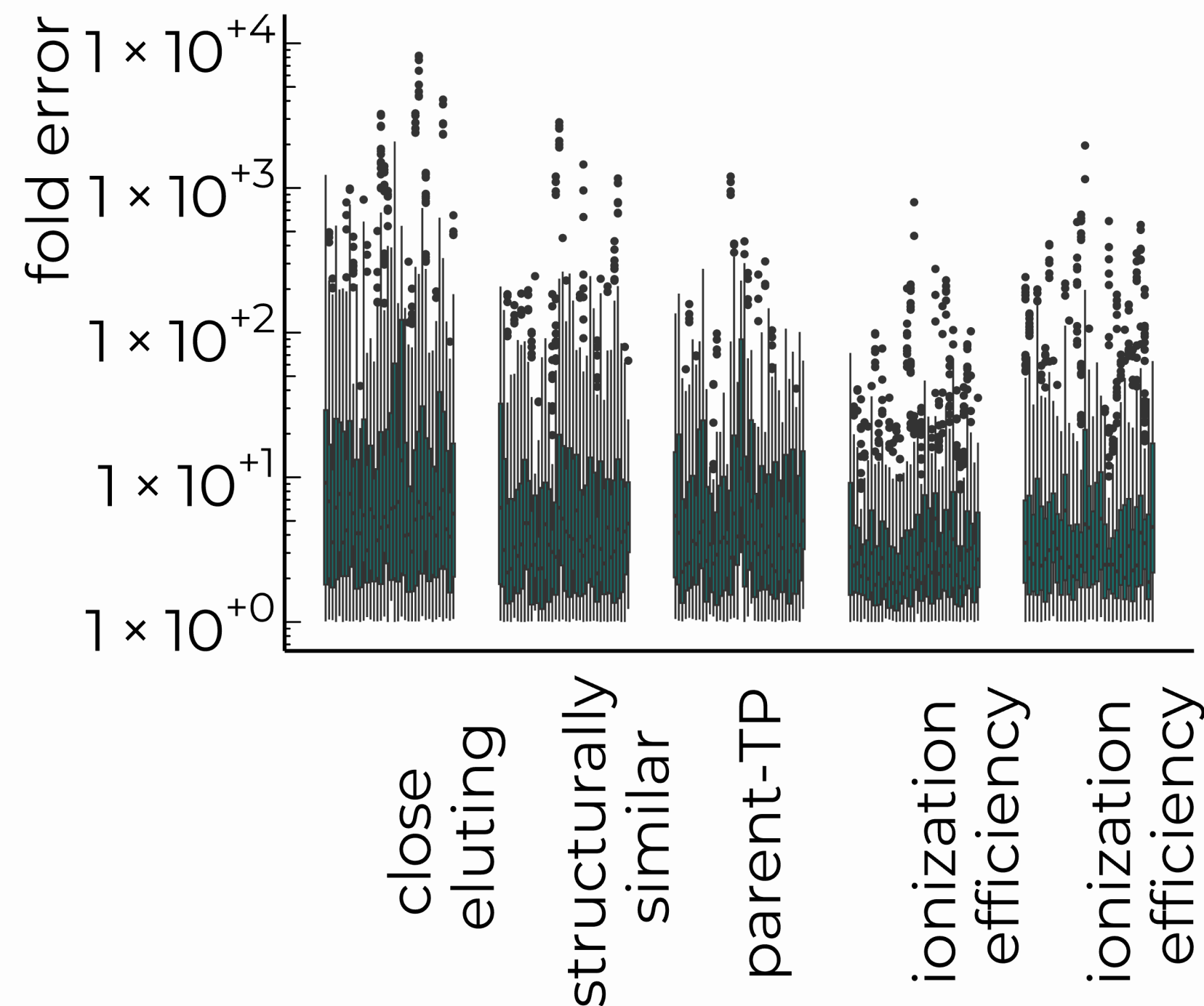
quantification

● from structure
mean error 1.7x



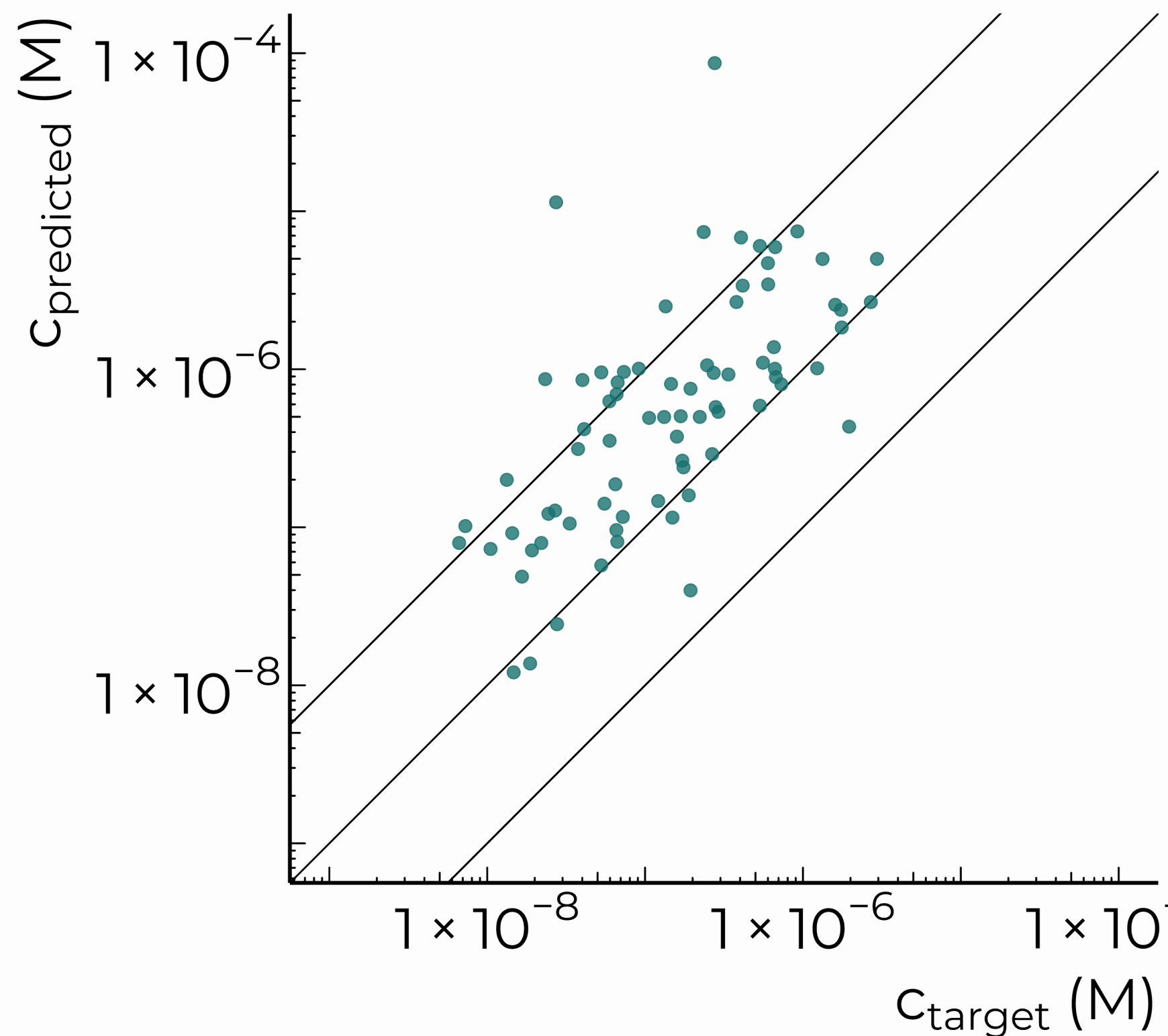
quantification

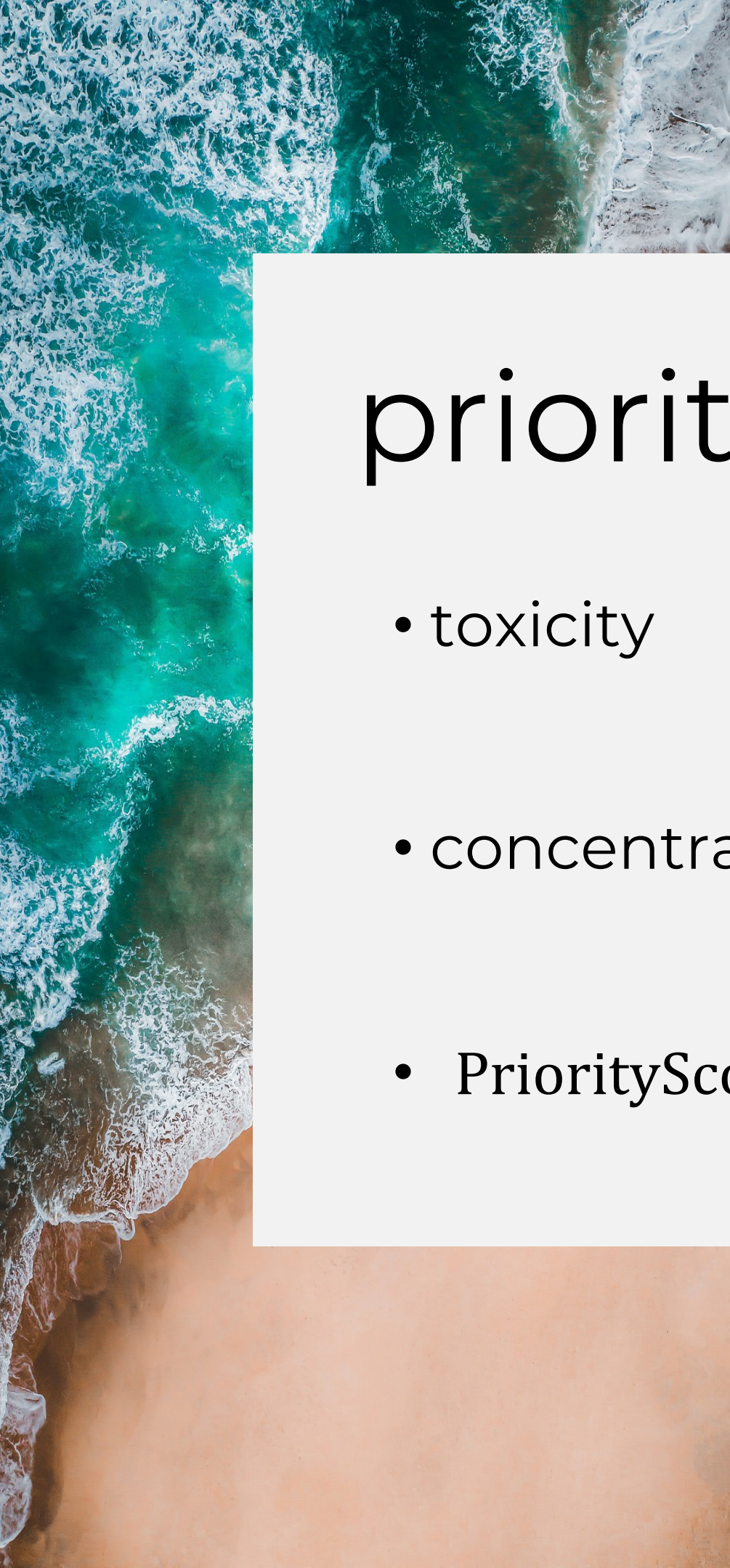
- from structure
mean error 1.7x
- tested in interlab
RMSE 0.88 log(M)



quantification

- from structure
mean error 1.7x
- tested in interlab
RMSE 0.88 log(M)
- from MS² data
mean error 9.5x





prioritization

- toxicity

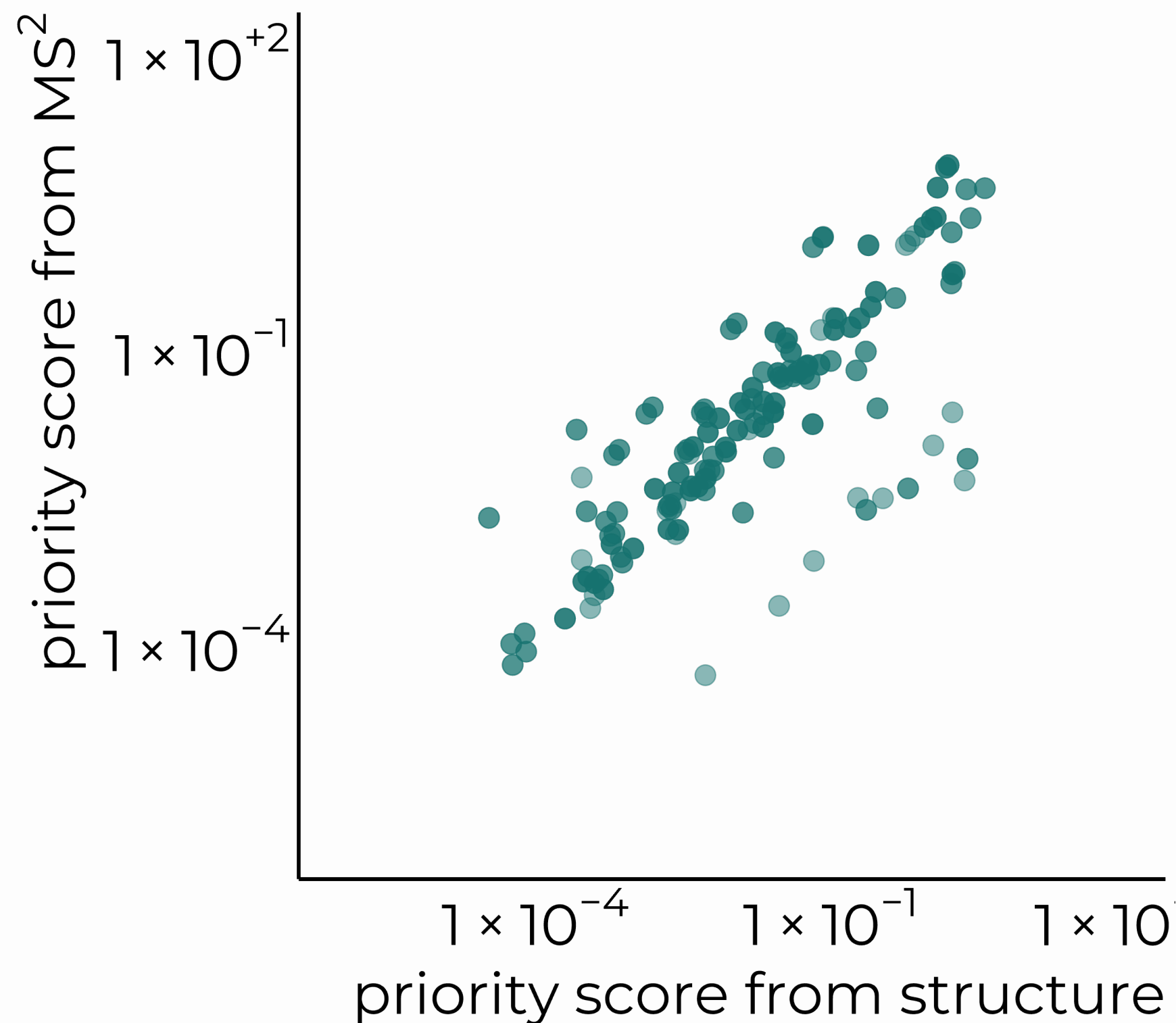
- concentration

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

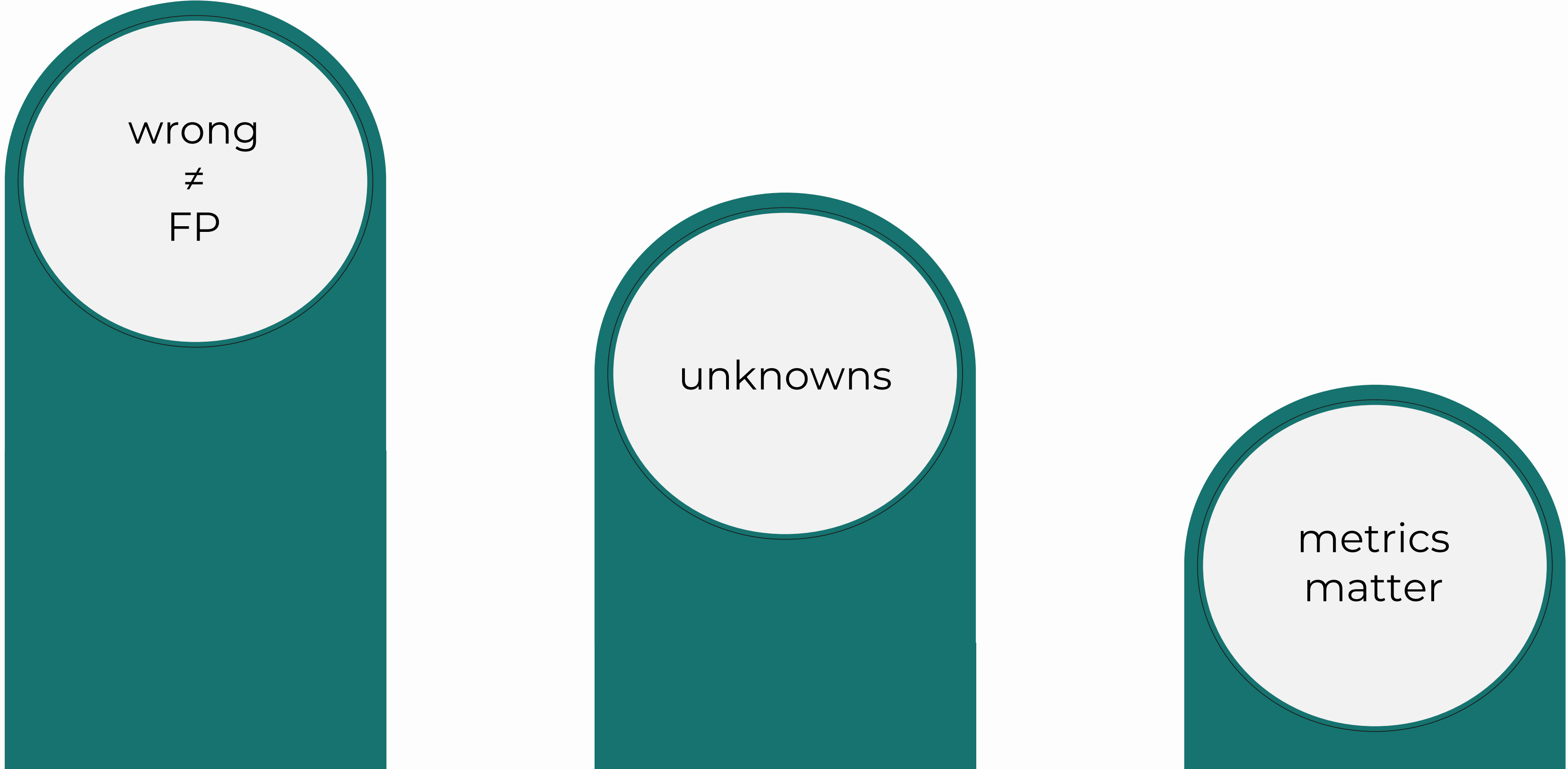


priority score

- fish LC_{50} & concentration
- MS^2 and structure based predictions agree
- acquisition mode impacts



take-home messages

The image features three teal-colored pillars of varying heights, each with a circular top section. The pillars are arranged horizontally. The leftmost pillar is the tallest, the middle one is medium height, and the rightmost one is the shortest. Each pillar's circular top contains text.

wrong
≠
FP

unknowns

metrics
matter

anneli kruve

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FORMAS



CARL TRYGGERS
STIFTELSE
FÖR VETENSKAPLIG FORSKNING

