Supporting information

Deep Learning assisted Peak Curation for large scale LC-MS Metabolomics

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Figure S1: Example of CS peaks detected by MZmine on dataset 2 and predicted as High quality, Acceptable quality and Noise by NeatMS.



Figure S2: Density plots representing the peak width distribution of the 3 peak classes of dataset 1 analysed using MZmine and NeatMS TL model. a. Width distribution of all peaks. b. Width distribution of peaks from the upper tercile of peak area. c. Width distribution of peaks from the intermediate peak area tercile. d. Width distribution of peaks from the lower peak area tercile.



Figure S3: Peak number and internal standard recovery for different tools on the different dilution points of dataset 2. a. Average number of peaks for each dilution point before and after using NeatMS. Only the high quality and acceptable quality classes of NeatMS are displayed here. The difference between the total number of peaks detected by MZmine and XCMS and the number of peaks included in NeatMS high and acceptable quality corresponds to peaks predicted as Noise and peaks rejected by the minimum scan number filter (set to the default value of 5). b. Number of (non diluted) internal standard compounds recovered by the different tools and the details of their predicted classes for each dilution point.



Figure S4: Average detected peak number and standard compound recovered by peakonly on dataset 1 when modulating peakonly "ROI minimum points" parameter. As previously described by the authors of peakonly, we found that lowering the ROI minimum points parameter significantly increases the number of reported noise peaks. A similar effect can be observed with peakonly "Peak minimum points" parameter (not shown). We conclude that the recommendations of peakonly authors are a good compromise between CS sensitivity and false positive peak detection. The vertical line (in black) represents the parameter that we have used for comparison.



Figure S5: Venn diagram comparing peakonly and the combination XCMS with NeatMS TL model. Numbers are averages over the 20 samples of dataset 1: total number of detected peaks (black), percent of recovered CS (red).

Extracted Ion Chromatogram: 559.2566 - 559.2603 m/z



Figure S6: Example of an extracted ion chromatogram of a peak reported by MZmine and classified as Noise by NeatMS matching a CS in dataset 1.

Table S1: Parameters of the different tools used for dataset 1 and 2. Only non default parameters are shown. Peakonly stable and published version 0.1 was used through command lines. XCMS version 3.10.0 was used in R 4.0.1. MZmine was used under version 2.53. NeatMS was used under version 0.6 with default parameters only.

Parameter	Value			
MZmine (Dataset 1 & 2)				
ADAP chroma	atogram builder			
Min group size in # of scans	5			
Group intensity threshold	5 x 10 ²			
Min highest intensity	1 x 10 ³			
m/z tolerance (m/z)	1 x 10 ⁻²			
Wavelets (ADAP) chron	matogram deconvolution			
S/N threshold	10			
S/N estimator	Intensity window SN			
Min feature height	1 x 10 ³			

Peak duration range	0.02 - 1.0				
RT wavelet range	0.001 - 0.05				
XCMS CentW	XCMS CentWave (Dataset 1)				
Min-Max peak width	3-85.16				
ppm	17				
mzdiff	-0.01145				
XCMS CentWave (Dataset 2)					
Min-Max peak width 3-81					
ppm	11.75				
mzdiff	-0.016				
Peakonly (D	ataset 1 & 2)				
Delta mz	0.01				
ROI minimum points	12				
Peak minimum points	6				

Table S2: Performance of peakonly with NeatMS using dataset 1 and the two models: Average number of peaks across 20 samples, average percentages of the 80 SC recovered using peakonly and different models. The input row shows the results returned by peakonly alone, other rows show the details of the three peak classes given by NeatMS. The total number of peaks after classification is smaller than the input due to the application of a minimum scan number filter that NeatMS uses (default value of 5 is used).

		Peakonly data with NeatMS TL model	Peakonly data with NeatMS PT model
	Input	1907	1907
Peak number	Classified	1724	1724
	High Quality	457	730

Acceptable Quality 747		747	684
	Noise	520	310
	Input & classified	79.44%	79.44%
CS found	High Quality	44.75%	74.69%
	Acceptable Quality	33.06%	3.81%
	Noise	1.44%	0.75%

Table S3: NeatMS resuls using TL model on dataset 1 analysed using MZmine with and without the gapfilling step. Peak number: average number of peaks across 20 samples. CS found: average percentages of the 80 CS recovered.

		MZmine without gapfilling	MZmine with gapfilling	
Peak number	Input	6977	21916	
	Classified 5505		14985	
	High quality 1069		1788	
	Acceptable	1945	2487	
	Noise	2491	10710	
CS found	Input & classified	94.25%	95.00%	
	High quality	79.31%	81.81%	
	Acceptable	11.25%	9.63%	
	Noise	3.69%	3.56%	

Table S4: Confusion matrix of DNN and NeatMS number of peaks per class on dataset 1 analyzed using MZmine with gapfilling. TL model was used for NeatMS. Numbers are averaged over 20 samples. No match represents peaks without an exclusive matching peak or without any matching peak at all. Column entries represent NeatMS results, row entries represent DNN results.

	High quality	Acceptable	Noise	No Match	Total
Good	832	1351	1188	5292	8663

Bad	305	346	5200	10770	16621
No Match	651	790	4322	/	
Total	4275		10710		

Table S5: Confusion matrix of DNN and NeatMS CS recovery per class on dataset 1 analyzed using MZmine with gapfilling. TL model was used for NeatMS. Numbers are averaged over 20 samples. No match represents CS missed by the specific method but found using the other. Column entries represent NeatMS results, row entries represent DNN results.

	High quality	Acceptable	Noise	No Match	Total
Good	73.31%	7.56%	2.69%	1.44%	85.00%
Bad	1.31%	0.25%	0.81%	0.12%	2.49%
No Match	7.20%	1.81%	0.06%	/	
Total	91.44%		3.56%		

Table S6: Confusion matrix of MetaClean and NeatMS number of peaks per class on dataset 1 analyzed using XCMS with alignment and gapfilling. TL model was used for NeatMS. Numbers are averaged over 20 samples. No match represents peaks without an exclusive matching peak or without any matching peak at all. Column entries represent NeatMS results, row entries represent MetaClean results.

	High quality	Acceptable	Noise	No Match	Total
Pass	1194	356	508	3464	5522
Fail	1079	410	1140	4940	7569
No Match	1024	282	2204	/	
Total	4345		3852		

Table S7: Confusion matrix of MetaClean and NeatMS CS recovery per class on dataset 1 analyzed using XCMS with alignment gapfilling. TL model was used for NeatMS. Numbers are averaged over 20 samples. No match represents CS missed by the specific method but found using the other. Column entries represent NeatMS results, row entries represent MetaClean results.

	High quality	Acceptable	Noise	No Match	Total
Pass	33.13%	0.94%	1.44%	0.05%	35.56%
Fail	39.25%	1.12%	1.25%	6.43%	43.31%
No Match	18.74%	0.5%	0.06%	/	
Total	93.68%		2.75%		