

Electronic Supplementary Information (ESI)

SMoESY: An Efficient and Quantitative Alternative to On-Instrument Macromolecular ^1H -NMR Signal Suppression†

Panteleimon G. Takis,^{*a,b} Beatriz Jiménez,^{a,b} Caroline J. Sands,^{a,b} Elena Chekmeneva^{a,b} and Matthew R. Lewis^{a,b}

^aSection of Bioanalytical Chemistry, Division of Systems Medicine, Department of Metabolism, Digestion and Reproduction, Imperial College London, South Kensington Campus, London, SW7 2AZ, United Kingdom

^bNational Phenome Centre, Department of Metabolism, Digestion and Reproduction, IRDB Building, Imperial College London, Hammersmith Campus, London, W12 0NN, United Kingdom

*Correspondence and any requests for materials should be addressed to Dr. Panteleimon G. Takis,
e-mail: p.takis@imperial.ac.uk

Table of Contents

| | |
|---|-----------|
| <i>SMoESY impact on time and cost for the NMR metabolomics pipeline</i> | 3 |
| Figure S1..... | 3 |
| <i>SMoESY approach in its simplest form</i> | 4 |
| Figure S2..... | 4 |
| <i>Testing metabolites intra-reproducibility by artificial mixtures of metabolites</i> | 5 |
| Figure S3..... | 8 |
| Table S1. | 9 |
| <i>3D scores plots of 1D-NOESY vs SMoESY PCA analyses</i> | 11 |
| Figure S4..... | 11 |
| <i>Correlation of SMoESY vs. CPMG spectra for 994 plasma-EDTA samples</i> | 12 |
| Figure S5..... | 25 |
| <i>Examples of SMoESY vs. CPMG and NOESY signals</i> | 26 |
| Figure S6..... | 26 |
| <i>SMoESY vs CPMG spectral bins correlation – spiking experiments in real plasma matrix</i> | 27 |
| Figure S7..... | 31 |
| <i>Comparison between SMoESY and signal deconvolution approaches for absolute quantification</i> | 32 |
| Figure S8..... | 32 |
| <i>An overview of the SMoESY_platform graphical user interface (GUI) toolbox</i> | 33 |
| Figure S9..... | 33 |
| <i>Computer code for the calculation of the Pearson correlation coefficients</i> | 34 |
| Data S1..... | 34 |
| <i>17 spiked metabolites 11 different concentrations spiked in real plasma sample</i> | 35 |
| Table S2 | 35 |
| <i>Example of a smoothing filter for denoising SMoESY</i> | 36 |
| Figure S10..... | 36 |

SMoESY impact on time and cost for the NMR metabolomics pipeline

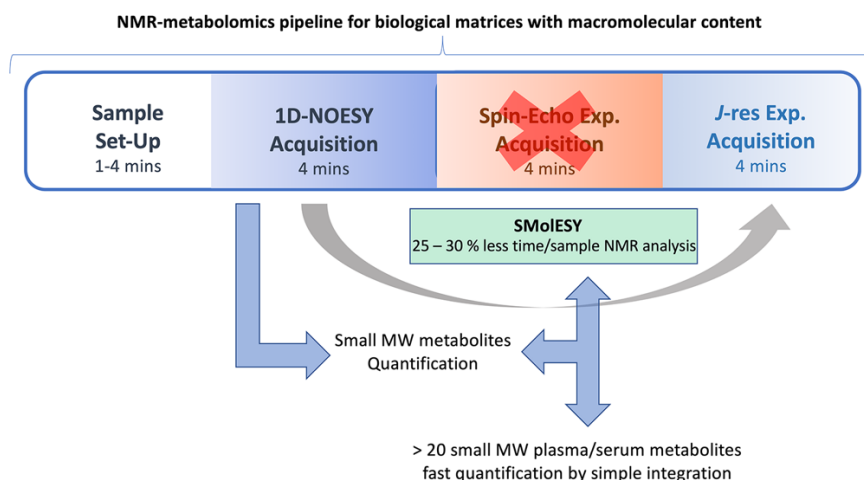


Figure S1. The experimental scheme of NMR based metabolomics pipeline for biofluids with macromolecular content (e.g. proteins, lipoproteins, lipids etc.) – SMoESY contribution. The total amount of experimental time is up to 16 minutes. The 1D-NOESY experiment [containing both macromolecules and low molecular weight (MW) metabolites signals] is followed by a one-dimensional spin-echo experiment (e.g. CPMG), which suppresses the broad signals of macromolecules by T_2 relaxation times filtering. Finally, a pseudo-2D experiment is acquired, J -res, employed for the assignment facilitation of low MW metabolites overlapped signals. As shown, spin-echo experiments (given the limited acquisition time) do not efficiently attenuate broad signals of macromolecules, and they cannot be employed for absolute quantification of metabolites as they are highly modulated by T_2 values (the same stands for the projections of J -res experiments that are highly modulated by the J -coupling constant values). On the other hand, SMoESY production requires < 1 sec per spectrum, supersedes spin-echo experiments in broad signals attenuation as well as maintains its quantitative ability for low MW metabolites absolute quantification. Consequently, a substitution of a spin-echo experiment by SMoESY could lead to up to 30% decrease of acquisition time per spectrum (or up to 30% increase of acquired spectra/samples), whereas it could speed up assignment/quantification of small MW metabolites. Consequently, SMoESY could significantly reduce the cost of NMR analyses for large scale epidemiological studies and provide more details for the low MW metabolites content.

SMoLESY approach in its simplest form

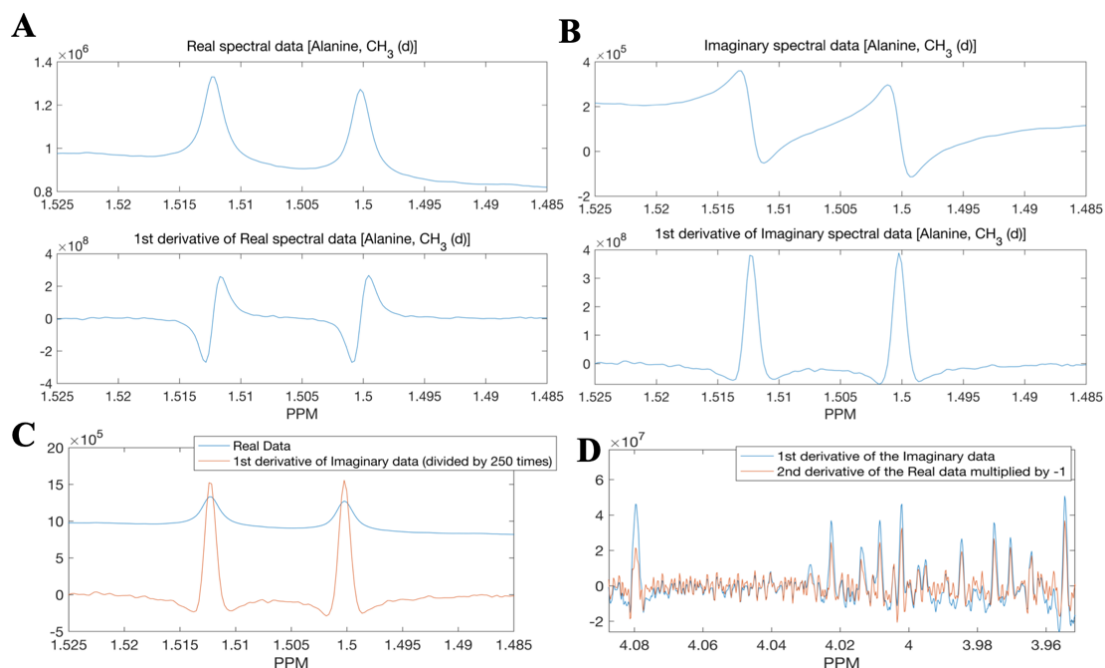
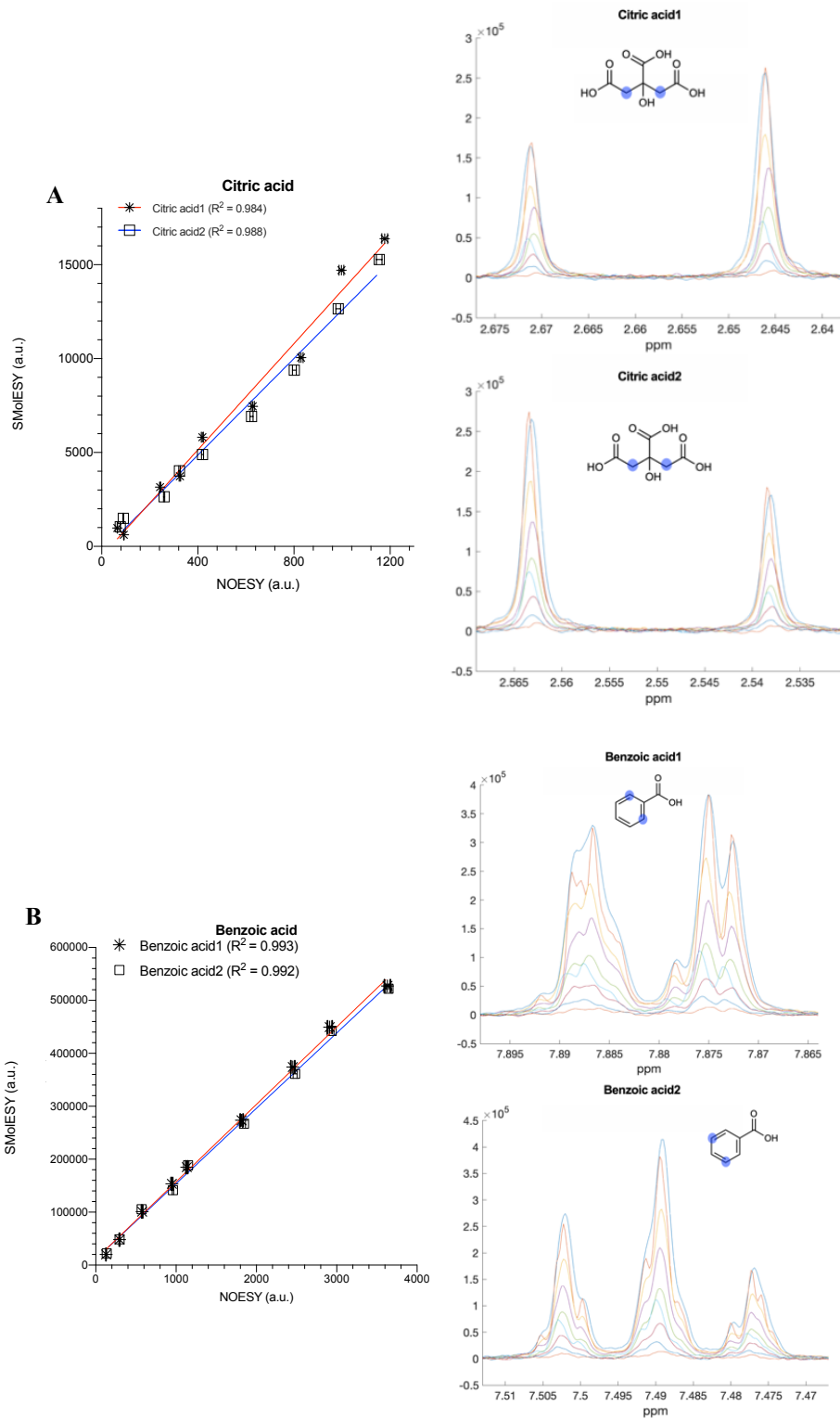
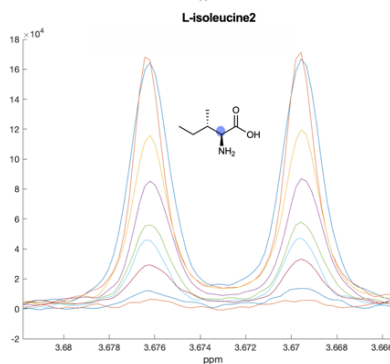
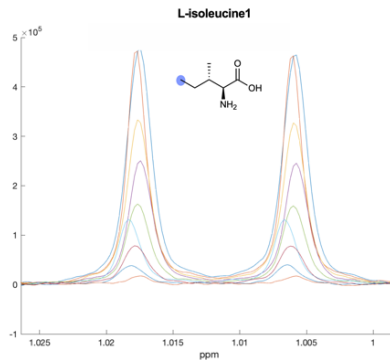
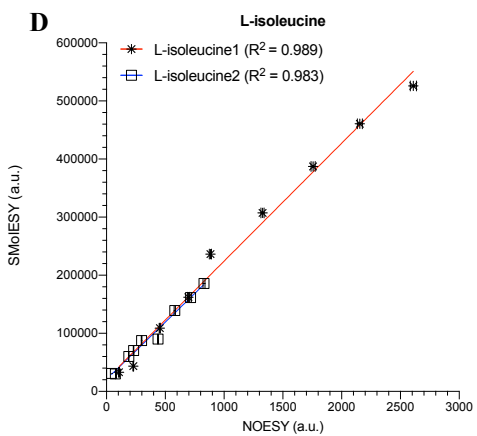
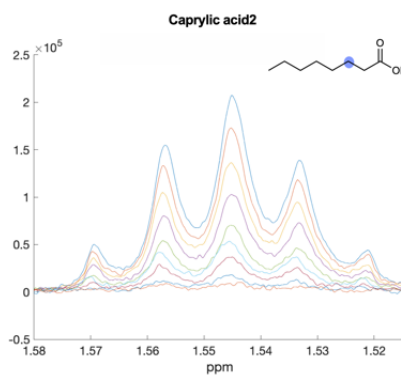
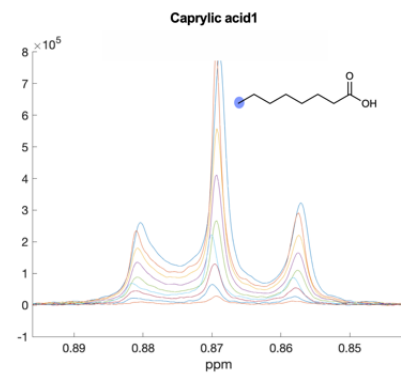
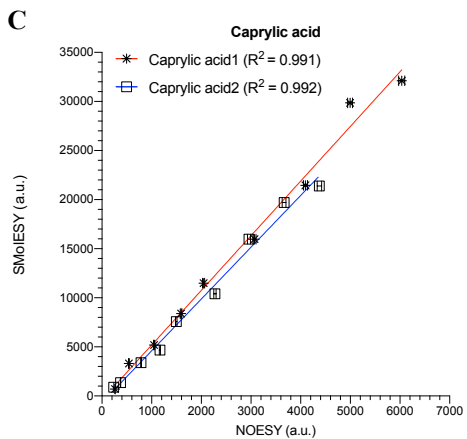


Figure S2. Examples of enhanced spectral resolution by the imaginary NMR spectral part differentiation. (A) The real spectral data (i.e. doublet, d) of the spin system from the -CH₃ group of L-alanine in a typical plasma/serum matrix (upper panel). The 1st numerical derivative of the real data from the L-alanine -CH₃ ¹H-NMR signal (after Fourier transform and phase correction) (bottom panel), produces an antisymmetric signal (positive on one side and negative on the other). (B) The imaginary spectral data of the spin system from the -CH₃ group of L-alanine in a typical plasma/serum matrix (upper panel). In contrary to the real data, the 1st derivative of the imaginary data, due to its gradient (namely positive-negative maxima per signal) (bottom panel), produces a positive transformed signal. (C) Overlaid real and 1st derivative of the imaginary part of the L-alanine -CH₃ doublet spectral regions, show no chemical shifting, without the need of applying any symmetrisation algorithms. The transformed signal from the imaginary spectral data could be immediately employed for any NMR-based metabolomics or analytical study. (D) Comparison between the 2nd derivative of the real data of the NMR spectrum multiplied by -1 (this could be the same for the 2nd power derivative) and the 1st derivative of the imaginary part of the same spectral region, taken from a ¹H-NMR plasma spectrum. It is immediately appreciated that the signal-to-noise ratio of the 2nd derivative of the real spectral data is decreased compared to the 1st derivative of the imaginary part.

Testing metabolites intra-reproducibility by artificial mixtures of metabolites





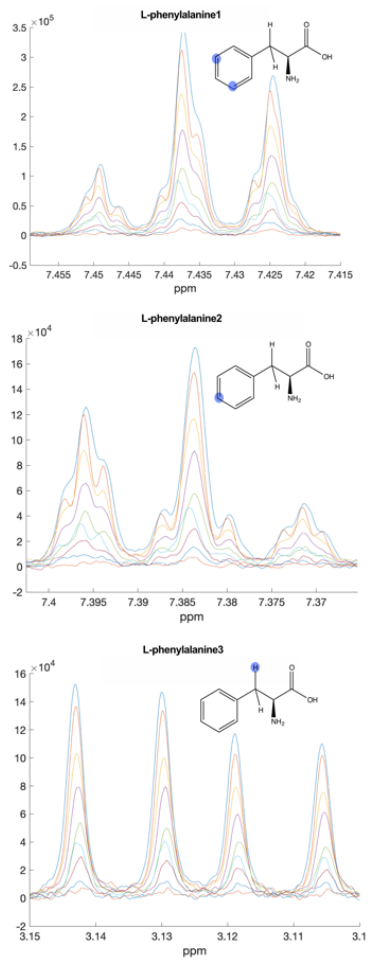
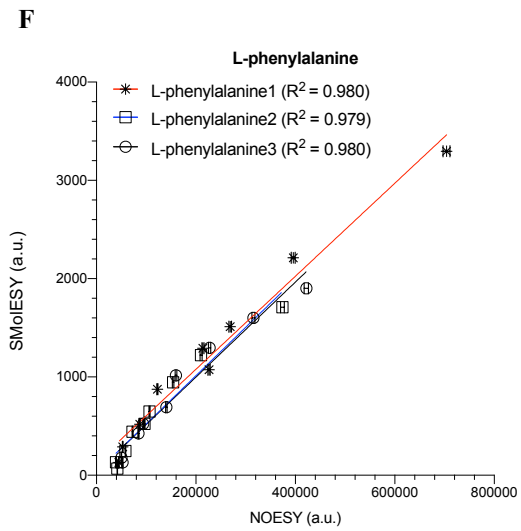
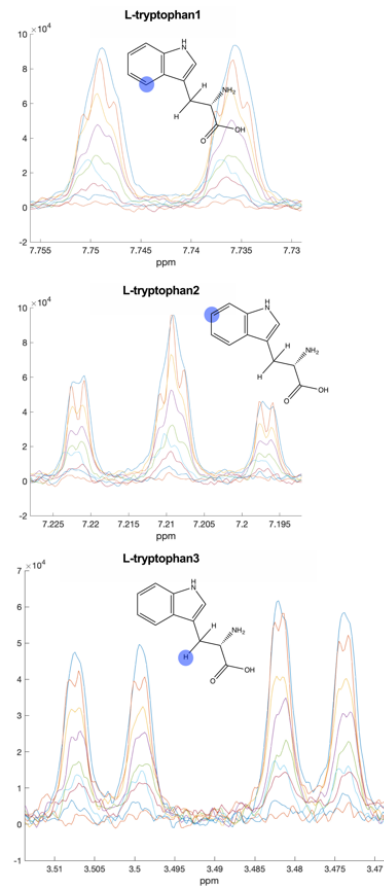
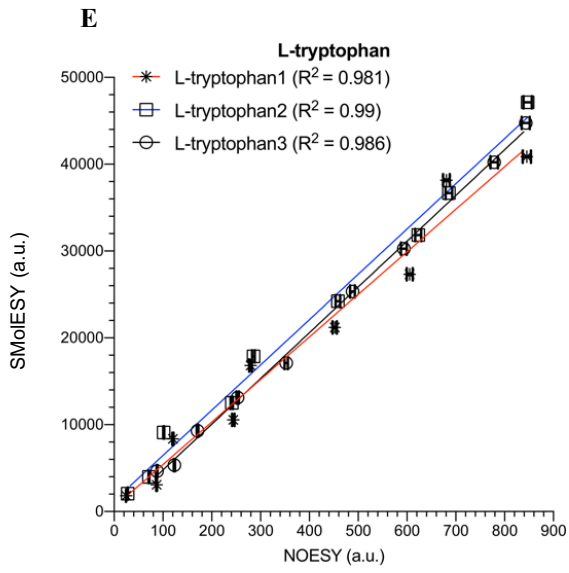


Figure S3. Validation of SMoESY intra-metabolites signals reproducibility. Plots of the fitted linear regression curves ($R^2 \geq 0.98$) based upon the SMoESY versus 1D-NOESY integrals from various ^1H spin systems (highlighted by light blue circles) with different multiplicities (horizontal/vertical error bars point at $\pm 1\%$ error in integration) of (A) Citric acid, (B) Benzoic acid, (C) Caprylic acid, (D) L-isoleucine, (E) L-tryptophan and (F) L-phenylalanine in 9 concentrations in the artificial metabolites mixtures. All slope, intercepts as well as one-way ANOVA tests for the confirmed all intercepts/slopes coincidence for each metabolite (see below Table S1). These results highlight the intra-metabolites SMoESY signals reproducibility.

Table S1. Statistical analyses data for SMoESY intra-metabolites signals reproducibility tests. Statistical data of the fitted linear regression curves for the SMoESY versus 1D-NOESY integrals of NMR signals from various ¹H spin systems from several metabolites in 9 concentrations in the artificial metabolites mixtures. All linear curves statistically pass through the zero point and according to the one-way ANOVA tests both slopes and intercepts of all curves for each metabolite coincide.

| | | | | | | |
|---------------------------------|------------------------|------------------------|-----------------------|-------------------|-------------------|---|
| Best-fit values | Cytidine 1 | Cytidine 2 | Cytidine 3 | Cytidine 4 | Cytidine 5 | one-way ANOVA results |
| Slope | 65.41 | 66.63 | 68.62 | 67.53 | 65.03 | For the slopes F = 0.265 DF _n ¹ = 4, DF _d ² = 35 P = 0.8984 The pooled slope equals 66.59 |
| Y-intercept | 3672 | 30.63 | -3676 | -1133 | -2303 | |
| X-intercept | -56.14 | -0.4597 | 53.57 | 16.77 | 35.41 | |
| 1/slope | 0.01529 | 0.01501 | 0.01457 | 0.01481 | 0.01538 | |
| Std. Error | | | | | | |
| Slope | 1.854 | 3.673 | 2.827 | 2.406 | 3.347 | |
| Y-intercept | 2413 | 4415 | 3544 | 3079 | 4310 | |
| 95% Confidence Intervals | | | | | | |
| Slope | 61.02 to 69.79 | 57.94 to 75.32 | 61.93 to 75.31 | 61.84 to 73.22 | 57.12 to 72.94 | |
| Y-intercept | -2034 to 9377 | -10409 to 10470 | -12056 to 4704 | -8413 to 6148 | -12493 to 7887 | |
| X-intercept | -151.6 to 29.54 | -175.8 to 142 | -74.53 to 163.1 | -97.89 to 116.7 | -134.9 to 175.3 | For the intercepts F = 1.291 DF _n ¹ = 4, DF _d ² = 39 P = 0.2905 The pooled intercept equals 630 |
| Best-fit values | Citric acid 1 | Citric acid 2 | | | | one-way ANOVA results |
| Slope | 14.15 | 12.87 | | | | For the slopes F = 2.165 DF _n ¹ = 1, DF _d ² = 14 P = 0.1633 The pooled slope equals 13.53 |
| Y-intercept | -523.2 | -287.5 | | | | |
| X-intercept | 36.98 | 22.35 | | | | |
| 1/slope | 0.07068 | 0.07772 | | | | |
| Std. Error | | | | | | |
| Slope | 0.6842 | 0.5335 | | | | |
| Y-intercept | 445.7 | 341.9 | | | | |
| 95% Confidence Intervals | | | | | | |
| Slope | 12.53 to 15.77 | 11.61 to 14.13 | | | | |
| Y-intercept | -1577 to 530.6 | -1096 to 520.9 | | | | |
| X-intercept | -41.41 to 102.3 | -44.08 to 78.99 | | | | For the intercepts F = 1.728 DF _n ¹ = 1, DF _d ² = 15 P = 0.2084 The pooled intercept equals 414.2 |
| Best-fit values | Benzoic acid 1 | Benzoic acid 2 | | | | one-way ANOVA results |
| Slope | 146.2 | 142.3 | | | | For the slopes F = 0.91 DF _n ¹ = 1, DF _d ² = 14 P = 0.3563 The pooled slope equals 144.2 |
| Y-intercept | 11191 | 11549 | | | | |
| X-intercept | -76.55 | -81.15 | | | | |
| 1/slope | 0.00684 | 0.007027 | | | | |
| Std. Error | | | | | | |
| Slope | 2.747 | 3.021 | | | | |
| Y-intercept | 5276 | 5887 | | | | |
| 95% Confidence Intervals | | | | | | |
| Slope | 139.7 to 152.6 | 135.2 to 149.4 | | | | |
| Y-intercept | -1285 to 23667 | -2371 to 25468 | | | | |
| X-intercept | -167.9 to 8.492 | -186.6 to 16.02 | | | | For the intercepts F = 1.434 DF _n ¹ = 1, DF _d ² = 15 P = 0.2496 The pooled intercept equals 11376 |
| Best-fit values | Caprylic acid 1 | Caprylic acid 2 | | | | one-way ANOVA results |
| Slope | 5.566 | 5.273 | | | | For the slopes F = 0.9582 DF _n ¹ = 1, DF _d ² = 14 P = 0.3443 The pooled slope equals 5.465 |
| Y-intercept | -378.0 | -671.1 | | | | |
| X-intercept | 67.92 | 127.3 | | | | |
| 1/slope | 0.1797 | 0.1896 | | | | |
| Std. Error | | | | | | |
| Slope | 0.2014 | 0.1998 | | | | |
| Y-intercept | 657.2 | 475.5 | | | | |
| 95% Confidence Intervals | | | | | | |
| Slope | 5.090 to 6.042 | 4.801 to 5.746 | | | | |
| Y-intercept | -1932 to 1176 | -1795 to 453.3 | | | | |
| X-intercept | -227.2 to 325.2 | -92.74 to 318.1 | | | | For the intercepts F = 3.619. DF _n ¹ = 1, DF _d ² = 15 P = 0.0765 The pooled intercept equals -576.7 |
| Best-fit values | L-isoleucine 1 | L-isoleucine 2 | | | | one-way ANOVA results |
| Slope | 203.0 | 197.0 | | | | For the slopes F = 0.08475 DF _n ¹ = 1, DF _d ² = 14 P = 0.7752 The pooled slope equals 202.4 |
| Y-intercept | 21257 | 20708 | | | | |
| X-intercept | -104.7 | -105.1 | | | | |
| 1/slope | 0.004927 | 0.005076 | | | | |
| Std. Error | | | | | | |
| Slope | 8.231 | 9.841 | | | | |
| Y-intercept | 11589 | 4539 | | | | |
| 95% Confidence Intervals | | | | | | |
| Slope | 183.5 to 222.4 | 173.7 to 220.3 | | | | |
| Y-intercept | -6147 to 48661 | -9975 to 31441 | | | | |
| X-intercept | -260.6 to 28.12 | -177.8 to -46.09 | | | | For the intercepts F = 0.15 DF _n ¹ = 1, DF _d ² = 15 P = 0.7039 The pooled intercept equals 20266 |
| Best-fit values | L-tryptophan 1 | L-tryptophan 2 | L-tryptophan 3 | | | one-way ANOVA results |
| Slope | 48.99 | 52.14 | 52.68 | | | For the slopes |
| Y-intercept | 508.7 | 1243 | -483.6 | | | |

| | | | | | | |
|---------------------------------|--------------------------|--------------------------|--------------------------|--|--|---|
| X-intercept | -10.38 | -23.85 | 9.18 | | | F = 0.787 DFn [†] = 2, DFd [‡] = 21 P = 0.4682 The pooled slope equals 51.26 |
| 1/slope | 0.02041 | 0.01918 | 0.01898 | | | |
| Std. Error | | | | | | |
| Slope | 3.162 | 2.022 | 0.8736 | | | |
| Y-intercept | 1456 | 939.6 | 426.8 | | | For the intercepts F = 2.722 DFn [†] = 2, DFd [‡] = 23 P = 0.0869 The pooled intercept equals 445.7 |
| 95% Confidence Intervals | | | | | | |
| Slope | 41.51 to 56.47 | 47.36 to 56.92 | 50.62 to 54.75 | | | |
| Y-intercept | -2935 to 3953 | -978.3 to 3465 | -1493 to 525.7 | | | |
| X-intercept | -92.48 to 53.52 | -71.89 to 17.49 | -10.32 to 27.44 | | | |
| Best-fit values | L-phenylalanine 1 | L-phenylalanine 2 | L-phenylalanine 3 | | | one-way ANOVA results |
| Slope | 0.004732 | 0.004916 | 0.004833 | | | For the slopes F = 0.0642 DFn [†] = 2, DFd [‡] = 21 P = 0.9380 The pooled slope equals 0.004785 |
| Y-intercept | 131.5 | 29.58 | 31.80 | | | |
| X-intercept | -27794 | -6016 | -6580 | | | |
| 1/slope | 211.3 | 203.4 | 206.9 | | | |
| Std. Error | | | | | | For the intercepts F = 0.7911 DFn [†] = 2, DFd [‡] = 23 P = 0.4653 The pooled intercept equals 68.56 |
| Slope | 0.0002872 | 0.0004470 | 0.0003886 | | | |
| Y-intercept | 88.23 | 73.07 | 81.53 | | | |
| 95% Confidence Intervals | | | | | | |
| Slope | 0.004053 to 0.005411 | 0.003859 to 0.005973 | 0.003914 to 0.005751 | | | |
| Y-intercept | -77.10 to 340.1 | -143.2 to 202.4 | -161.0 to 224.6 | | | |
| X-intercept | -81350 to 14700 | -50122 to 25080 | -55497 to 28945 | | | |

[†]DFn: degrees of freedom numerator.

[‡]DFd: degrees of freedom denominator.

3D scores plots of 1D-NOESY vs SMoESY PCA analyses

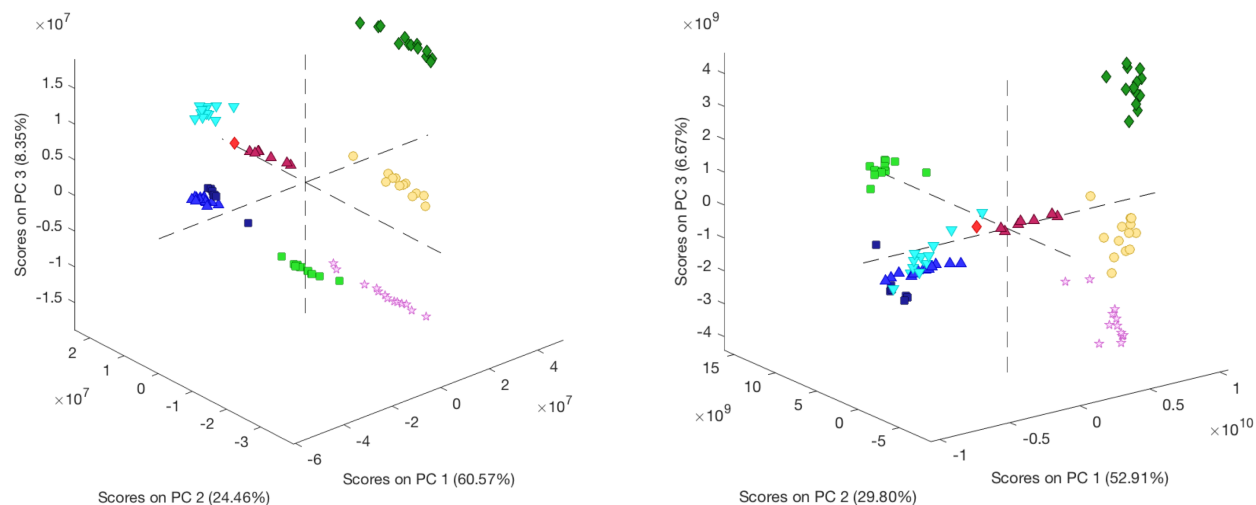
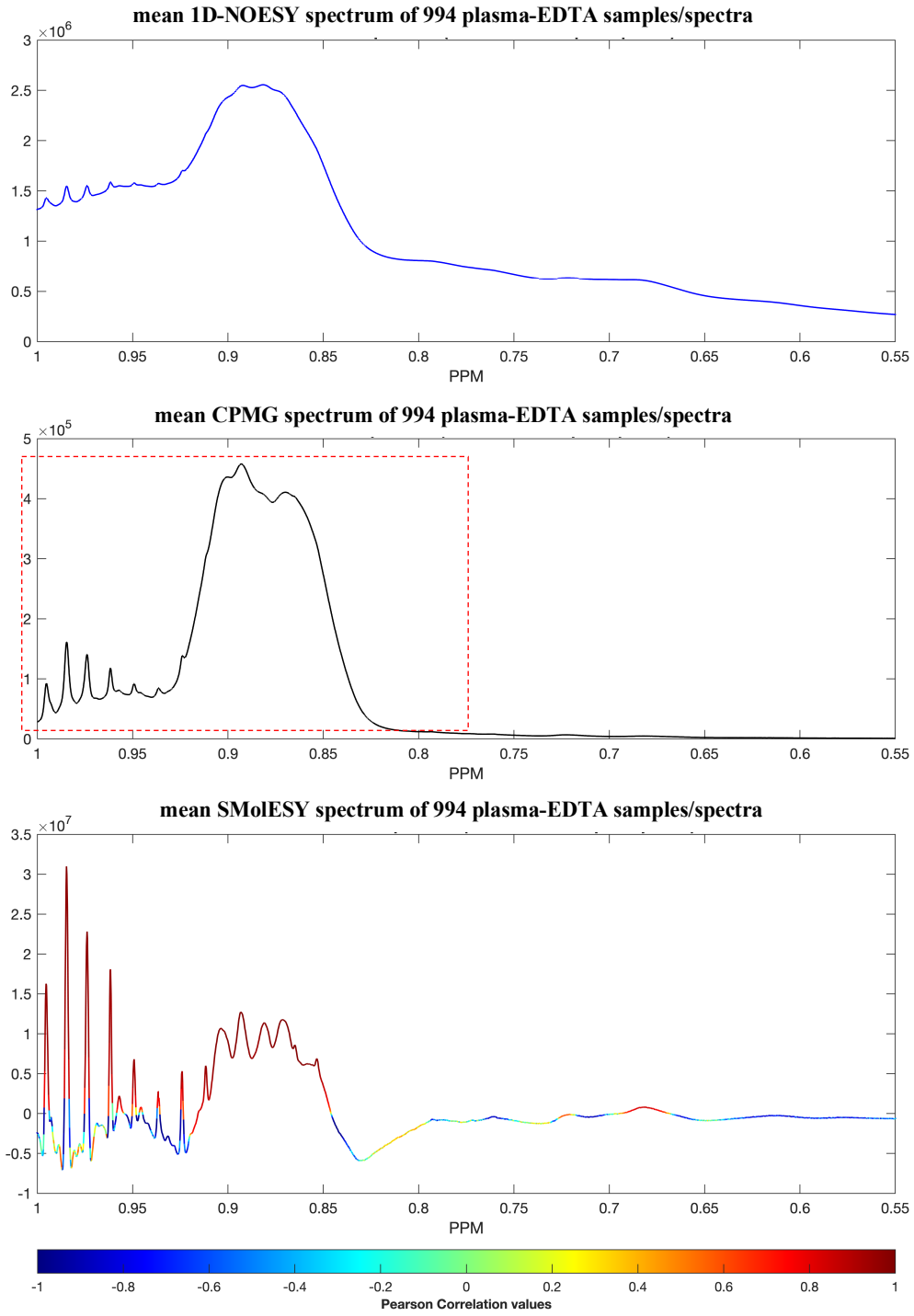
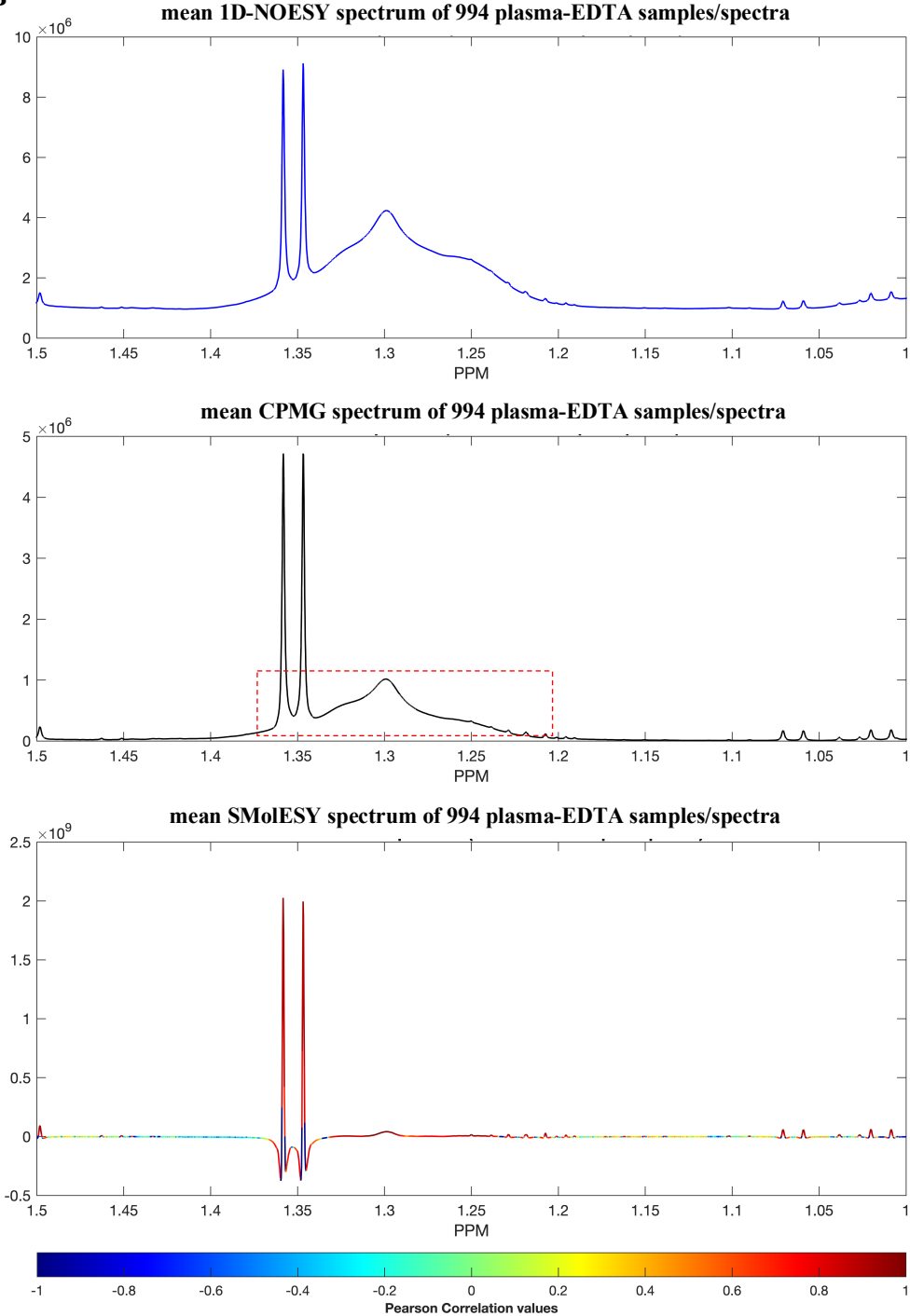


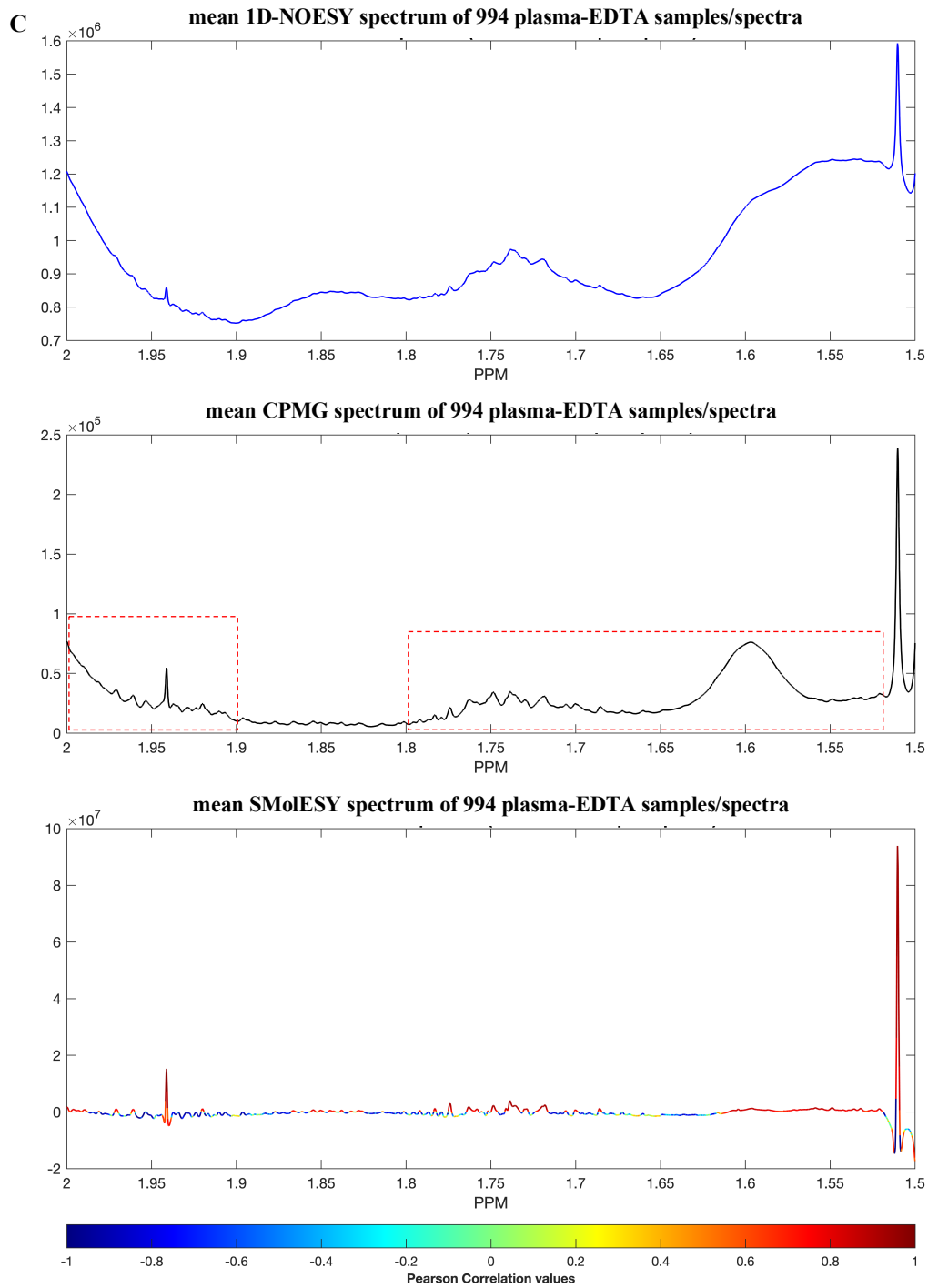
Figure S4. 3D scores plots of the first three components from the PCA of a urine dataset for both 1D-NOESY and SMoESY spectra. From the analysis, it is observed that both NOESY and SMoESY spectra capture similar cumulative variability, 93.4% and 89.4% respectively, and provide similar discrimination of sample groups.

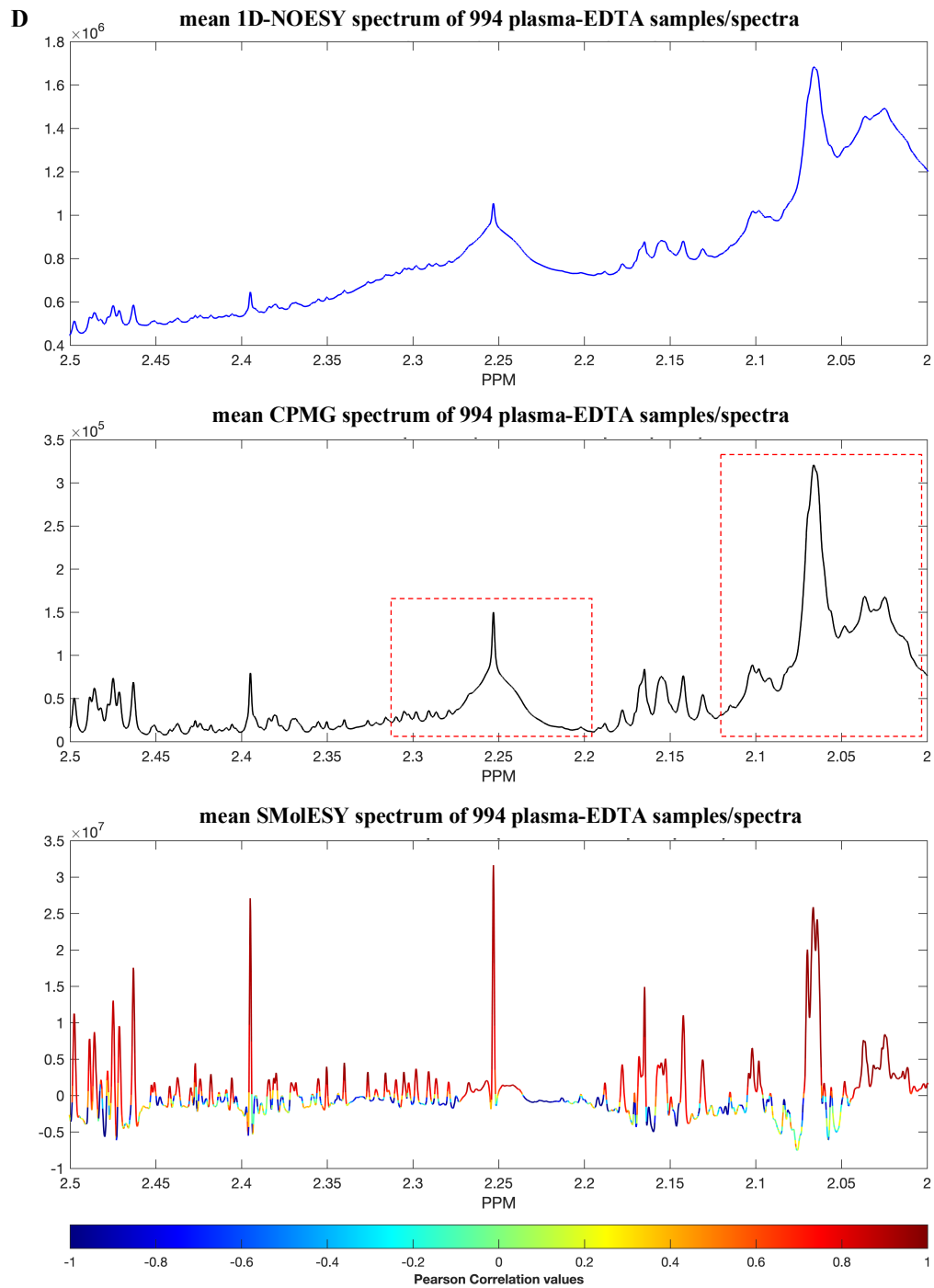
Correlation of SMoESY vs. CPMG spectra for 994 plasma-EDTA samples

A

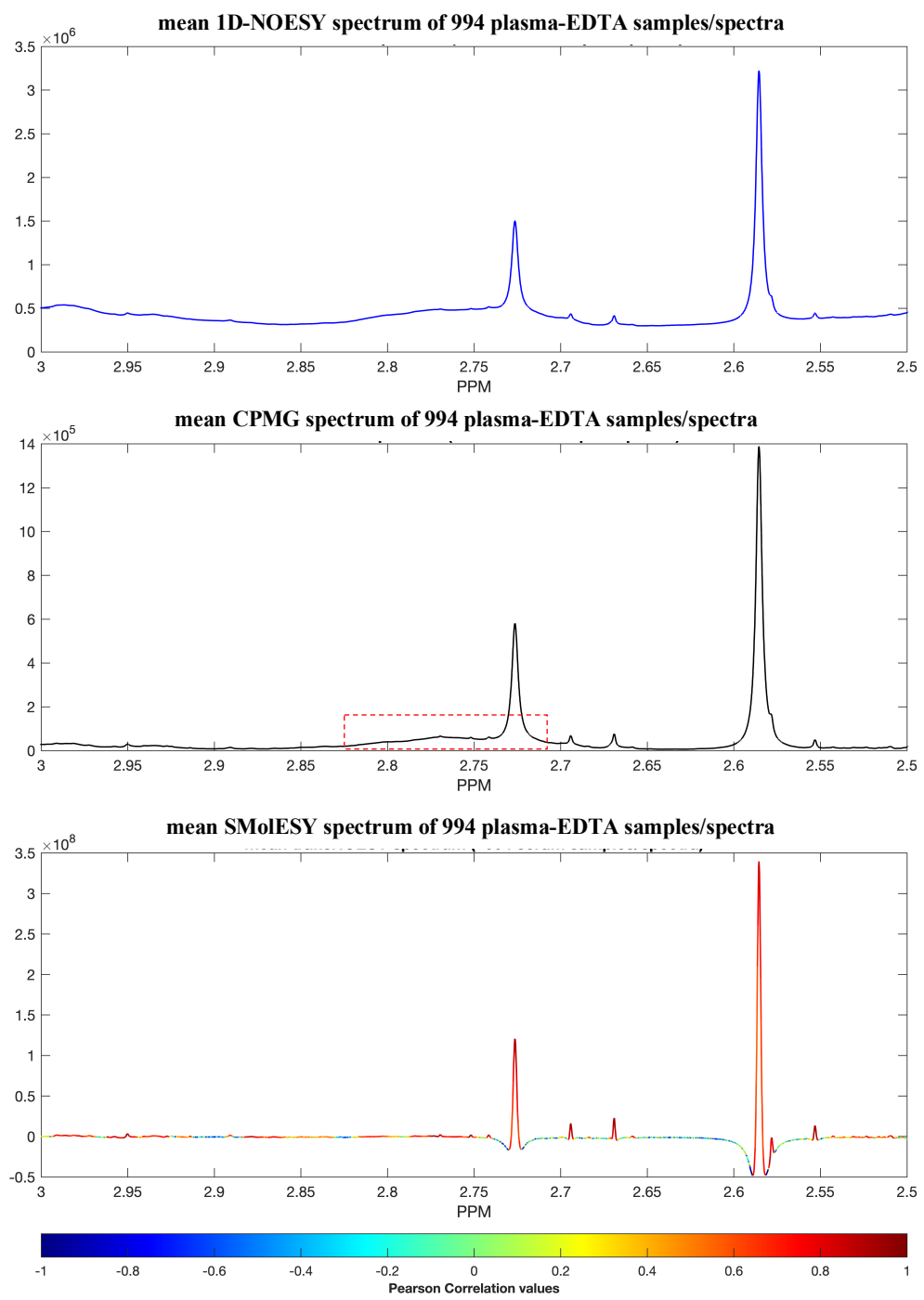


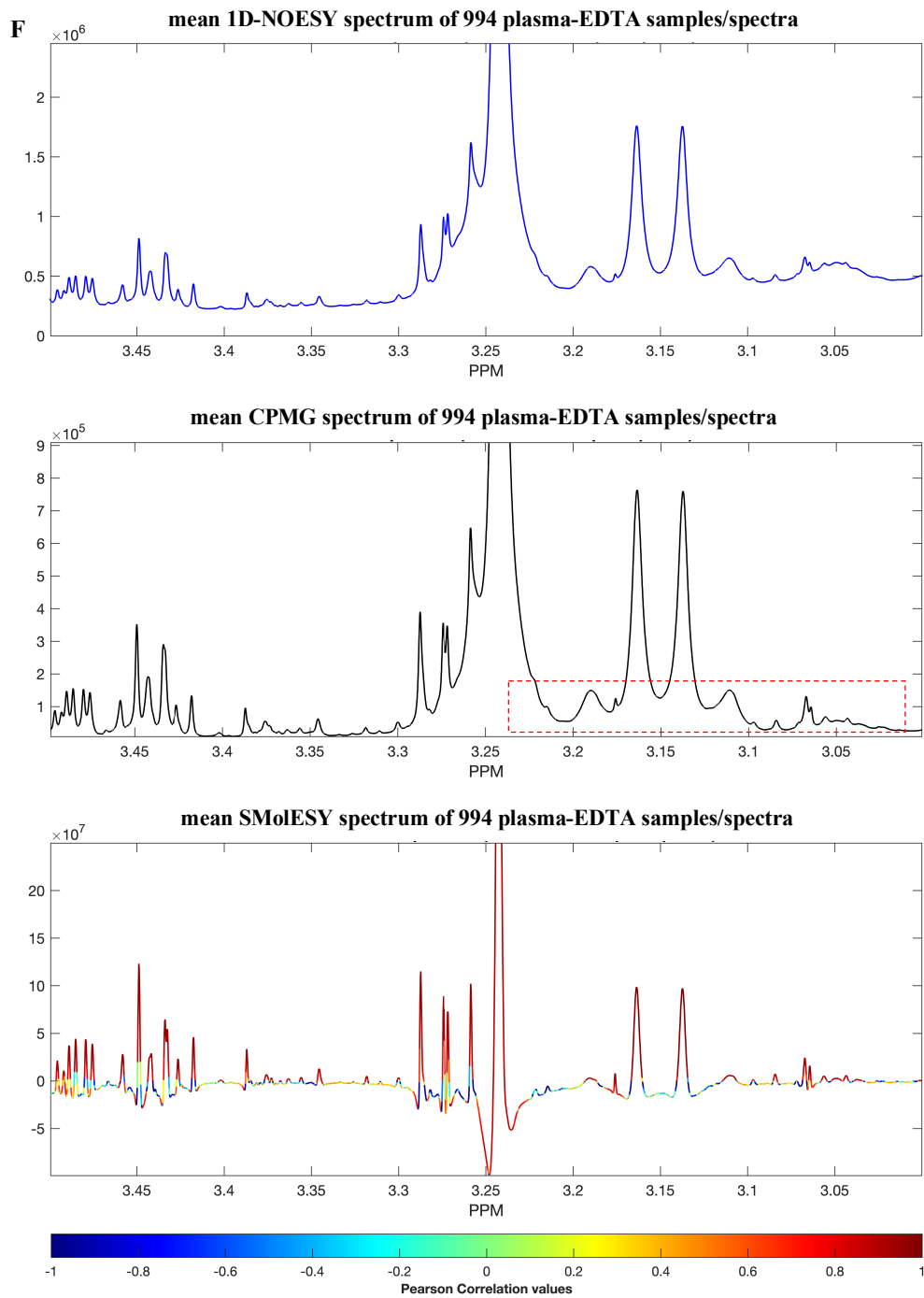
B

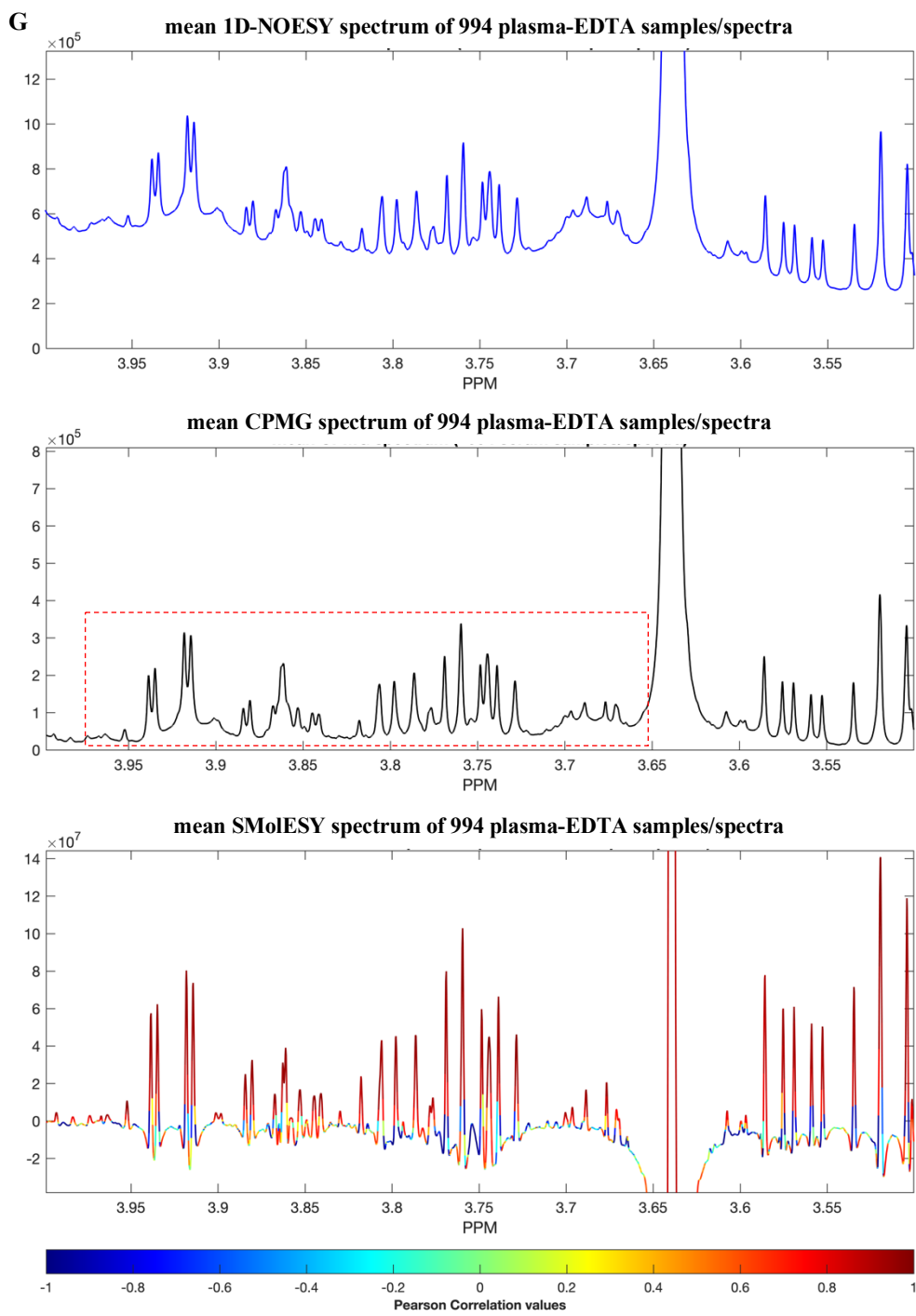




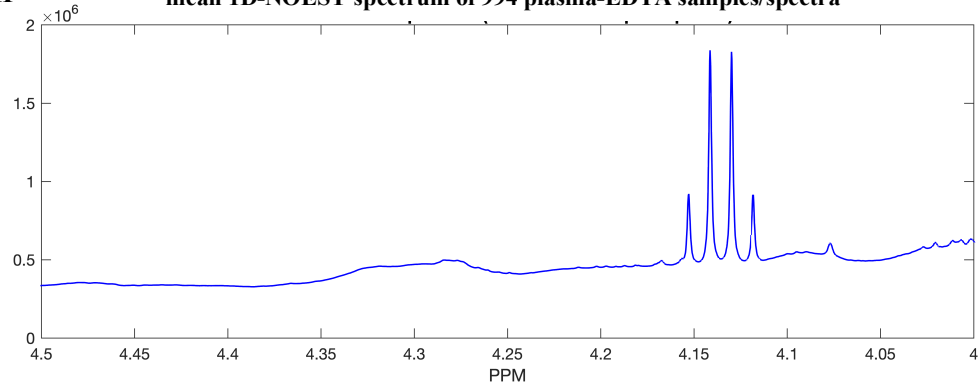
E



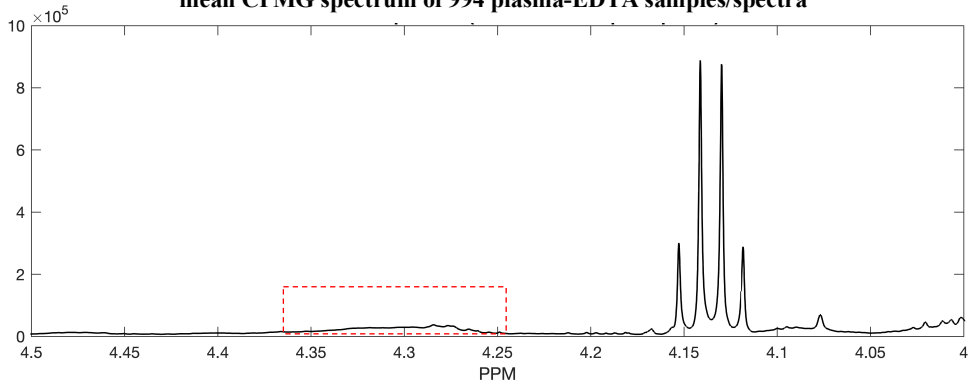




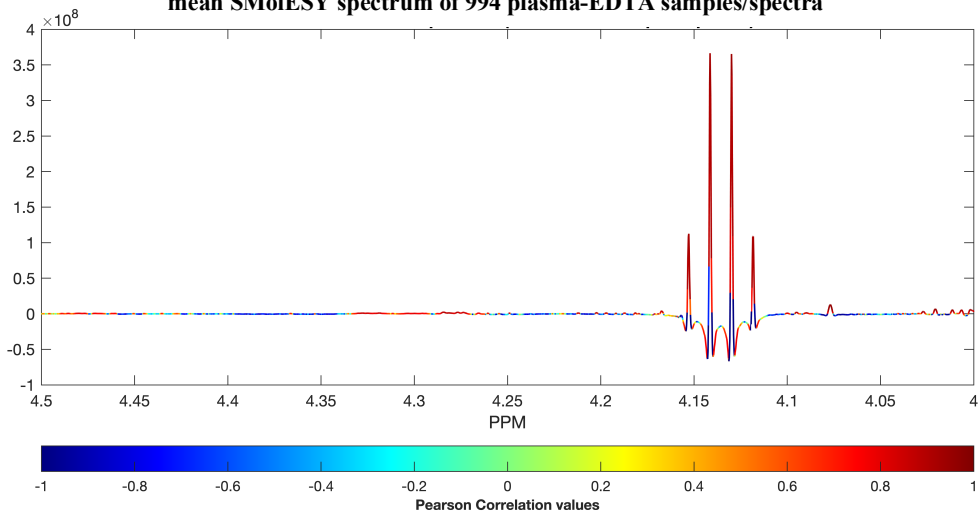
H mean 1D-NOESY spectrum of 994 plasma-EDTA samples/spectra

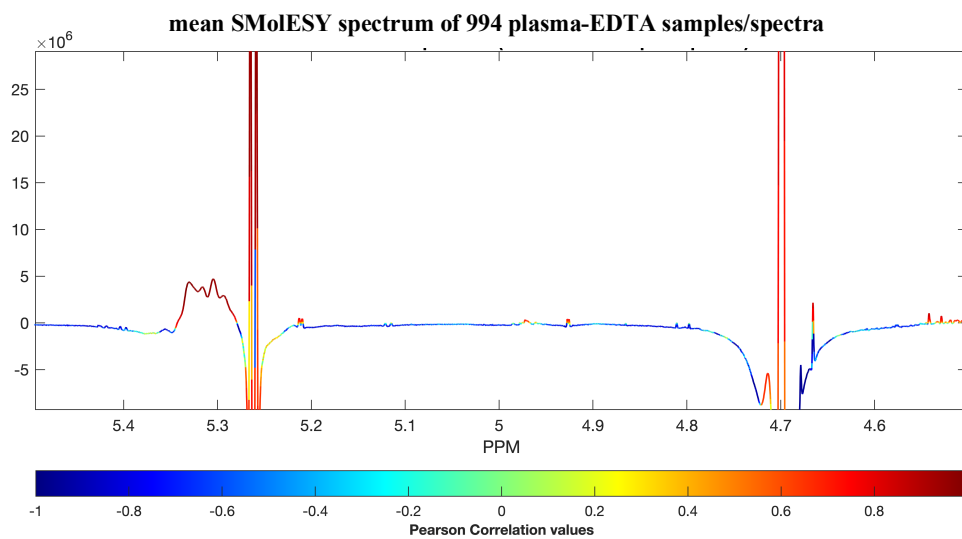
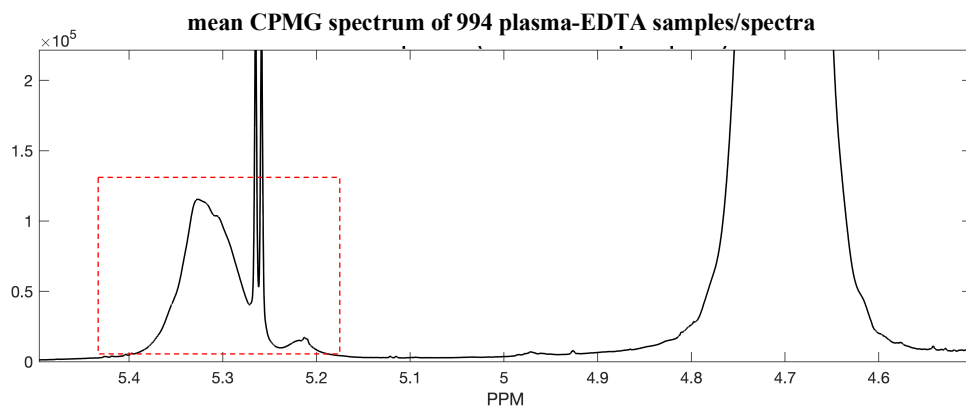
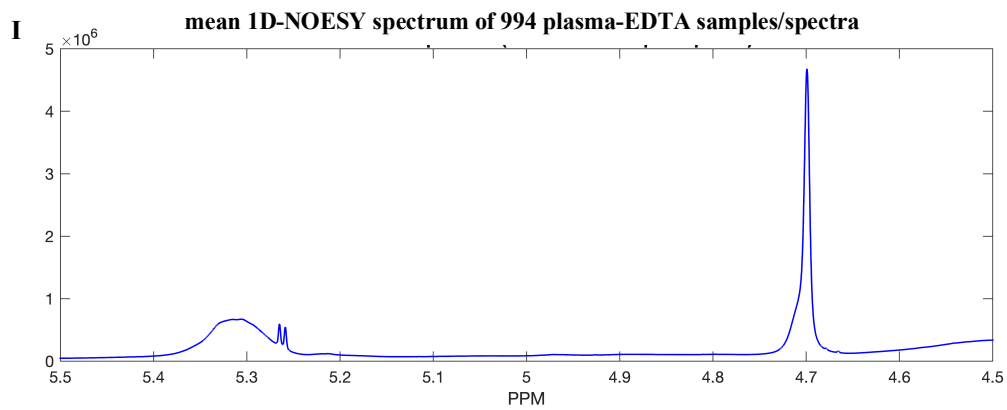


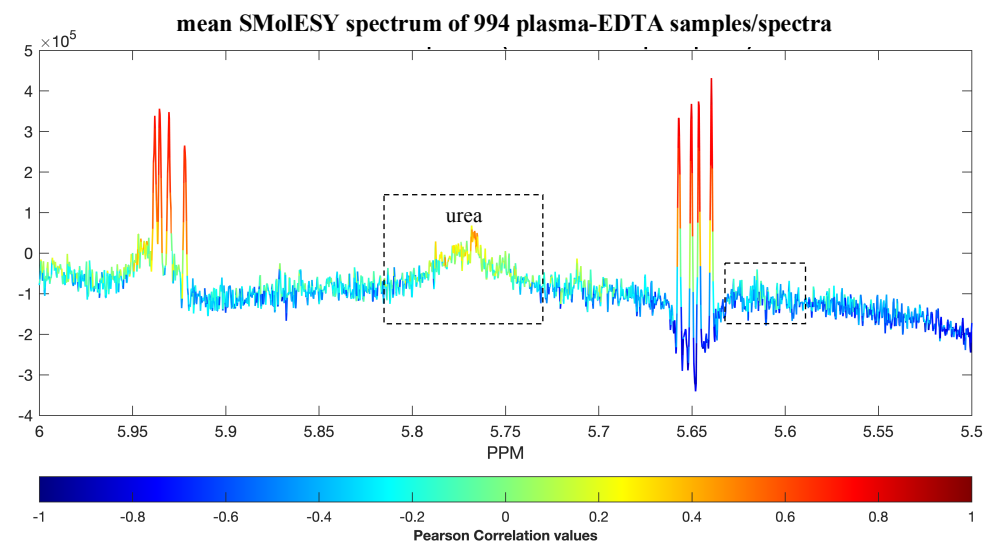
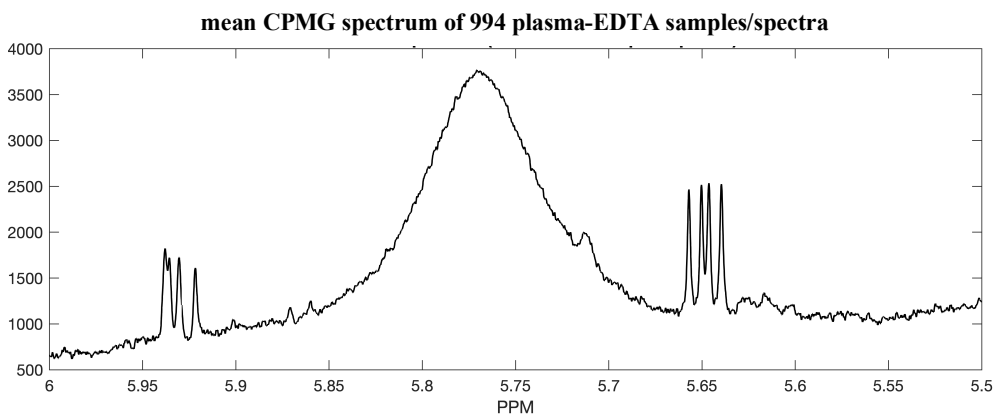
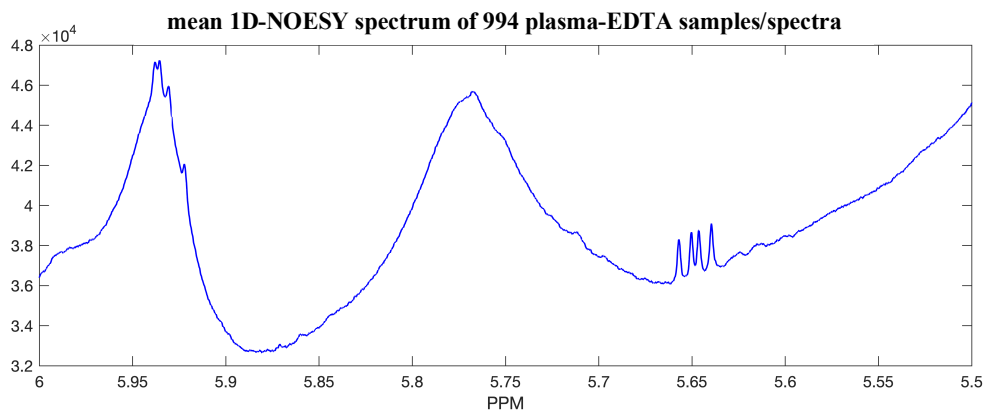
mean CPMG spectrum of 994 plasma-EDTA samples/spectra

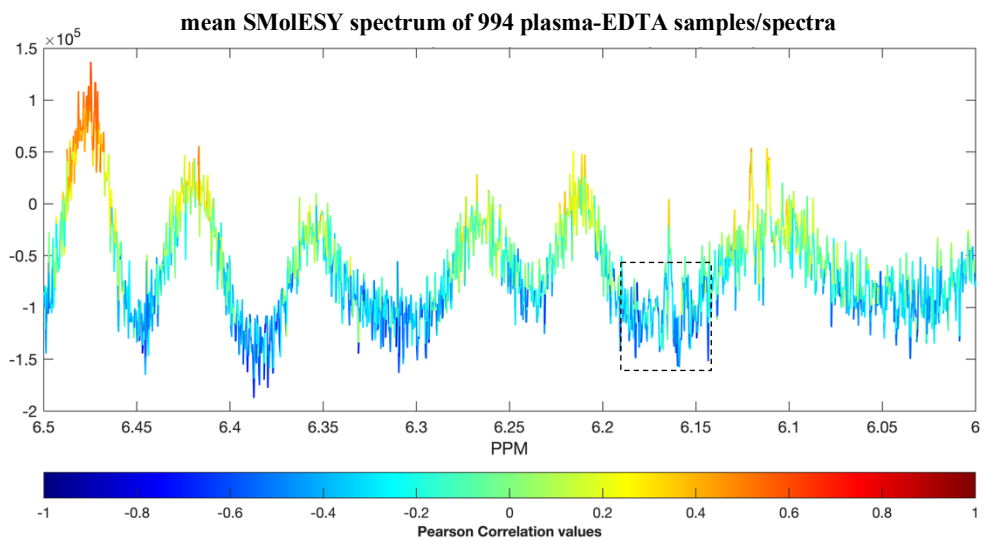
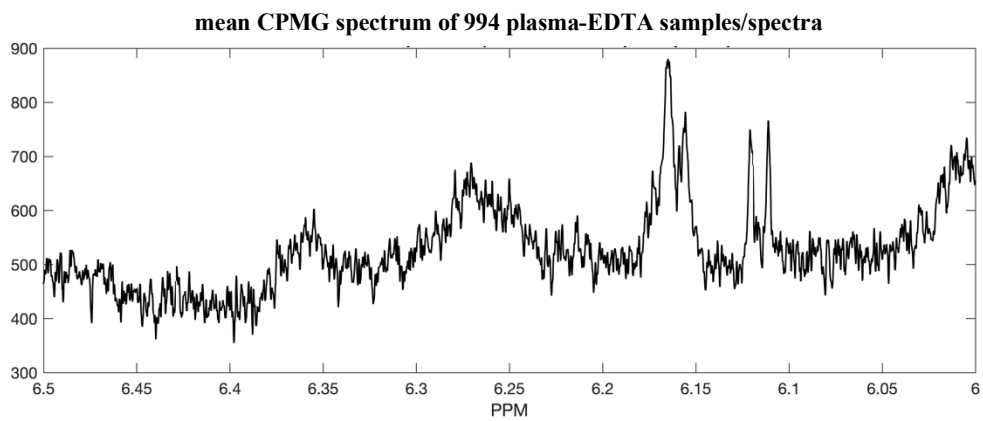
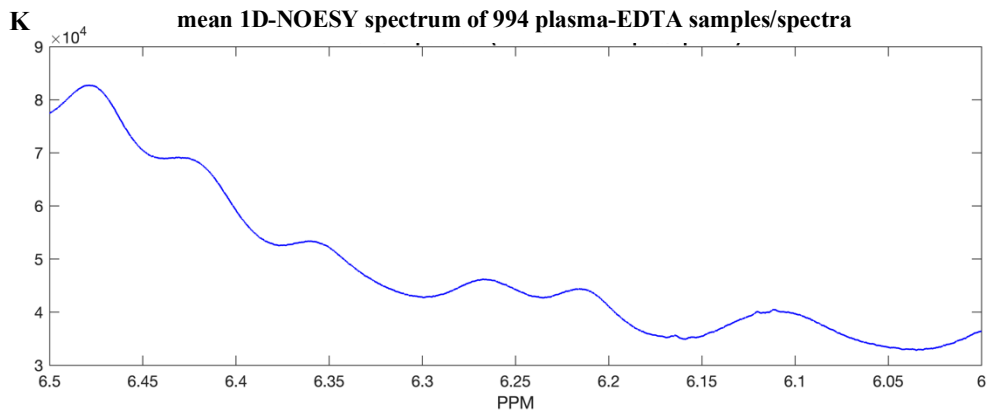


mean SMoESY spectrum of 994 plasma-EDTA samples/spectra

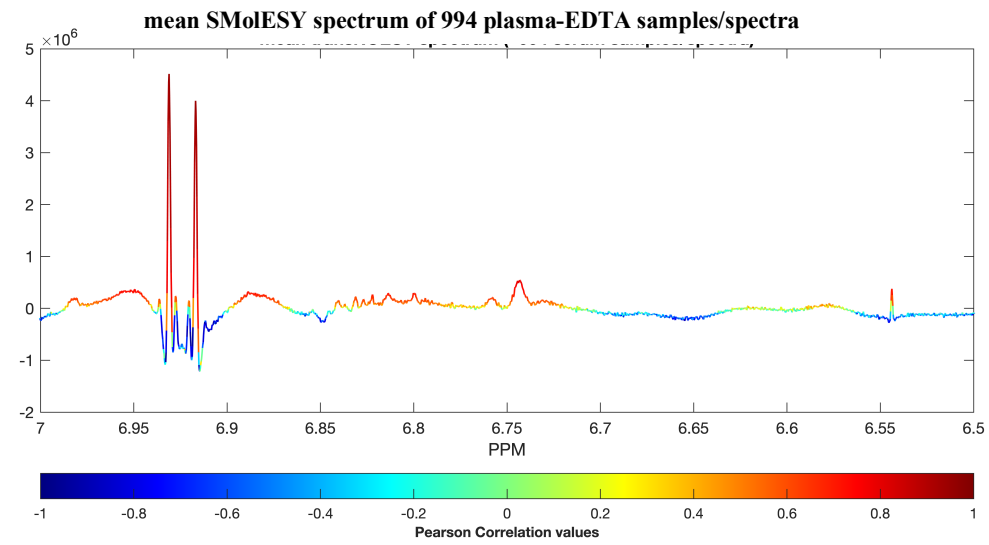
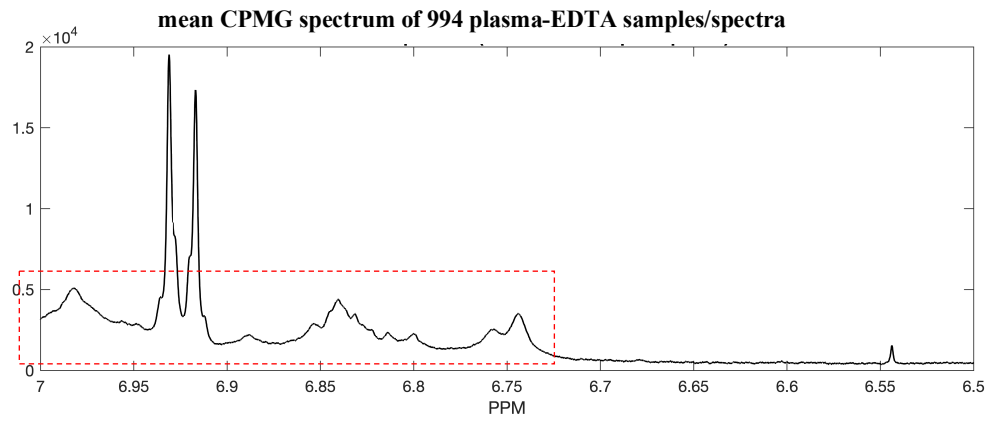
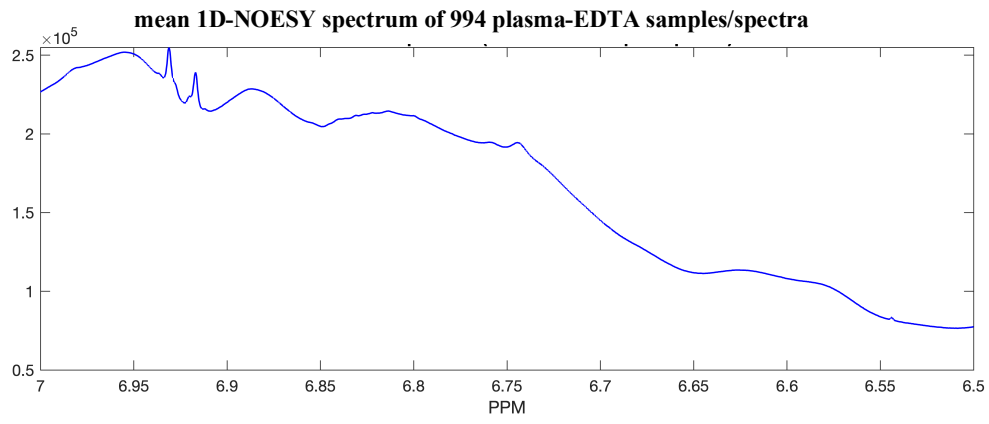




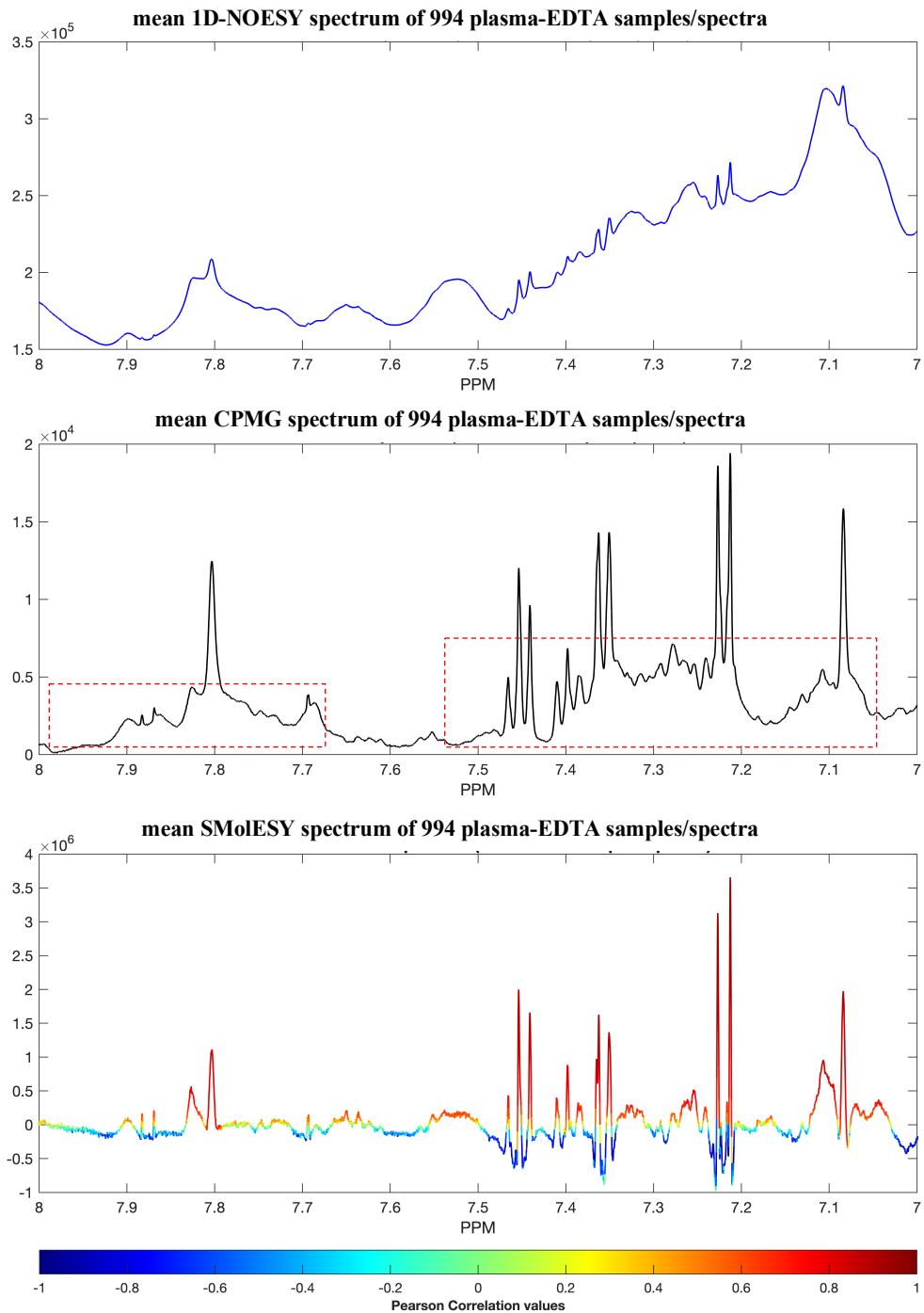
J



L



M



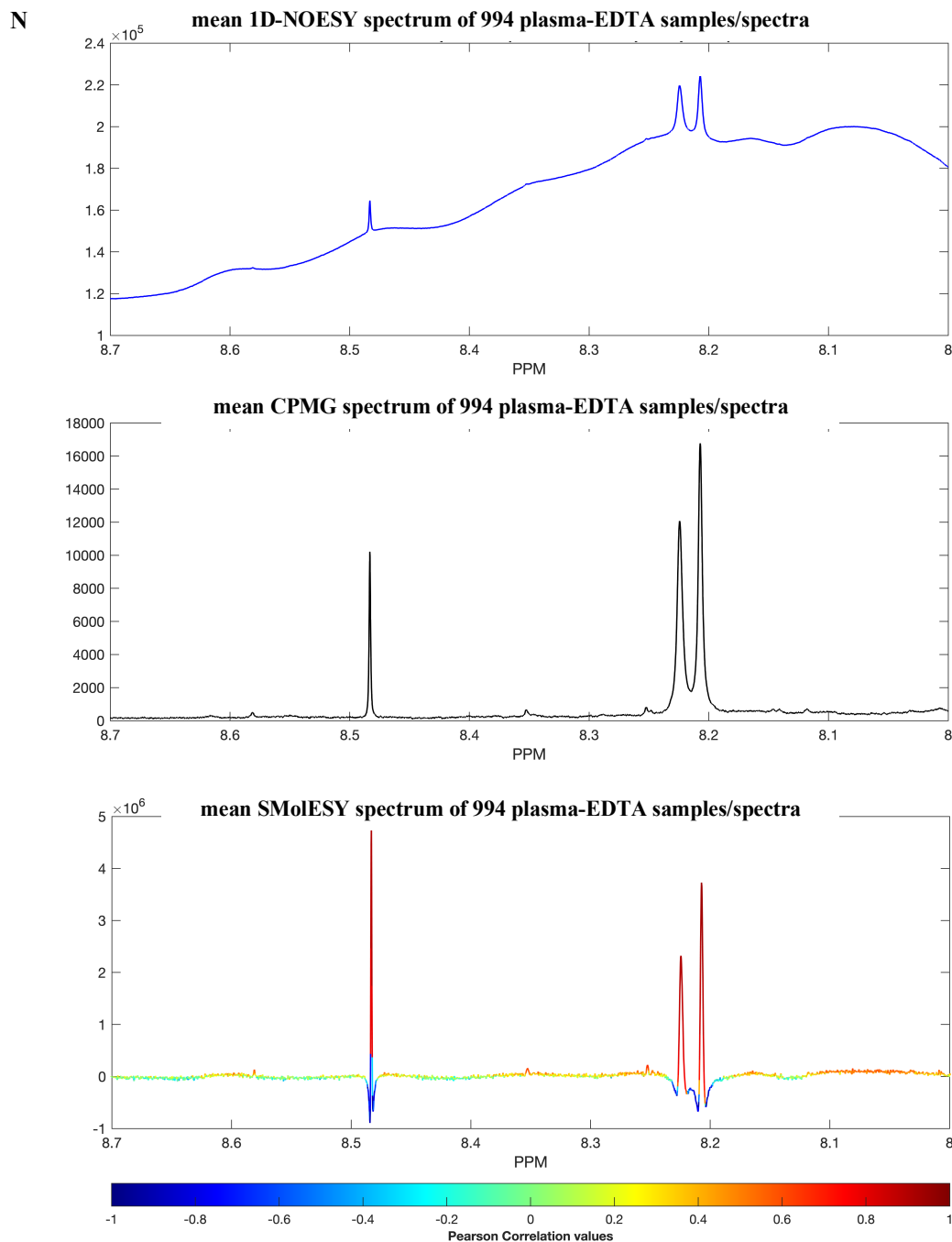


Figure S5. SMoIESY performance in 994 plasma-EDTA samples. (A-N) Mean spectrum of 994 plasma-EDTA 1D-NOESY, CPMG and SMoIESY ¹H-NMR spectra (0.55–8.7 ppm) zoomed at ~0.5-0.7 ppm windows. SMoIESY is colored according to the Pearson coefficients from SMoIESY versus CPMG signals correlation in 994 spectra. The majority of highly resolved SMoIESY signals are linearly correlated to the CPMG and > 99.5 % of CPMG features of low molecular weight metabolites are maintained, while successfully suppressing the broad signals of macromolecules in contrast to CPMG (examples of unsuppressed CPMG broad signals are highlighted by red dashed boxes in panels A-I, L, M). It is noted that broad signal of urea along with 3 broad signals of very low abundance metabolites (< 1.5 times the signal-noise-ratio) are highly suppressed and low correlated to the CPMG (black dashed boxes in panels K, J), even though being recovered by the SMoIESY.

Examples of SMoESY vs. CPMG and NOESY signals

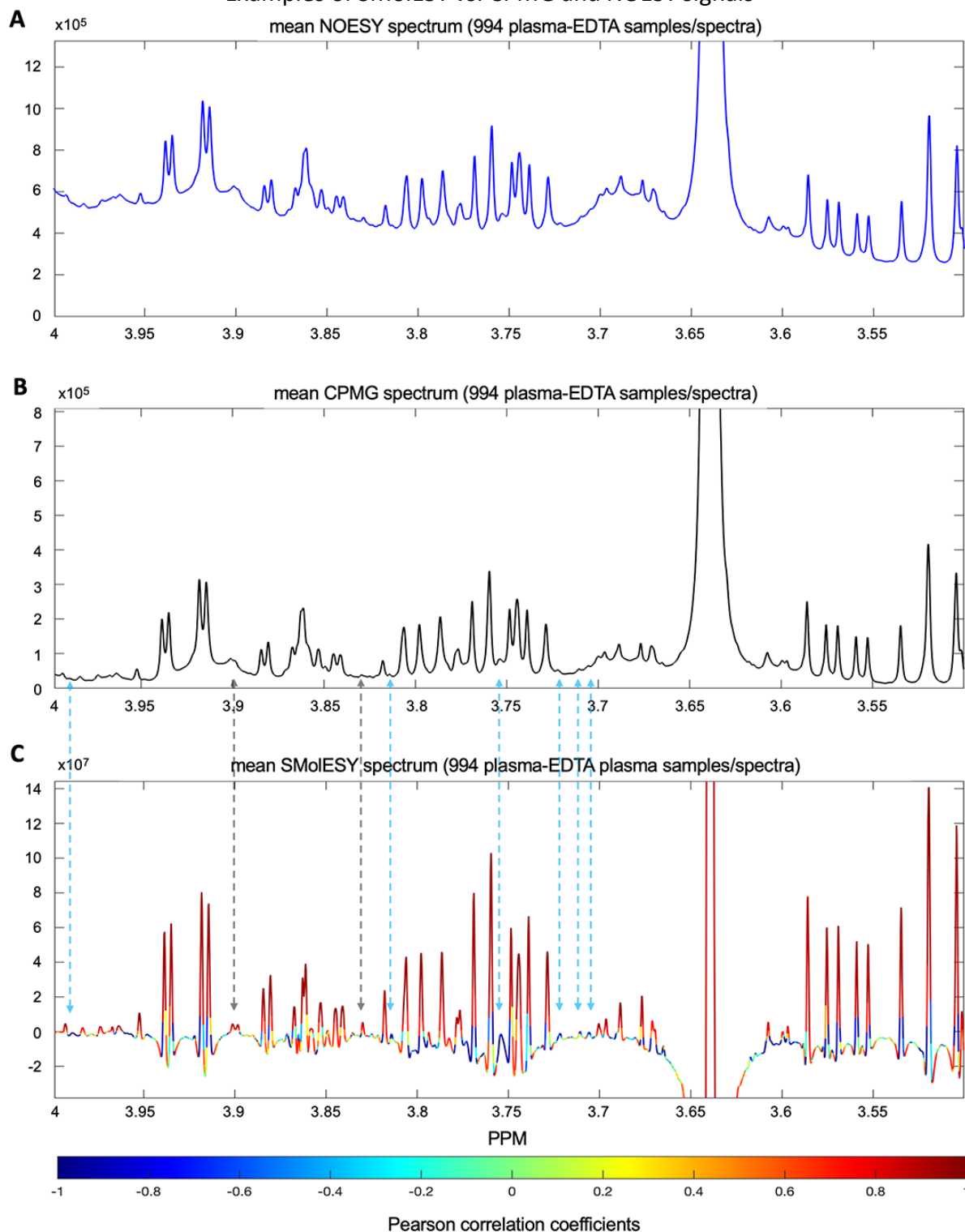
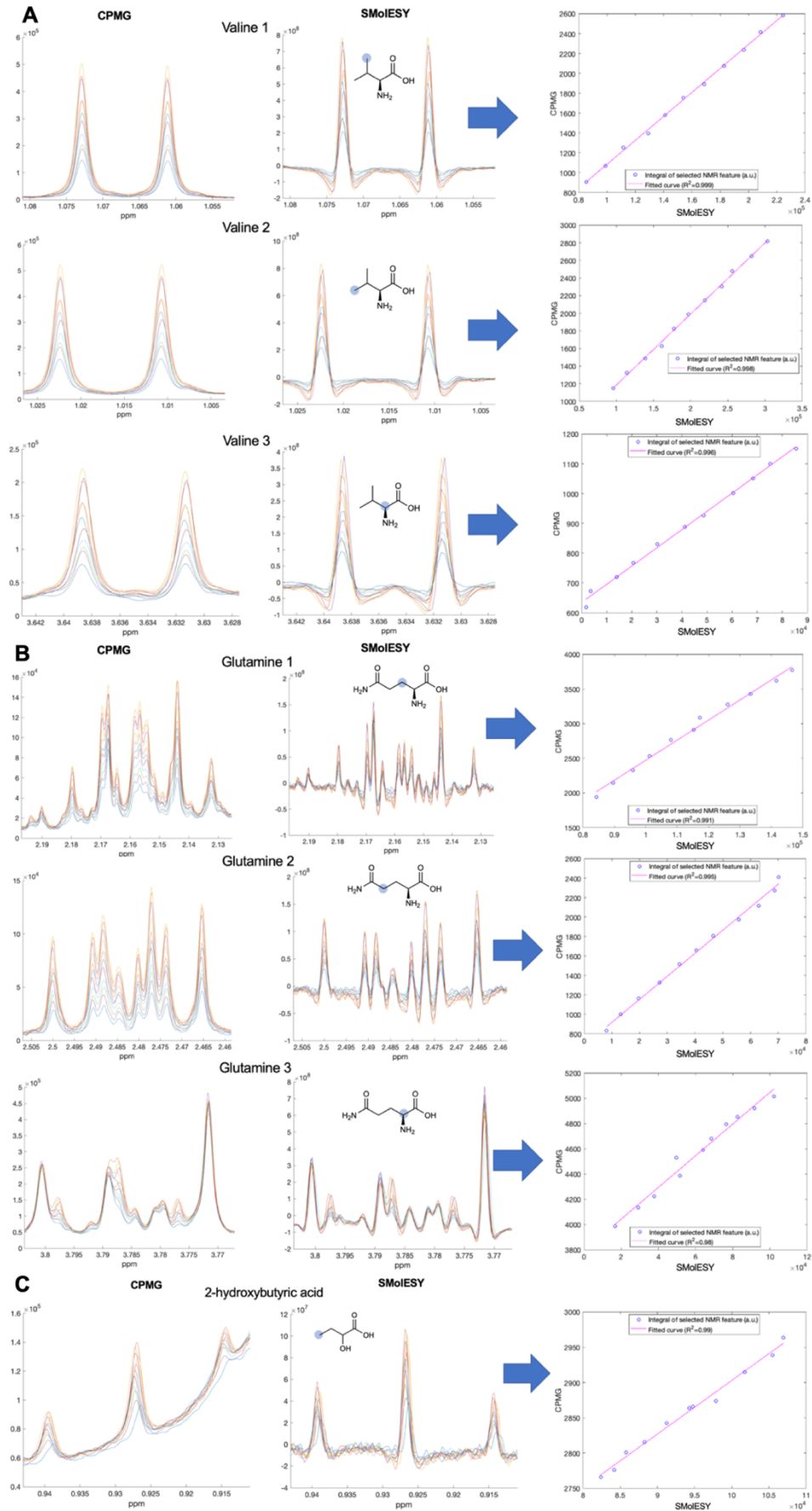
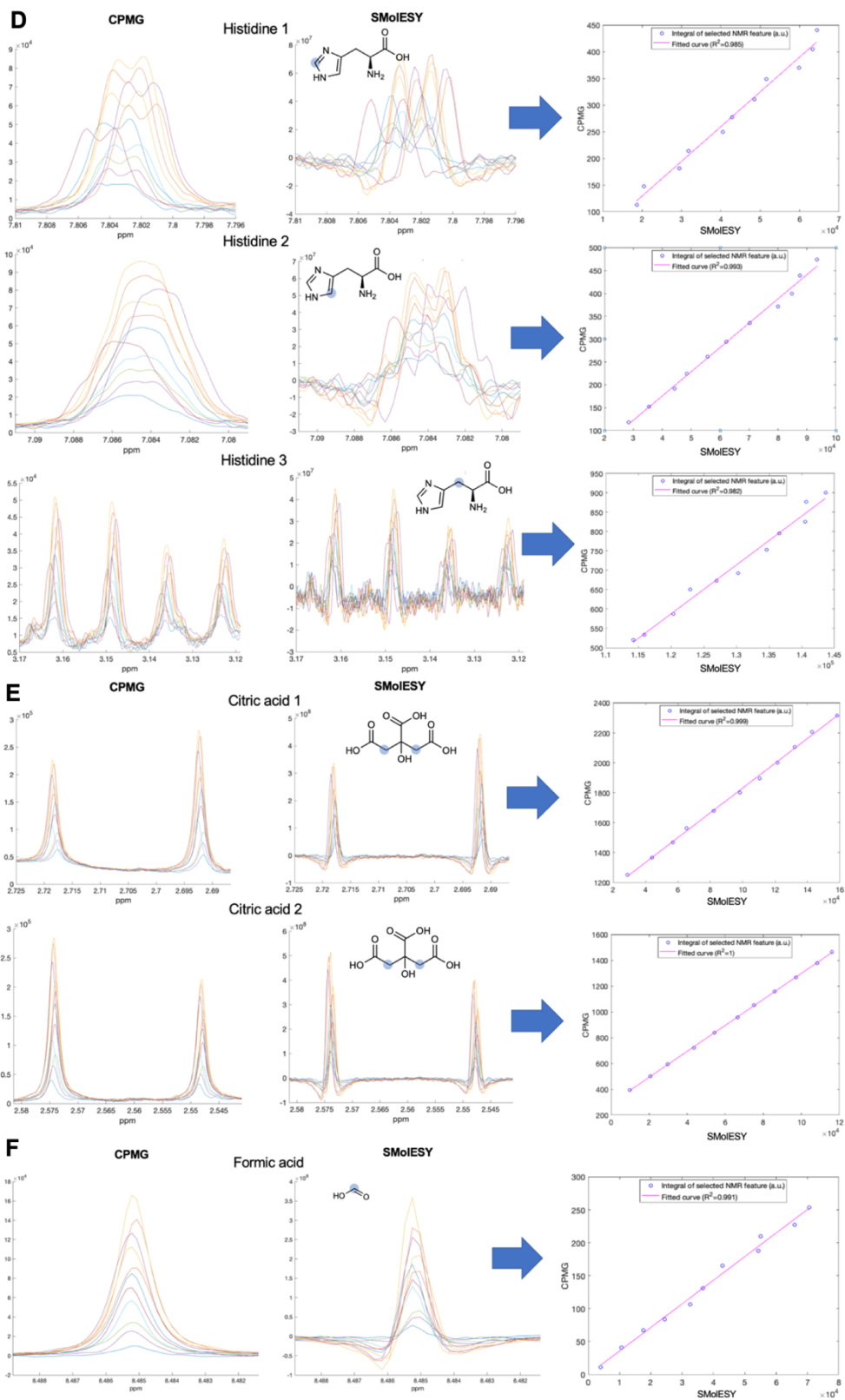
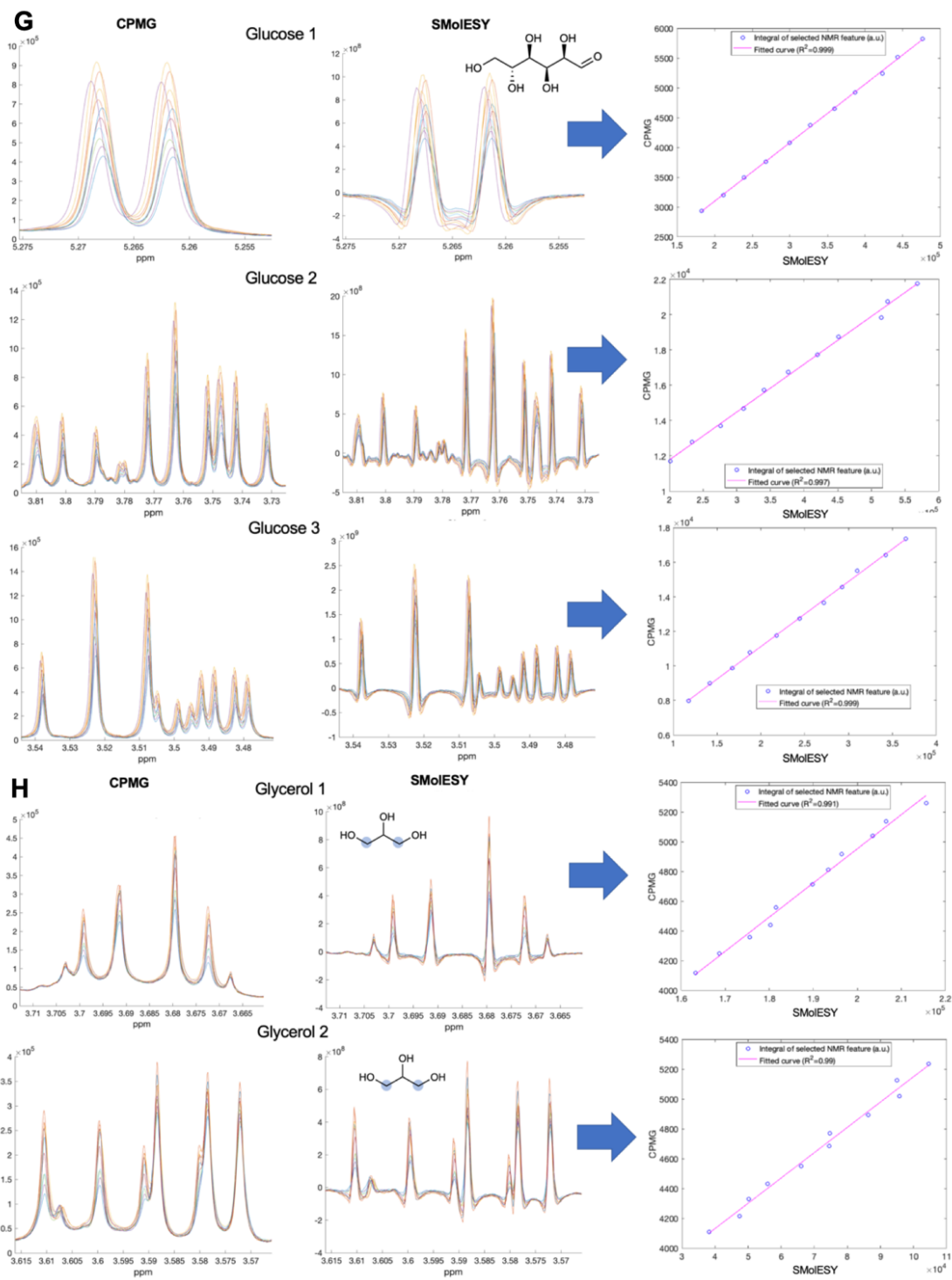


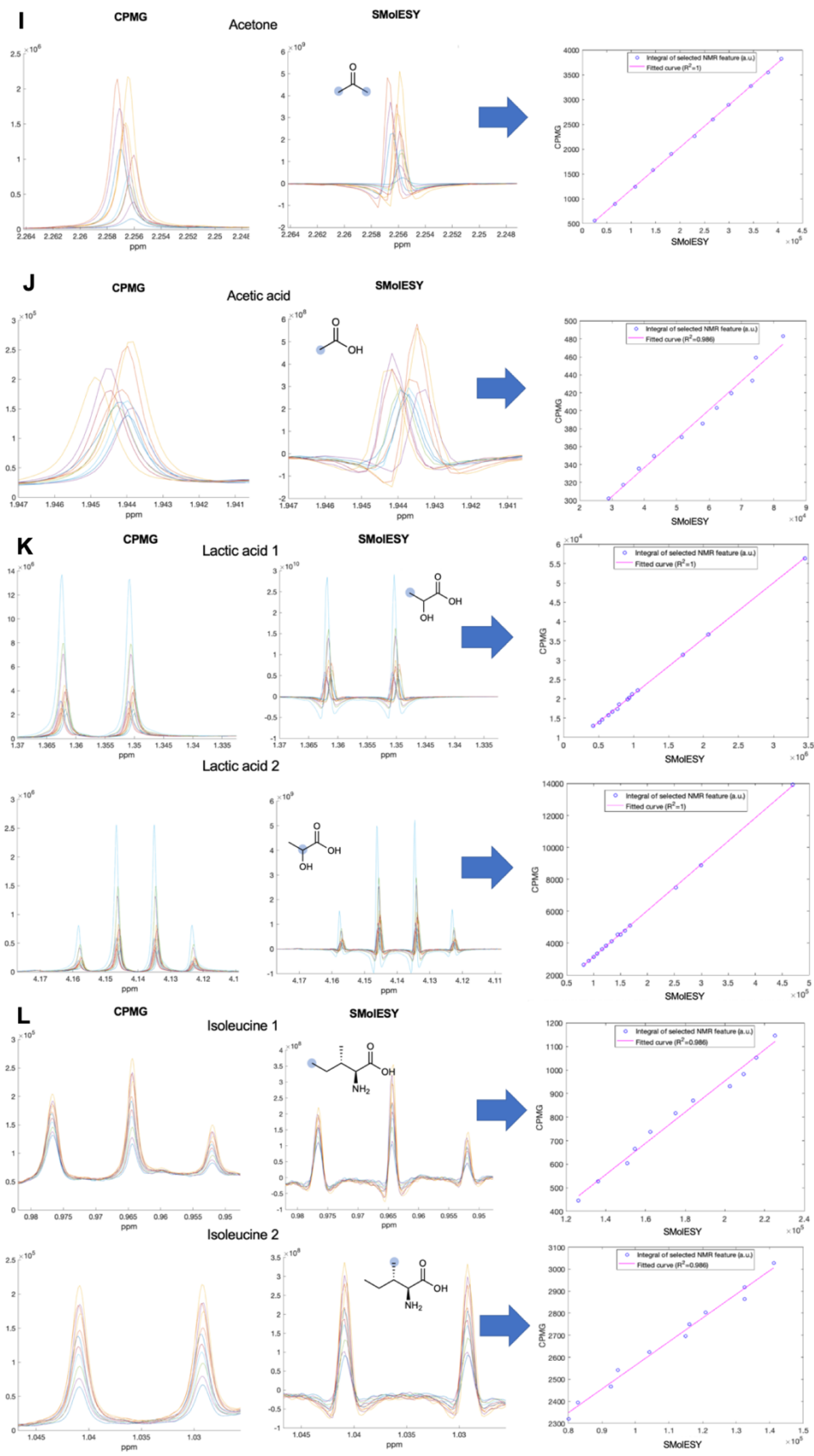
Figure S6. Mean spectrum of 994 plasma-EDTA samples spectra focusing on the 3.5-4.0 ^1H -NMR ppm region. Mean spectrum is calculated by the (A) 1D-NOESY, (B) CPMG and (C) SMoESY 994 spectra. SMoESY is colored according to the Pearson coefficients from SMoESY versus CPMG signals correlation for the 994 spectra. The majority of highly resolved (e.g. pointed by grey arrows) SMoESY signals are linearly correlated to the CPMG. Few signals are zero or anti-correlated to the CPMG spectra (light blue arrows), due to partial overlapping with other signals and/or baseline distortions in the CPMG (i.e. ineffective suppression of broad signals).

SMoESY vs CPMG spectral bins correlation – spiking experiments in real plasma matrix









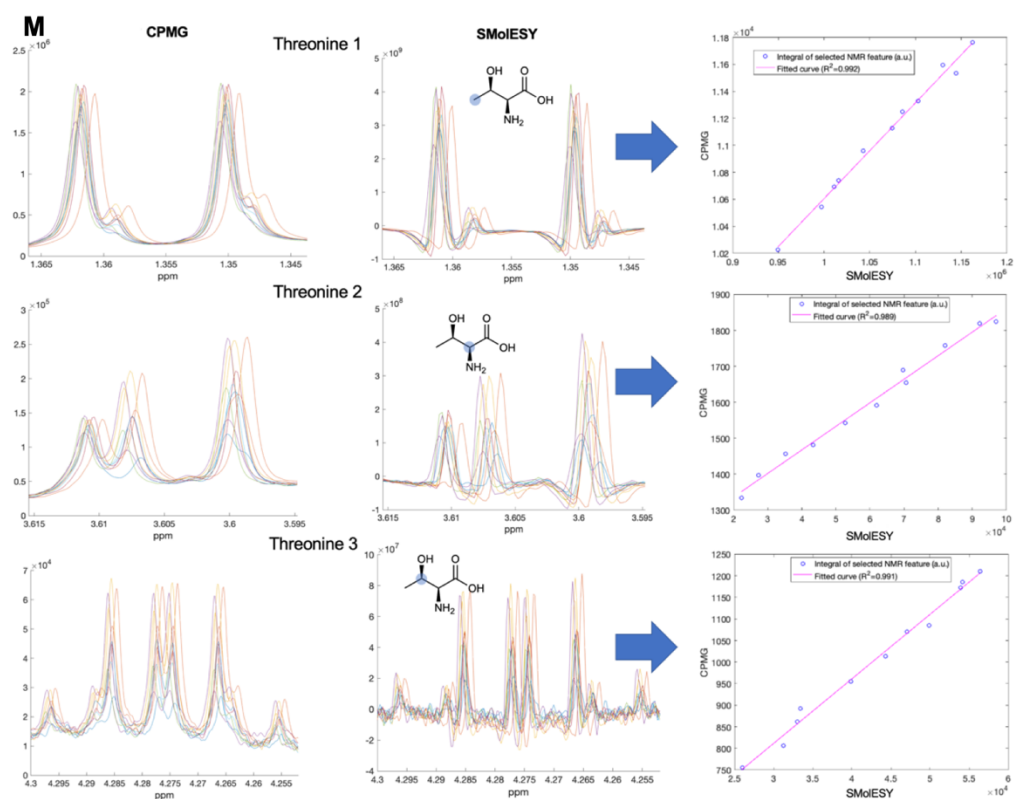


Figure S7. SMoESY application and reproducibility validation to spectra binning. Comparison between CPMG (left panel) and SMoESY (middle panel) spectral bins including various ^1H -NMR signals of 13 spiked metabolites (11 concentrations) in a plasma sample: (A) L-valine, (B) L-glutamine, (C) 2-hydroxybutyric acid, (D) L-histidine, (E) Citric acid, (F) Formic acid, (G) Glucose, (H) Glycerol, (I) Acetone, (J) Acetic acid (K) L-lactic acid, (L) L-isoleucine and (M) L-threonine. Linear regression curves (right panel) exhibit $0.98 < R^2 < 1$ for all metabolites, indicate that SMoESY is highly reproducible while superseding CPMG in broad signals suppression resulting in zero baseline distortions. Due to ~ 0 error in bin integration, horizontal and vertical error bars are not plotted in all linear regression plots. Light blue circles indicate the proton spin system of each metabolite included in each spectral bin which might include signals of other (not spiked) metabolites.

Comparison between SMoIESY and signal deconvolution approaches for absolute quantification

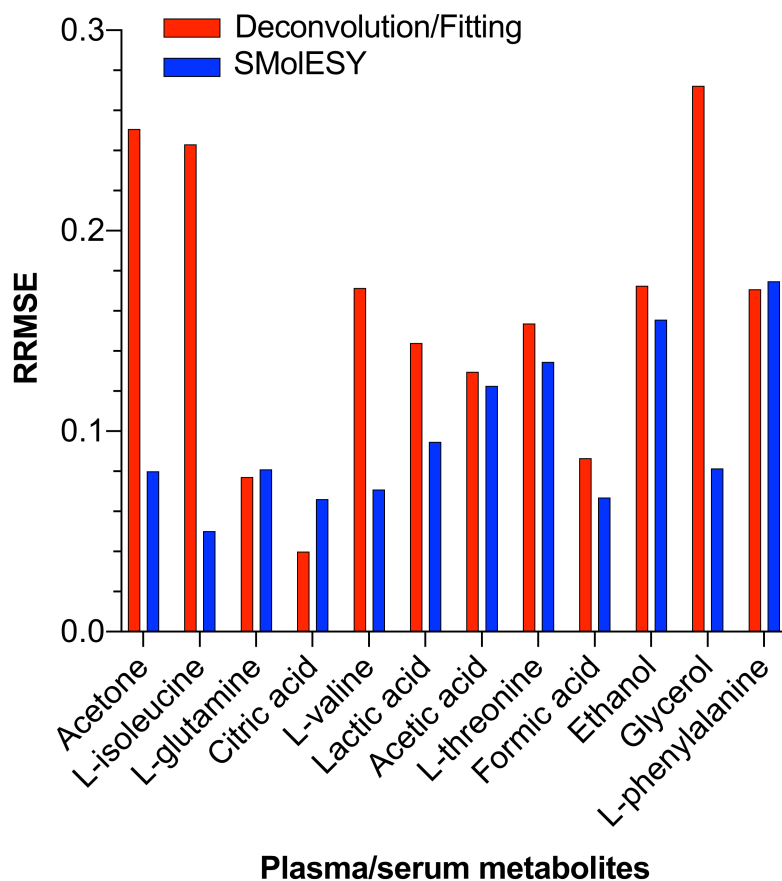


Figure S8. The relative root mean square errors (RRMSE) of SMoIESY and deconvolution algorithms used for 12 spiked plasma metabolites absolute quantification were calculated based upon the regression analyses (see Fig. 5). RRMSE values show that absolute quantification via SMoIESY signals integration provides less errors compared to the fitting procedures followed by deconvolution algorithms.

An overview of the SMoIESY_platform graphical user interface (GUI) toolbox.

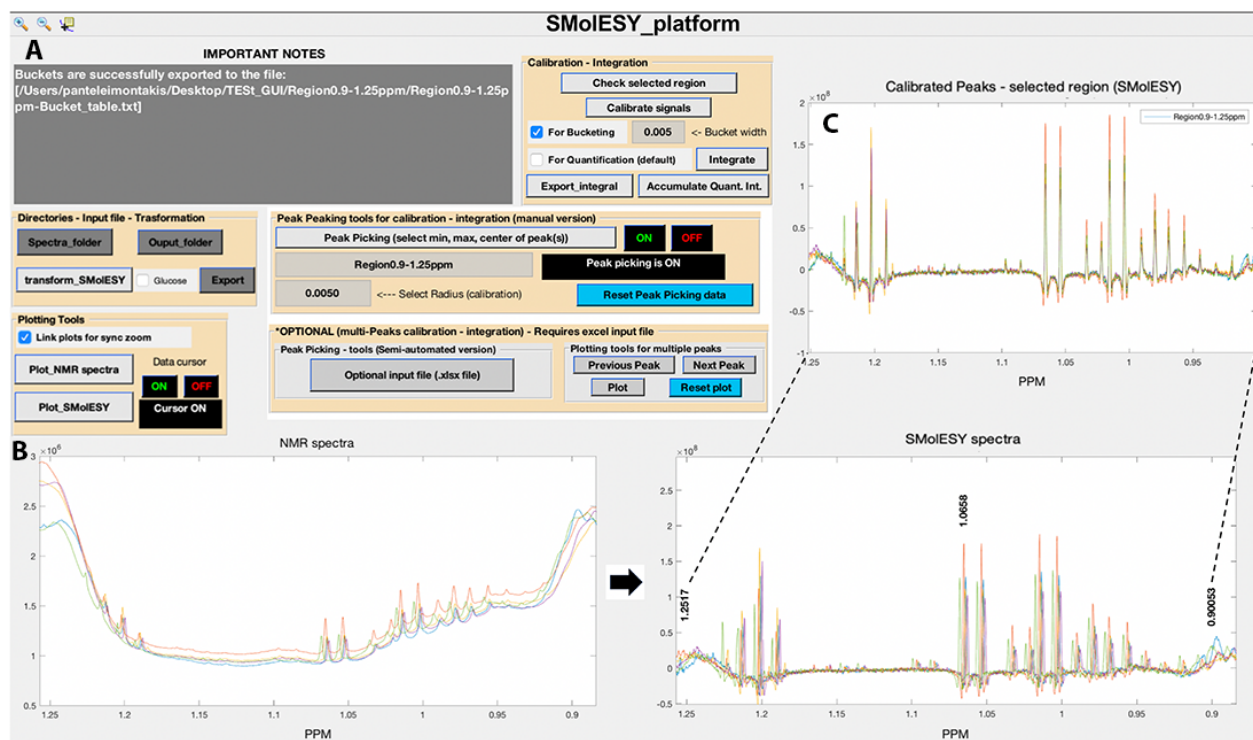


Figure S9. (A) The development of SMoIESY_platform provides the opportunity for any user to load 1D ^1H -NMR raw spectra, so as to transform them into SMoIESY and export them into a .txt file. In addition, the user has the opportunity to calibrate the SMoIESY spectra to the doublet of the anomeric proton of glucose (~ 5.25 ppm) in case of plasma/serum/CSF etc. acquired spectra. (B) It offers the possibility to plot both 1D ^1H -NMR and SMoIESY spectra for a synchronized zoom in both panels, (C) as well as to align into specific reference peaks a set of spectral bins or individual signals so as to integrate SMoIESY features either for qualitative (i.e. option of variable-size SMoIESY spectra binning) or quantitative purposes (i.e. option of SMoIESY signals integration for quantification). The alignment of the signals could be performed both manually or in a semi-automated way upon users experience and request. More details and user guidelines of the software could be found at: https://github.com/pantakis/SMoIESY_platform.

Computer code for the calculation of the Pearson correlation coefficients.

```
function [SMoIESY_ints,CPMG_ints,Corrv,Pv] = pearson_cor_SMoIESYvsCPMG(ppm_SMoIESY,ppm_CPMG,SMoIESY_data,CPMG_data)

%
% Calculating the integrals of each feature/point of SMoIESY for a
% comparison to CPMG equal width of ppm.
% NOTE: SMoIESY - CPMG spectral data should be aligned to the same
% reference peak (e.g. glucose, TSP etc.)
%
% Inputs
% ppm_SMoIESY: PPM data of all SMoIESY data (one vector)
% ppm_CPMG: PPM data of all CPMG data (one vector)
% SMoIESY_data: SMoIESY data (intensities) (matrix, rows: spectra, columns:
% intensities)
% CPMG_data: CPMG data (intensities) (matrix, rows: spectra, columns:
% intensities)
%
% Outputs
% SMoIESY_ints: Integral of each SMoIESY feature
% CPMG_ints: Integral of each CPMG feature (describing the same ppm region
% as the corresponding SMoIESY feature)
% Corrv: calculated Pearson's Linear Correlation Coefficient
%
%
%
num_of_spectra = size(SMoIESY_data,1);
XAXIS_CPMG_rounded = round(ppm_CPMG,4);
XAXIS_SMoIESY_rounded = round(ppm_SMoIESY,4);

% equal Xaxis buckets for both CPMG and SMoIESY/NOESY
for l = 2:length(ppm_SMoIESY)
    [~,ii] = find(ppm_CPMG <= ppm_SMoIESY(:,l-1) & ppm_CPMG >= ppm_SMoIESY(:,l));
    KL(l-1).aa = ii;
    clearvars ii
    for i = 1:num_of_spectra
        SMoIESY_ints(i,l-1) = trapz(fliplr(XAXIS_SMoIESY_rounded(1,l-1:1)),fliplr(SMoIESY_data(i,l-1:1)));
        if length(KL(l-1).aa) > 1
            CPMG_ints(i,l-1) = trapz(fliplr(XAXIS_CPMG_rounded(1,min(KL(l-1).aa):max(KL(l-1).aa))),fliplr(CPMG_data(i,min(KL(l-1).aa):max(KL(l-1).aa))));
        else
            CPMG_ints(i,l-1) = CPMG_data(i,KL(l-1).aa);
        end
    end
end
[rho,pval] = corr(CPMG_ints(:,l-1),SMoIESY_ints(:,l-1),'Type','Pearson');
Corrv(l,l-1) = rho;
Pv(l,l-1) = pval;
end
```

Data S1 The described MATLAB code was followed to correlate each SMoIESY feature intensity (i.e. integral) versus its corresponding CPMG feature for more than 3000 plasma spectra from different individuals as depicted in Fig. 2 and Fig. S5.

17 spiked metabolites 11 different concentrations spiked in real plasma sample

Table S2

| Metabolites | Concentration (mM) # | | | | | | | | | | |
|----------------------------------|----------------------|-------|-------|-------|-------|-------|-------|-------|-------|-------|-------|
| | 0 | 0.333 | 0.666 | 0.999 | 1.332 | 1.665 | 1.998 | 2.331 | 2.664 | 2.997 | 3.330 |
| Lactic acid | 0 | 0.333 | 0.666 | 0.999 | 1.332 | 1.665 | 1.998 | 2.331 | 2.664 | 2.997 | 3.330 |
| 2-hydroxybutric acid sodium salt | 0 | 0.016 | 0.032 | 0.048 | 0.063 | 0.079 | 0.095 | 0.111 | 0.127 | 0.143 | 0.159 |
| Acetone | 0 | 0.034 | 0.069 | 0.103 | 0.138 | 0.172 | 0.207 | 0.241 | 0.275 | 0.310 | 0.344 |
| Citric acid | 0 | 0.078 | 0.156 | 0.234 | 0.312 | 0.390 | 0.469 | 0.547 | 0.625 | 0.703 | 0.781 |
| D-glucose | 0 | 0.611 | 1.221 | 1.832 | 2.442 | 3.053 | 3.663 | 4.274 | 4.885 | 5.495 | 6.106 |
| Ethanol | 0 | 0.043 | 0.087 | 0.130 | 0.174 | 0.217 | 0.261 | 0.304 | 0.347 | 0.391 | 0.434 |
| Glycerol | 0 | 0.054 | 0.109 | 0.163 | 0.217 | 0.271 | 0.326 | 0.380 | 0.434 | 0.489 | 0.543 |
| L-aspartic acid | 0 | 0.038 | 0.075 | 0.113 | 0.150 | 0.188 | 0.225 | 0.263 | 0.300 | 0.338 | 0.376 |
| L-glutamine | 0 | 0.082 | 0.164 | 0.246 | 0.328 | 0.411 | 0.493 | 0.575 | 0.657 | 0.739 | 0.821 |
| L-histidine | 0 | 0.045 | 0.090 | 0.135 | 0.181 | 0.226 | 0.271 | 0.316 | 0.361 | 0.406 | 0.451 |
| L-isoleucine | 0 | 0.023 | 0.046 | 0.069 | 0.091 | 0.114 | 0.137 | 0.160 | 0.183 | 0.206 | 0.229 |
| L-phenylalanine | 0 | 0.049 | 0.097 | 0.145 | 0.194 | 0.242 | 0.291 | 0.339 | 0.387 | 0.436 | 0.484 |
| L-threonine | 0 | 0.025 | 0.050 | 0.076 | 0.101 | 0.126 | 0.151 | 0.176 | 0.201 | 0.227 | 0.252 |
| L-tryptophan | 0 | 0.073 | 0.147 | 0.220 | 0.294 | 0.367 | 0.441 | 0.514 | 0.588 | 0.661 | 0.735 |
| L-valine | 0 | 0.085 | 0.170 | 0.256 | 0.341 | 0.427 | 0.512 | 0.598 | 0.683 | 0.768 | 0.854 |
| Sodium acetate | 0 | 0.024 | 0.049 | 0.073 | 0.098 | 0.122 | 0.146 | 0.171 | 0.195 | 0.219 | 0.244 |
| Sodium formate | 0 | 0.044 | 0.088 | 0.132 | 0.176 | 0.221 | 0.265 | 0.309 | 0.353 | 0.397 | 0.441 |

#Zero values correspond to the non-spiked plasma sample.

Example of a smoothing filter for denoising SMoIESY

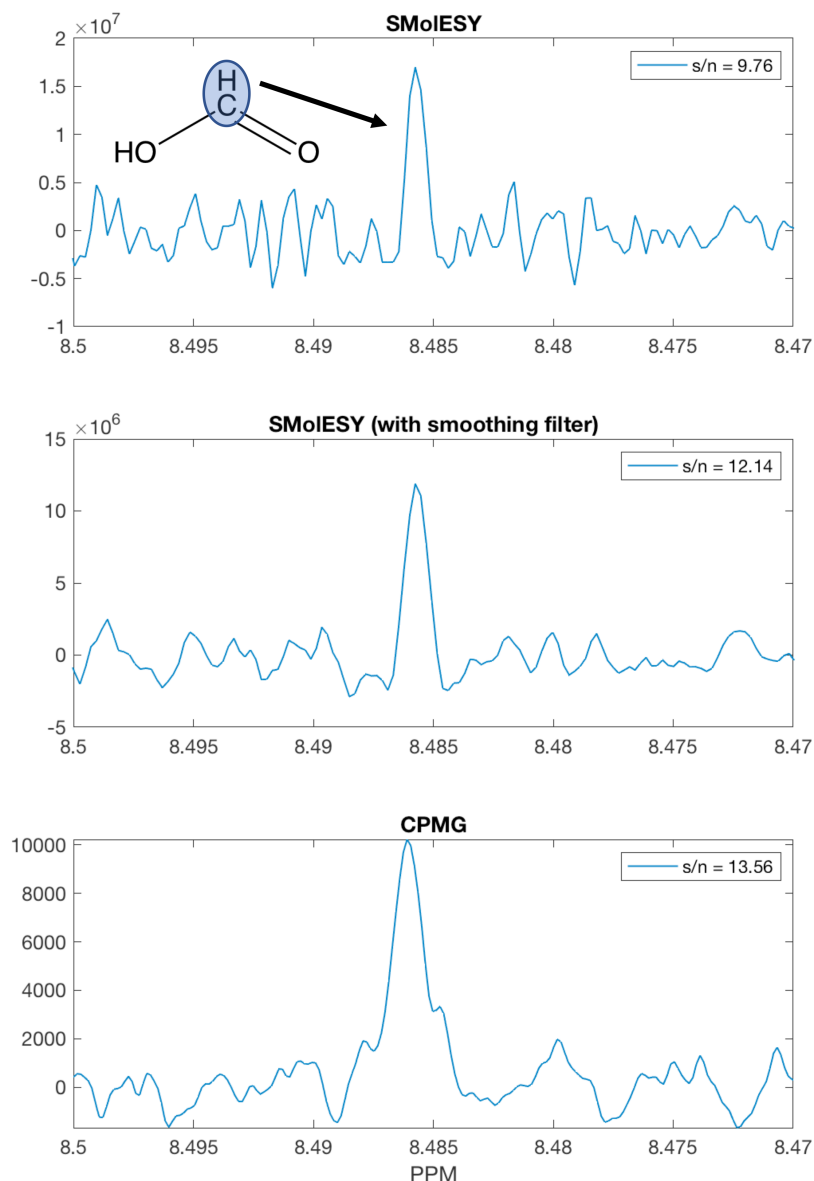


Figure S10. SMoIESY (upper panel), noise filtered SMoIESY (middle panel) and the CPMG (lower panel) spectral regions, focusing on the ^1H -NMR signal of the proton from Formic acid (at ~ 0.02 mM) in a plasma sample. The selected singlet resonates in usually noisy spectral region of a plasma ^1H -NMR profile, consequently a quite large s/n decrease is expected after its transformation from the 1D-NOESY spectrum. Indeed, SMoIESY spectrum shows the lowest s/n ($\sim 27\%$ decrease compared to CPMG), whereas the application of a simple lowpass filter (i.e. filter coefficients equal to the reciprocal of the window span) results in $\sim 18\%$ increase of s/n compared to pure SMoIESY signal, and exhibits quite similar s/n with CPMG ($\sim 10\%$ lower than CPMG).