

Resolving Severely Overlapping Ion Mobility Peaks Using Enhanced Fourier Self-Deconvolution

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November 14, 2024

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1 Derivation process of Equ.12	

Tom O'Haver¹ used a zero-centered Gaussian function (maximum at $t = 0$) as a deconvolving function, expressed as:

$$c(t) = \begin{cases} e^{-\frac{(t-t_{max})^2}{2\sigma^2}}, & t > \frac{t_{max}}{2} \\ e^{-\frac{t^2}{2\sigma^2}}, & t \leq \frac{t_{max}}{2} \end{cases} \quad (1)$$

where t_{max} represents the maximum value of t .

To address the asymmetric broadening observed in the IMS spectrum, this study introduces a modified deconvolving function $g(t)$ that combines Gaussian and Breit-Wigner functions for application in Fourier self-deconvolution. This newly defined function, represented in Equation 2, captures spectral asymmetry by centring its left and right components at t_{max} and 0, respectively. This configuration improves the model's capacity to handle asymmetrical broadening, thereby reducing errors commonly associated with tailing effects and enhancing overall deconvolution accuracy.

$$g(t) = \begin{cases} e^{-\frac{(t-t_{max})^2}{2\sigma^2}}, & t > \frac{t_{max}}{2} \\ \frac{w_{bw}^2}{w_{bw}^2 + t^2}, & t \leq \frac{t_{max}}{2} \end{cases} \quad (2)$$

According to the derived relationships, the half-width of the left (Gaussian) and right (Breit-Wigner) half-peaks, denoted as w_{gaus} and w_{bw} , respectively, is determined by the horizontal distance between the crest and the left (right) trough of the wavelet coefficient curve, represented by Δt_l and Δt_r .

$$w_{gaus} = \sqrt{\frac{2\ln 2}{3}} \Delta t_l \quad (3)$$

$$w_{bw} = \Delta t_r \quad (4)$$

For Gaussian functions, the half-width w_{gaus} relates to the standard deviation σ as follows:

$$w_{gaus} = \sigma \cdot \sqrt{2\ln 2} \quad (5)$$

By substituting Equation 3 into Equation 5, we derive:

$$\begin{aligned} \sigma &= \frac{w_{gaus}}{\sqrt{2\ln 2}} \\ &= \frac{\Delta t_l}{\sqrt{3}} \end{aligned} \quad (6)$$

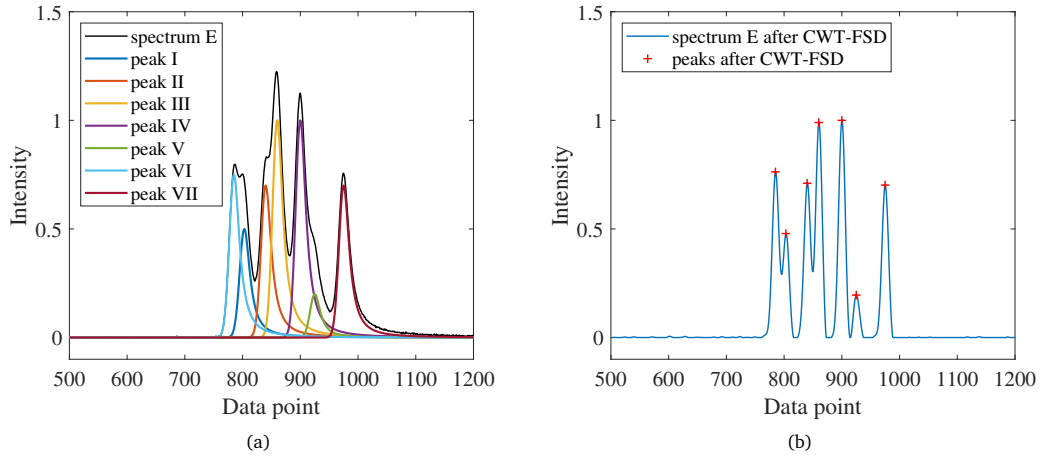
Finally, substituting Equations 6 and 4 into Equation 2 yields:

$$g(t) = \begin{cases} e^{-\frac{3(t-t_{max})^2}{2\Delta t_l^2}}, & t > \frac{t_{max}}{2} \\ \frac{\Delta t_r^2}{\Delta t_r^2 + t^2}, & t \leq \frac{t_{max}}{2} \end{cases} \quad (7)$$

2 Tables and Figures

Table S1 Parameters of spectra G and H discussed in this paper

Parameter	Spectrum G							Spectrum H						
	I	II	III	IV	V	VI	VII	I	II	III	IV	V	VI	VII
μ	664	675	715	724	744	755	804	664	675	714	726	744	755	804
w_l	5.25	5.03	5.79	5.02	4.97	5.57	4.67	4.90	5.18	5.51	5.12	5.19	4.85	5.01
w_r	5.89	6.58	5.48	6.09	5.86	5.40	6.37	6.71	6.39	6.32	6.76	6.25	6.85	6.92
h	0.28	0.77	1.00	0.87	0.59	0.28	0.16	1.00	0.86	0.27	0.78	0.58	0.28	0.16

**Figure S1** Spectrum E and the deconvolution results using CWT-FSD at a cutoff frequency of 150. (a) Spectrum E before deconvolution. (b) Use the combination of the Gaussian function and the Breit-Wigner function as the linear function for CWT-FSD.**Table S2** Deconvolution results for spectra G and H using different deconvolution functions at a cutoff frequency of 210

True value - Spectrum G	Estimated value / Absolute error		
	Function 1 ($w = 12.27$)	Function 2 ($w = 4.82 + 7.45$)	Function 3 ($w = 4.82 + 7.45$)
664	663 / -1	665 / 1	664 / 0
675	675 / 0	676 / 1	676 / 1
715	713 / -2	715 / 0	715 / 0
724	723 / -1	725 / 1	724 / 0
744	743 / -1	744 / 0	745 / 1
755	754 / -1	756 / 1	755 / 0
804	804 / 0	805 / 1	805 / 1
Misidentification	641, 652, 691, 701, 732, 764, 792	691	/

True value - Spectrum H	Estimated value / Absolute error		
	Function 1 ($w = 13.09$)	Function 2 ($w = 4.78 + 8.31$)	Function 3 ($w = 4.78 + 8.31$)
664	663 / -1	665 / 1	664 / 0
675	674 / -1	676 / 1	675 / 0
714	712 / -2	714 / 0	714 / 0
726	726 / 0	727 / 1	726 / 0
744	743 / -1	744 / 0	744 / 0
755	754 / -1	757 / 2	755 / 0
804	803 / -1	804 / 0	804 / 0
Misidentification	628, 639, 651, 686, 700, 765, 790	693	/

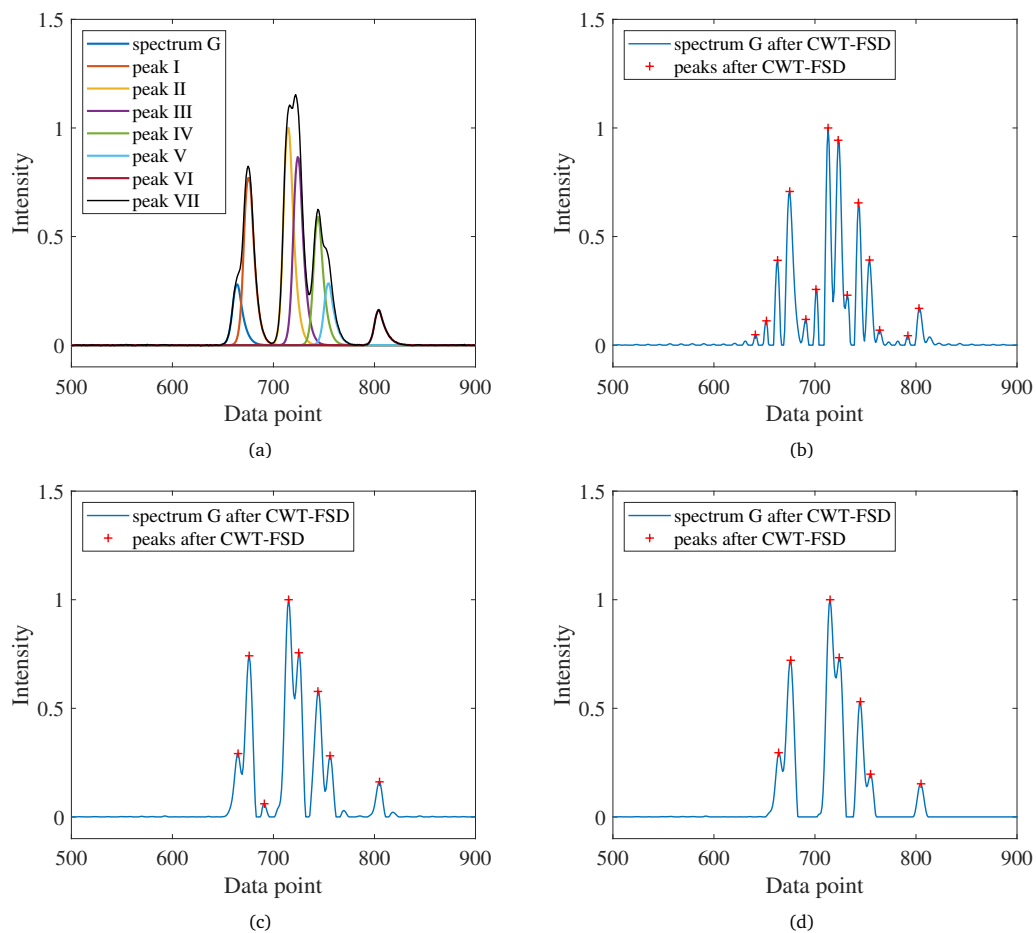


Figure S2 Spectrum G and the deconvolution results using different deconvolution functions at a cutoff frequency of 210. (a) Spectrum G before deconvolution. (b) Use the symmetric Gaussian function as the deconvolving function for CWT-FSD. (c) Use the asymmetric Gaussian function as the deconvolving function for CWT-FSD. (d) Use the combination of the Gaussian function and the Breit-Wigner function as the deconvolving function for CWT-FSD.

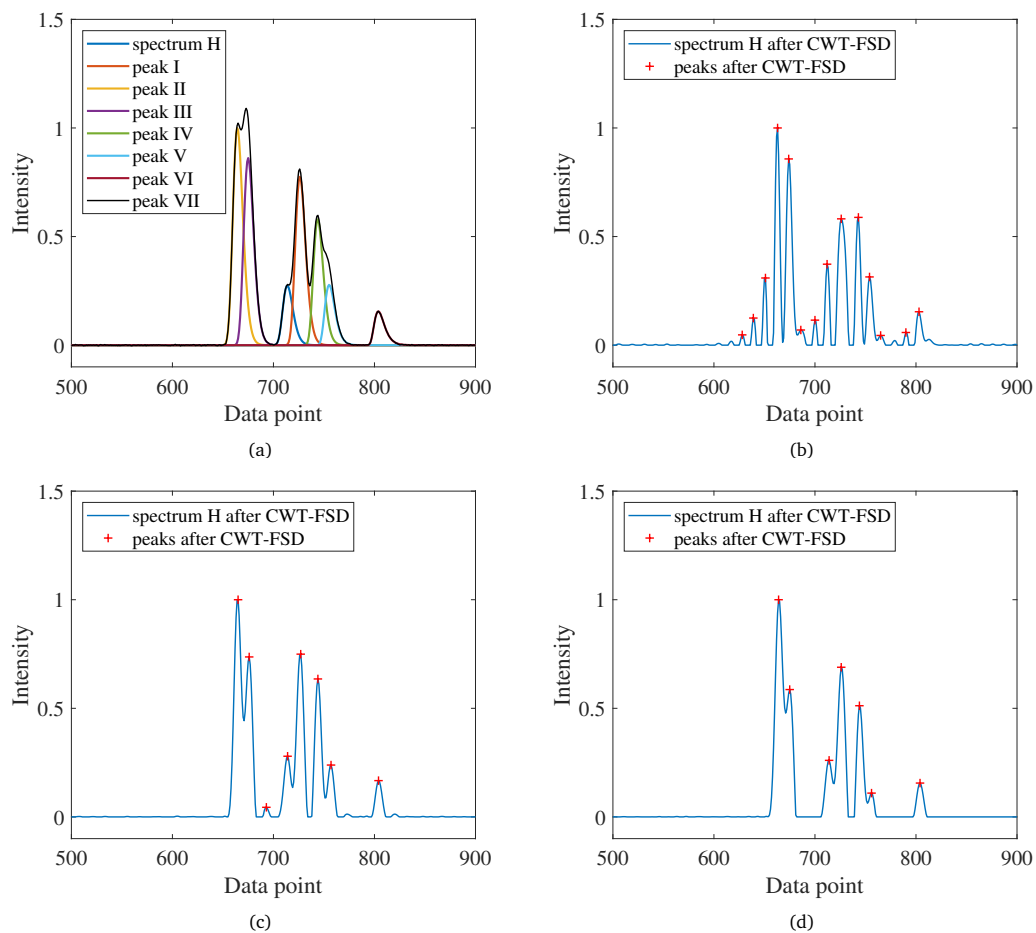


Figure S3 Spectrum H and the deconvolution results using different deconvolution functions at a cutoff frequency of 210. (a) Spectrum H before deconvolution. (b) Use the symmetric Gaussian function as the deconvolving function for CWT-FSD. (c) Use the asymmetric Gaussian function as the deconvolving function for CWT-FSD. (d) Use the combination of the Gaussian function and the Breit-Wigner function as the deconvolving function for CWT-FSD.

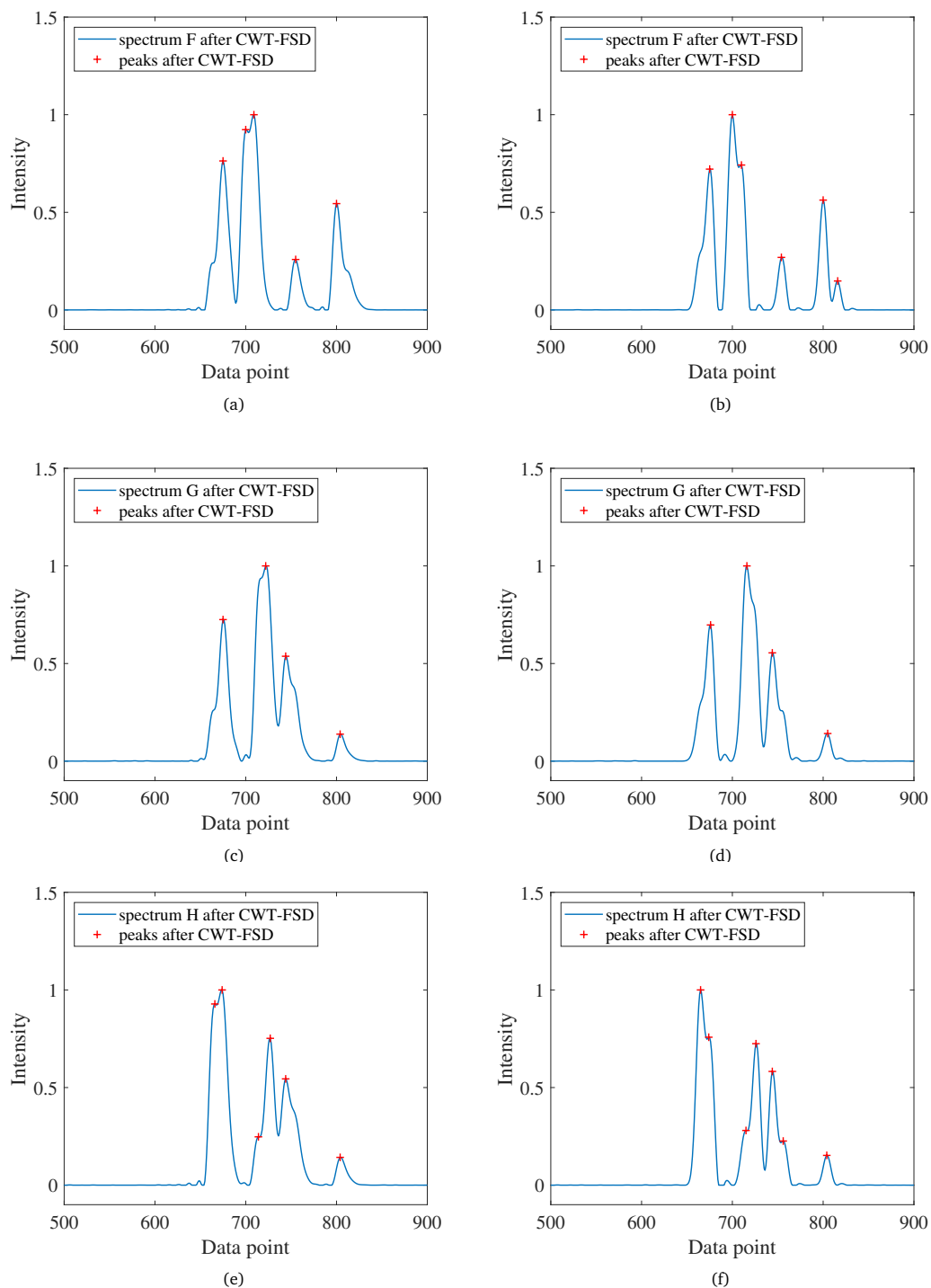


Figure S4 Deconvolution results of spectra F, G, and H at lower cut-off frequencies. (a) Spectrum F processed with a Gaussian function as the convolution function at a cut-off frequency of 130. (b) Spectrum F processed with an asymmetric Gaussian function at a cut-off frequency of 160. (c) Spectrum G processed with a Gaussian function at a cut-off frequency of 140. (d) Spectrum G processed with an asymmetric Gaussian function at a cut-off frequency of 160. (e) Spectrum H processed with a Gaussian function at a cut-off frequency of 130. (f) Spectrum H processed with an asymmetric Gaussian function at a cut-off frequency of 160.

Table S3 Results of different methods on spectrum F, G, and H

Spectrum F	True value	Estimated value / Absolute error		
		WFD	DWT-fitting-FSD	CWT-FSD
I	664	/	660 / -4	664 / 0
II	675	676 / 1	676 / 1	675 / 0
III	699	702 / 3	699 / 0	699 / 0
VI	710	708 / -2	710 / 0	710 / 0
V	754	753 / -1	755 / 1	754 / 0
VI	800	800 / 0	800 / 0	800 / 0
VII	814	812 / -2	817 / 3	815 / 1
Misidentification		673, 755	733, 780	/
Run time (s)		0.223	0.170	0.018
FWHM	5.12+6.62	39.71	18.92	4.88+8.53

Spectrum G	True value	Estimated value / Absolute error		
		WFD	DWT-fitting-FSD	CWT-FSD
I	664	/	660 / -4	664 / 0
II	675	676 / 1	676 / 1	676 / 1
III	715	718 / 3	/	715 / 0
VI	724	724 / 0	723 / -1	724 / 0
V	744	744 / 0	744 / 0	745 / 1
VI	755	750 / -5	/	755 / 0
VII	804	804 / 0	804 / 0	805 / 1
Misidentification		/	695	/
Run time (s)		0.213	0.170	0.014
FWHM	5.18+5.95	33.87	18.94	4.82+7.45

Spectrum H	True value	Estimated value / Absolute error		
		WFD	DWT-fitting-FSD	CWT-FSD
I	664	665 / 1	665 / 1	664 / 0
II	675	672 / -3	674 / -1	675 / 0
III	714	713 / -1	709 / -5	714 / 0
VI	726	725 / -1	726 / 0	726 / 0
V	744	744 / 0	745 / 1	744 / 0
VI	755	750 / -5	/	755 / 0
VII	804	804 / 0	804 / 0	804 / 0
Misidentification		/	644	/
Run time (s)		0.216	0.289	0.017
FWHM	5.11+6.60	45.95	20.66	4.78+8.31

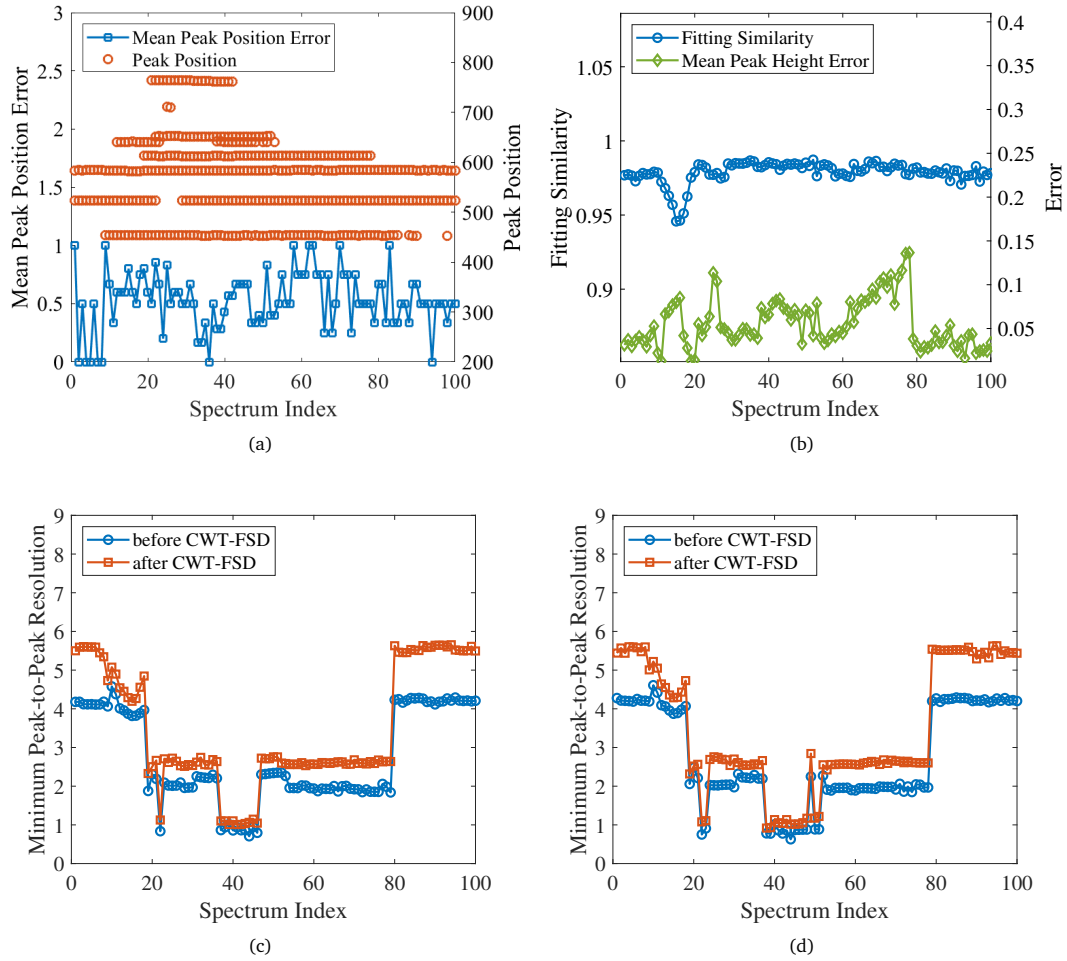


Figure S5 Results of human exhaled gas signals using CWT-FSD. (a) Peak position and average absolute error detected after deconvolution of spectral sequence B. (b) Peak height mean relative error and spectral similarity detected after deconvolution of spectral sequence B. (c) Change of minimum resolution before and after deconvolution of spectral sequence A. (d) Change of minimum resolution before and after deconvolution of spectral sequence B.

3 The code used in this article

The following code is used for processing spectral data using Continuous Wavelet Transform (CWT) and Fourier Self-Deconvolution (FSD) methods. Each function's input and output are specified below:

```
1
2 % Main Program
3 % This program performs Continuous Wavelet Transform (CWT) and Fourier Self-Deconvolution (FSD) on spectral data to detect peaks.
4 %
5 % Input: Spectral data from an external file.
6 % Output: Processed data with identified peak positions and plots.
7 %
8 % Function Descriptions:
9 % 1. Cwt_my - Performs Continuous Wavelet Transform.
10 %   - Input: s (array) - spectral data, N (int) - length of data.
11 %   - Output: L_l (float) - the weighted average horizontal distances between each peak and its corresponding left trough,
12 %             L_r (float) - the weighted average horizontal distances between each peak and its corresponding right trough.
13 %
14 % 2. FSD - Performs Fourier Self-Deconvolution.
15 %   - Input: s (array) - spectral data, N (int) - length of data,
16 %             L_l (float) - results of Cwt_my, L_r (float) - results of Cwt_my.
17 %   - Output: Ln_ (array) - deconvoluted data.
18 %
19 % 3. Peak_find - Detects peaks based on thresholds.
20 %   - Input: s (array) - spectral data, Ln_ (array) - wavelet coefficient curve, locs (array), locs1 (array), Ln_max (float).
21 %   - Output: locs_ (array) - detected peak locations.
22 %
23 % 4. findFuzzyMode - Calculates the fuzzy mode of an array within a tolerance.
24 %   - Input: values (array), tolerance (float).
25 %   - Output: fuzzyMode (float) - mode value within tolerance.
26 %
27 % 5. gaussian - Generates a Gaussian-shaped function.
28 %   - Input: x (array) - x-axis values, pos (float) - position of peak, wid (float) - width of peak.
29 %   - Output: g (array) - Gaussian function values.
30 %
31 % 6. bw - Generates a Breit-Wigner-shaped function.
32 %   - Input: x (array) - x-axis values, pos (float) - position of peak, wid (float) - width of peak.
33 %   - Output: g (array) - Breit-Wigner function values.
34 %
35 % 7. FourierFilter - Applies Fourier filtering to a signal based on specified frequency parameters.
36 %   - Input: xvector (array) - x-axis values, yvector (array) - y-axis (signal) values,
37 %             centerfrequency (float) - center frequency, filterwidth (float) - width of filter,
38 %             filtershape (int) - shape of the filter, filtermode (int) - mode of filter.
39 %   - Output: ry (array) - filtered signal.
40 %
41 % 8. shapefunction - determines the sharpness of the cut-off.
42 %   - Input: x (array) - x-axis values, pos (float) - peak position, wid (float) - width,
43 %             n (int) - shape type (0 for Lorentzian, 1 or more for Gaussian).
44 %   - Output: g (array) - function values.
45
46
47
48 clearvars
49 tic
50 s = importdata('C:/Users/19464/Desktop/FFT/WT/data.txt'); % Read data file
51 N = length(s);
52 t = 1:N;
53
54 [L_l, L_r] = Cwt_my(s, N); % Continuous Wavelet Transform, returns L_l and L_r
55 Ln_ = FSD(s, N, L_l, L_r); % Fourier Self-Deconvolution
56 Ln_ = Ln_';
57 Ln_max = max(Ln_);
58
59 for m = 1:length(Ln_)
60     if Ln_(m) < 0
61         Ln_(m) = 0; % Set negative values to zero
62     end
63 end
64
65 [pks, locs] = findpeaks(Ln_, 'minpeakheight', 0.05 * Ln_max); % Find peaks after deconvolution
66 locs_ = locs';
67 toc
68
69 % Plot deconvoluted curve and its peaks
70 hold on;
71 plot(t, Ln_, locs_, 'r+', 'linewidth', 1)
72 disp('Peaks after CWT-FSD processing:');
73 disp(locs_);
74 disp(Ln_(locs_));
75
76 legend({'Spectrum E after CWT-FSD', 'Peaks after CWT-FSD'}, 'Location', 'NorthWest');
77 ylim([-0.1 1.5])
78 xlabel('Data point');
79 ylabel({'Intensity'});
```

```

80
81 % Peak detection on original curve
82 L_max = max(s);
83 [pks2, locs2] = findpeaks(s, 'minpeakheight', 0.05 * L_max);
84 disp('Peaks detected by MATLAB built-in function:');
85 disp(locs2');
86 set(gcf, 'color', 'white');
87
88 % Function: Cwt_my
89 function [L_l, L_r] = Cwt_my(s, N)
90     % Filter the raw function
91     fs = N;
92     fc = 220; % Cut-off frequency for CWT (It can be adjusted according to the signal)
93     [b, a] = butter(4, fc / (fs / 2)); % 4th order low-pass filter
94     filtered_y = filtfilt(b, a, s); % Zero-phase filtering
95
96     % Continuous Wavelet Transform
97     p = 1:1;
98     scales = 2 .^ p; % scales as powers of 2
99     wname = 'gaus2'; % Using Gaussian wavelet
100     coefs = cwt(filtered_y, scales, wname);
101
102     % Find peaks and troughs in wavelet coefficients
103     L_max = max(coefs);
104     [pks_0, locs_0] = findpeaks(coefs);
105     [pks_1, locs_1] = findpeaks(L_max - coefs);
106     peaks = Peak_find(filtered_y, coefs, locs_0, locs_1, L_max); % Peak locations
107     Ln_ = max(coefs) - coefs;
108     Ln_max = max(Ln_);
109     [pks0, locs0] = findpeaks(Ln_);
110     [pks1, locs1] = findpeaks(Ln_max - Ln_);
111     troughs = Peak_find(filtered_y, Ln_, locs0, locs1, Ln_max); % Find troughs
112
113     left_differences = zeros(1, length(peaks));
114     right_differences = zeros(1, length(peaks));
115     peak_differences = zeros(1, length(peaks));
116     for i = 1:length(peaks)
117         % Find the nearest trough on the left of the current peak
118         left_troughs = troughs(troughs < peaks(i));
119         if ~isempty(left_troughs)
120             left_diff = peaks(i) - left_troughs(end);
121         else
122             left_diff = NaN; % If no left trough, set to NaN
123         end
124         left_differences(i) = left_diff;
125
126         % Find the nearest trough on the right of the current peak
127         right_troughs = troughs(troughs > peaks(i));
128         if ~isempty(right_troughs)
129             right_diff = right_troughs(1) - peaks(i);
130         else
131             right_diff = NaN; % If no right trough, set to NaN
132         end
133         right_differences(i) = right_diff;
134         if i < length(peaks)
135             peak_differences(i) = peaks(i + 1) - peaks(i);
136         end
137     end
138
139     disp('Differences between each peak and the nearest left trough:');
140     disp(left_differences);
141
142     disp('Differences between each peak and the nearest right trough:');
143     disp(right_differences);
144
145     disp('Differences between peaks:');
146     disp(peak_differences);
147     max_peak_diff = max(peak_differences);
148     if isnan(left_differences(1))
149         left_differences(1) = findFuzzyMode(left_differences, 2);
150     end
151
152     L_l = left_differences(1) * max_peak_diff;
153     peak_diff = max_peak_diff;
154     for j = 2:length(peaks)
155         if abs(left_differences(j) - left_differences(1)) < 5
156             L_l = L_l + left_differences(j) * peak_differences(j - 1);
157             peak_diff = peak_diff + peak_differences(j - 1);
158         end
159     end
160     L_l = L_l / peak_diff;
161     disp('Weighted average distance between each peak and its nearest left trough:');
162     disp(L_l);
163     if isnan(right_differences(length(peaks)))
164         right_differences(length(peaks)) = findFuzzyMode(right_differences, 2);

```

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165 end
166 peak_diff = max_peak_diff;
167 L_r = right_differences(length(peaks)) * max_peak_diff;
168 for j = 1:length(peaks) - 1
169     if abs(right_differences(j) - right_differences(length(peaks))) < 5
170         L_r = L_r + right_differences(j) * peak_differences(j);
171         peak_diff = peak_diff + peak_differences(j);
172     end
173 end
174 L_r = L_r / peak_diff;
175 disp('Weighted average distance between each peak and its nearest right trough:');
176 disp(L_r);
177 end
178
179 % Function to find peaks based on thresholds
180 function locs_ = Peak_find(s, Ln_, locs, locs1, Ln_max)
181     q = Ln_max * 0.1; % Height difference threshold between peaks and troughs
182     m = Ln_max * 0.1; % Peak height threshold
183     p = 1;
184     locs_ = [];
185     locs_(1) = 0;
186     if locs(1) > locs1(1) % First trough, then peak
187         for i = 2:length(locs) - 1
188             if Ln_(locs(i)) - Ln_(locs1(i)) > q || Ln_(locs(i)) - Ln_(locs1(i + 1)) > q
189                 if Ln_(locs(i)) > m && locs(i) - locs_(max(1, p - 1)) > 4
190                     locs_(p) = locs(i);
191                     p = p + 1;
192                 end
193             end
194         end
195     elseif locs(1) < locs1(1) % First peak, then trough
196         for i = 2:length(locs) - 1
197             if Ln_(locs(i)) - Ln_(locs1(i)) > q || Ln_(locs(i)) - Ln_(locs1(i - 1)) > q
198                 if Ln_(locs(i)) > m && locs(i) - locs_(max(1, p - 1)) > 4
199                     locs_(p) = locs(i);
200                     p = p + 1;
201                 end
202             end
203         end
204     end
205 end
206
207 % Function to find the fuzzy mode of an array
208 function fuzzyMode = findFuzzyMode(values, tolerance)
209     % Remove NaN values
210     values = values(~isnan(values));
211
212     % Calculate fuzzy mode
213     unique_vals = unique(values);
214     counts = zeros(size(unique_vals));
215     for i = 1:length(unique_vals)
216         counts(i) = sum(abs(values - unique_vals(i)) <= tolerance);
217     end
218     % Find the value with maximum frequency
219     [~, maxIdx] = max(counts);
220     fuzzyMode = unique_vals(maxIdx);
221
222 end
223
224 function xe = FSD(s, N, L_l, L_r)
225     % Set parameters for Fourier Self-Deconvolution
226     FrequencyCutoff = 150; % Frequency cutoff (simulation: 210, real: 165, Gaussian-BW simulation: 150) (It can be adjusted according to
227     % the signal)
228     CutOffRate = 1;
229     DA = 1;
230     index = 1:N;
231
232     % Calculate Gaussian and Breit-Wigner width parameters
233     % Theoretical value
234     dw_1 = 0.6798 * (L_l) * 2;
235     dw_2 = 1 * L_r * 2;
236     % True value (can be used in the deconvolution of the true spectrum)
237     % dw_1 = 0.65*(L_l)*2;
238     % dw_2 = 1.09*L_r*2;
239
240     dw = (dw_1 + dw_2) / 2;
241     F = s';
242
243     % Combine Gaussian and Breit-Wigner function
244     df = bw(index, min(index), dw_2) + gaussian(index, max(index), dw_1);
245
246     % Asymmetric Gaussian function
247     % df=gaussian(index,min(index),dw_2)+gaussian(index,max(index),dw_1);
248     % Symmetric Gaussian function
249     % df=gaussian(index,min(index),dw)+gaussian(index,max(index),dw);

```

```

249
250 % Perform Fourier Transform and Inverse Fourier Transform for FSD
251 ca5_FFT = fft(df);
252 X_iftt = ifft(fft(F) ./ (ca5_FFT + DA * 0.01 * max(ca5_FFT))) * sum(df);
253 syDA = FourierFilter(F, X_iftt, 0, FrequencyCutoff, CutOffRate, 1);
254
255 % Normalize the result
256 xe = syDA / max(syDA);
257 end
258
259 function g = gaussian(x, pos, wid)
260 % Generate a Gaussian-shaped function
261 % g = exp(-((x - pos) / (0.60056120439323 * wid))^ 2);
262 g = exp(-((x - pos) / (0.60056120439323 * wid)) .^ 2);
263 end
264
265 function g = bw(x, pos, wid)
266 % Generate a Breit-Wigner-shaped function
267 w_bw = wid * 0.5;
268 h = 1;
269 g = (w_bw .^ 2 * h) ./ (w_bw .^ 2 + (x - pos) .^ 2);
270 end
271
272 function ry = FourierFilter(xvector, yvector, centerfrequency, filterwidth, filtershape, filtermode)
273 % Computes a Fourier filter for a signal
274 % Centerfrequency and filterwidth specify the frequency range of the pass band.
275 % filtershape determines the sharpness of the cutoff, and filtermode specifies the filter type.
276
277 % Adjust x and y vectors to row format
278 xvector = reshape(xvector, 1, length(xvector));
279 yvector = reshape(yvector, 1, length(yvector));
280 fy = fft(yvector);
281 lft1 = 1:(length(fy) / 2);
282 lft2 = (length(fy) / 2 + 1):length(fy);
283
284 % Compute filter shape based on filtermode
285 if filtermode == 1 % 'Band-pass'
286     ffilter1 = shapefunction(lft1, centerfrequency + 1, filterwidth, filtershape);
287     ffilter2 = shapefunction(lft2, length(fy) - centerfrequency + 1, filterwidth, filtershape);
288     ffilter = [ffilter1, ffilter2];
289 elseif filtermode == 2 % 'High-pass'
290     centerfrequency = length(xvector) / 2;
291     ffilter1 = shapefunction(lft1, centerfrequency + 1, filterwidth, filtershape);
292     ffilter2 = shapefunction(lft2, length(fy) - centerfrequency + 1, filterwidth, filtershape);
293     ffilter = [ffilter1, ffilter2];
294 elseif filtermode == 3 % 'Low-pass'
295     centerfrequency = 0;
296     ffilter1 = shapefunction(lft1, centerfrequency + 1, filterwidth, filtershape);
297     ffilter2 = shapefunction(lft2, length(fy) - centerfrequency + 1, filterwidth, filtershape);
298     ffilter = [ffilter1, ffilter2];
299 elseif filtermode == 4 % 'Band-reject (notch)'
300     ffilter1 = shapefunction(lft1, centerfrequency + 1, filterwidth, filtershape);
301     ffilter2 = shapefunction(lft2, length(fy) - centerfrequency + 1, filterwidth, filtershape);
302     ffilter = 1 - [ffilter1, ffilter2];
303 elseif filtermode == 5 % 'Comb pass'
304     n = 2;
305     ffilter1 = shapefunction(lft1, centerfrequency + 1, filterwidth, filtershape);
306     ffilter2 = shapefunction(lft2, length(fy) - centerfrequency + 1, filterwidth, filtershape);
307     while n < 50
308         ffilter1 = ffilter1 + shapefunction(lft1, n * (centerfrequency + 1), filterwidth, filtershape);
309         ffilter2 = ffilter2 + shapefunction(lft2, length(fy) - n * (centerfrequency + 1), filterwidth, filtershape);
310         n = n + 1;
311     end
312     ffilter = [ffilter1, ffilter2];
313 elseif filtermode == 6 % 'Comb notch'
314     n = 2;
315     ffilter1 = shapefunction(lft1, centerfrequency + 1, filterwidth, filtershape);
316     ffilter2 = shapefunction(lft2, length(fy) - centerfrequency + 1, filterwidth, filtershape);
317     while n < 50
318         ffilter1 = ffilter1 + shapefunction(lft1, n * (centerfrequency + 1), filterwidth, filtershape);
319         ffilter2 = ffilter2 + shapefunction(lft2, length(fy) - n * (centerfrequency + 1), filterwidth, filtershape);
320         n = n + 1;
321     end
322     ffilter = 1 - [ffilter1, ffilter2];
323 end
324
325 if length(fy) > length(ffilter)
326     ffilter = [ffilter ffilter(1)];
327 end
328 ffy = fy .* ffilter; % Apply the filter in the frequency domain
329 ry = real(ifft(ffy));
330 end
331
332 function g = shapefunction(x, pos, wid, n)
333 % Generates a shaped peak function (Gaussian or Lorentzian)

```

```

334 % Shape is Lorentzian (1/x^2) when n=0, Gaussian (exp(-x^2)) when n=1,
335 % and becomes more rectangular as n increases.
336 if n == 0
337     g = ones(size(x)) ./ (1 + ((x - pos) / (0.5 * wid)) .^ 2);
338 else
339     g = exp(-((x - pos) / (0.6 * wid)) .^ (2 * round(n)));
340 end
341 end
342

```

Notes and references

- [1] T. O'Haver, A pragmatic introduction to signal processing, University of Maryland at College Park (1997).