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ENVELOPE ESTIMATION BY TANGENTIALLY **CONSTRAINED SPLINE**

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Introduction

- An envelope of a signal is a smooth wrapping curve containing much information about the signal. →it is widely considered in signal processing. (e.g., empirical mode decomposition (EMD)).
- The accuracy of EMD is determined by the accuracy of envelope estimation because subtraction of the mean of upper and lower envelopes is the central step in the process of EMD.
- Conventional envelope estimation method is interpolating all extrema using the cubic C^2 -spline interpolation. It can obtain smooth envelopes, but often causes the undershoot problems. \rightarrow **leads to erroneous results in EMD.**
- In this paper, tangentially constrained spline together with a gradient-based tangential points optimization method is proposed.

-Spline functions and cubic C²-spline interpolation-

Spline functions

• A spline is a continuous function defined by piecewise polynomials that has been widely used in signal processing owing to its flexibility and

<u>Cubic C²-spline interpolations</u>

• Finding a cubic C^2 -spline function passing through all given points (x_k, y_k) is characterized by

Mean envelope

optimality. $s_k(t)$: Polynomials. $s_0(t) \circ s_1(t)$

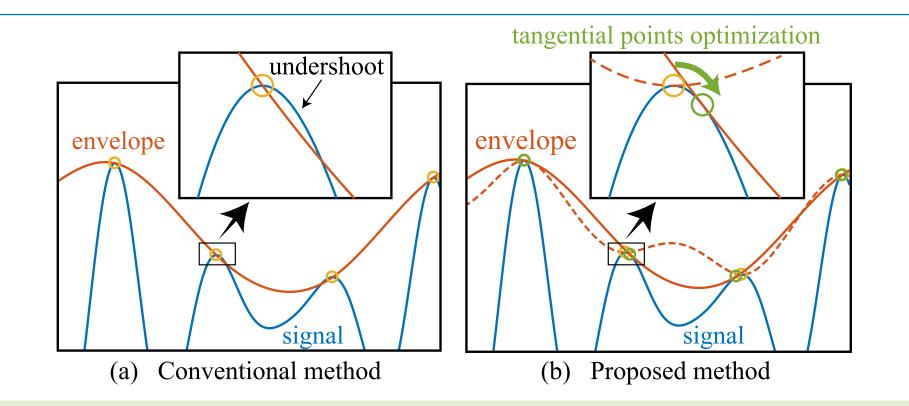
• These polynomials are different, but each polynomial is **connected continuously**.

Proposed method

- In order to estimate the envelope without undershoot problem while maintaining smoothness, a tangentially constrained spline with tangential points optmization is proposed.
- Tangentially constrained spline \rightarrow An estimated envelope is strictly tangent at interpolation points. • **Tangential points optmization** \rightarrow An estimated envelope is **optimally smooth**.

$\underset{s \in S_3^2(\mathbf{x})}{\text{minimize}} \quad \frac{1}{2} \int_{x_0}^{x_n} |s''(t)|^2 dt \quad \text{subject to} \quad s(x_k) = y_k \text{ for all } k.$

 $S_3^2(\mathbf{x})$: Set of all spline functions of degree 3 and C^2 -continuous.



Lower envelope

Upper envelope

Tangentially constrained spline

• Tangentially constrainde spline s_{TC} : A quartic C^2 -spline function constrained with first derivatives

 $\underset{s_{\mathrm{TC}}\in S_{4}^{2}(\boldsymbol{\tau})}{\text{minimize}} \quad \frac{1}{2} \int_{\tau_{0}}^{\tau_{n}} |s_{\mathrm{TC}}''(t)|^{2} dt \quad \text{subject to} \quad s_{\mathrm{TC}}(\tau_{k}) = u(\tau_{k}), \ s_{\mathrm{TC}}'(\tau_{k}) = u'(\tau_{k}) \quad \text{for all } k.$

u(t) : Signal. $S_4^2(\boldsymbol{\tau})$: Set of all spline functions of degree 4 and C^2 -continuous. $\boldsymbol{\tau} = [\tau_0, \tau_1, \dots, \tau_n]^T$: Tangential points.

 $\mathbf{s} = [\mathbf{z}^T, \mathbf{p}^T, \mathbf{P}^T]^T, \ \mathbf{z} = [z_0, \dots, z_n]^T, \ \mathbf{p} = [p_0, \dots, p_n]^T, \ \mathbf{P} = [P_0, \dots, P_n]^T, \ z_k = s_{\mathrm{TC}}(\tau_k), \ p_k = s_{\mathrm{TC}}'(\tau_k), \ P_k = s_{\mathrm{TC}}'(\tau_k), \ h_k = \tau_{k+1} - \tau_k$

Tangential points optimization

- Smoothness of tangentially constrained spline depends on the choice of τ .
- \rightarrow The tangentially constrained spline is smooth only when appropriate tangential points are chosen as the interpolation points.
- Optimization problem of tangential points is formulated as

 $\underset{\boldsymbol{\tau} \in \mathbb{R}^{n+1}}{\text{minimize}} \quad I(\boldsymbol{\tau}) = \frac{1}{2} \mathbf{s}^{\star}(\boldsymbol{\tau})^T \mathbf{A}(\boldsymbol{\tau}) \mathbf{s}^{\star}(\boldsymbol{\tau}).$ s^{\star} : The solution of Eq. (*).

It can be rewritten as

 $\underset{\boldsymbol{\tau} \in \mathbb{R}^{n+1}}{\text{minimize}} \quad \frac{1}{2} \boldsymbol{\xi}^T \tilde{\mathbf{A}}(\boldsymbol{\tau}) \boldsymbol{\xi} \quad \text{subject to} \quad \mathbf{K}(\boldsymbol{\tau}) \boldsymbol{\xi} - \tilde{\mathbf{b}}(\boldsymbol{\tau}) = \mathbf{0}.$

•
$$\int_{\tau_{k}}^{\tau_{k+1}} |s_{\mathrm{TC}}'(t)|^{2} dt = \frac{6}{5h_{k}} \Big(p_{k} - p_{k+1} + \frac{h_{k}}{12} (P_{k} + P_{k+1}) \Big)^{2} \\ + \frac{h_{k}}{24} \Big(3P_{k}^{2} - 2P_{k}P_{k+1} + 3P_{k+1}^{2} \Big), \\ = \mathbf{s}_{k}^{T} \mathbf{A}_{k} \mathbf{s}_{k}$$
• $s_{\mathrm{TC}}(\tau_{k}) = u(\tau_{k}), \ s_{\mathrm{TC}}'(\tau_{k}) = u'(\tau_{k}) \\ \Leftrightarrow \mathbf{Es} = \mathbf{b}$

This problem can be descritized as

$$\underset{\mathbf{s}\in\mathbb{R}^{3n+3}}{\text{ninimize}} \quad \frac{1}{2}\mathbf{s}^{T}\mathbf{A}\,\mathbf{s} \text{ subject to } \mathbf{E}\mathbf{s} = \mathbf{b}. \qquad \cdots (\mathbf{s}^{n+1})^{T}\mathbf{A}\,\mathbf{s} \mathbf{s}^{T}\mathbf{A}\,\mathbf{s} \mathbf{s}^{T}\mathbf{A}\,\mathbf{s} \mathbf{s}^{T}\mathbf{s}^$$

 Its solution is obtained by solving Karush–Kuhn–Tucker system, $\mathbf{K}\boldsymbol{\xi} - \tilde{\mathbf{b}} = \mathbf{0}. \qquad \boldsymbol{\xi} = \begin{bmatrix} \mathbf{s} \\ \boldsymbol{\nu} \end{bmatrix}, \quad \mathbf{K} = \begin{bmatrix} \mathbf{A} & \mathbf{E}^T \\ \mathbf{E} & \mathbf{O} \end{bmatrix}, \quad \tilde{\mathbf{b}} = \begin{bmatrix} \mathbf{0} \\ \mathbf{b} \end{bmatrix}$

$\tilde{\mathbf{A}} = \begin{bmatrix} \mathbf{A} & \mathbf{O} \\ \mathbf{O} & \mathbf{O} \end{bmatrix}, \quad \boldsymbol{\xi}^T \tilde{\mathbf{A}}(\boldsymbol{\tau}) \, \boldsymbol{\xi} = \begin{bmatrix} \mathbf{s} \\ \boldsymbol{\nu} \end{bmatrix}^T \begin{bmatrix} \mathbf{A} & \mathbf{O} \\ \mathbf{O} & \mathbf{O} \end{bmatrix} \begin{bmatrix} \mathbf{s} \\ \boldsymbol{\nu} \end{bmatrix} = \mathbf{s}^T \mathbf{A}(\boldsymbol{\tau}) \, \mathbf{s}$

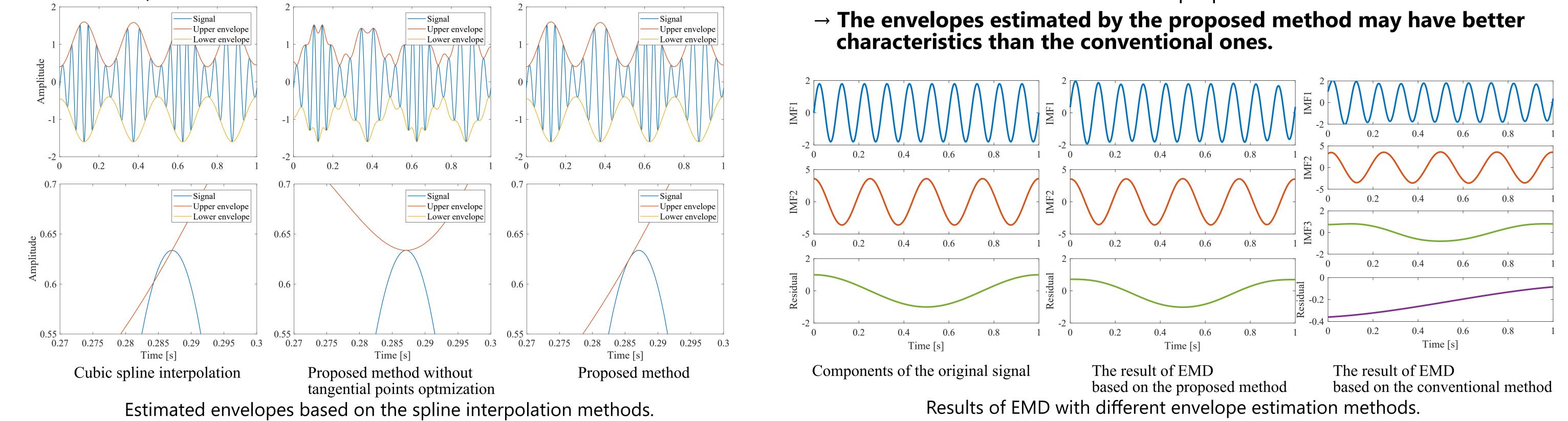
 In order to solve the problem using a gradient-based optimization method, gradient $\nabla_{\tau} I(\tau)$ is calculated by the following steps through the adjoint-state method:

> Step 1. Solve $K\xi^* = b$. Step 2. Solve $K\lambda^* = \tilde{A}\xi^*$. Step 3. Calculate the gradient by $\nabla_{\boldsymbol{\tau}} I(\boldsymbol{\tau}) = \frac{1}{2} \mathbf{D}_{\mathbf{A}}^{T} \boldsymbol{\xi}^{\star} - \left(\mathbf{D}_{\mathbf{K}} - \frac{\partial \tilde{\mathbf{b}}}{\partial \boldsymbol{\tau}} \right)^{T} \boldsymbol{\lambda}^{\star}. \quad \mathbf{D}_{\mathbf{A}} = \left[\frac{\partial \tilde{\mathbf{A}}}{\partial \tau_{0}} \boldsymbol{\xi}, \dots, \frac{\partial \tilde{\mathbf{A}}}{\partial \tau_{n}} \boldsymbol{\xi} \right], \quad \mathbf{D}_{\mathbf{K}} = \left[\frac{\partial \mathbf{K}}{\partial \tau_{0}} \boldsymbol{\xi}, \dots, \frac{\partial \mathbf{K}}{\partial \tau_{n}} \boldsymbol{\xi} \right].$

Numerical experiment

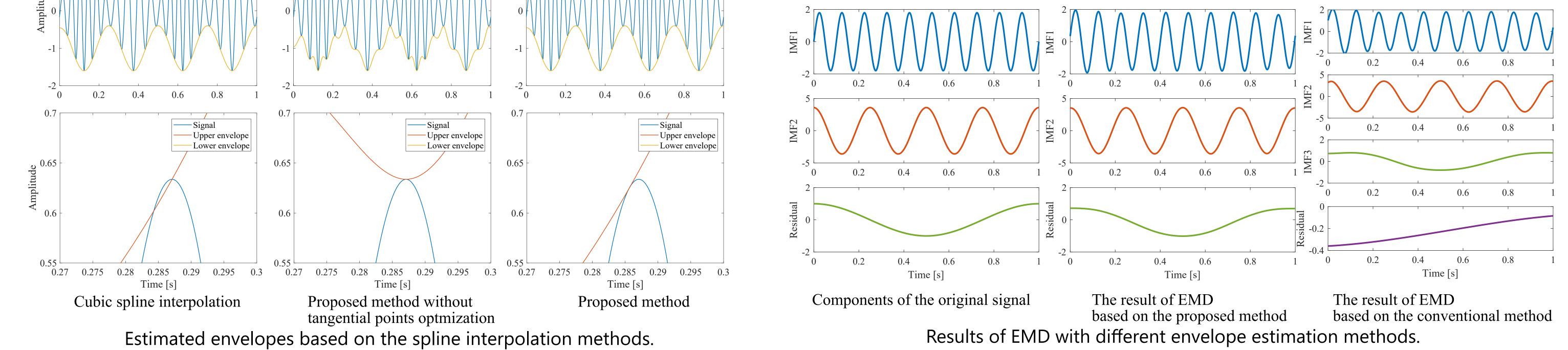
<u>Comparison of estimated envelopes</u>

- An envelope estimation problem of a simulated signal was considered.
- The proposed method can estimate a smooth tangential envelope without undershoot problem.



<u>Application of the EMD</u>

- EMD was applied to a signal consisting of three components.
- EMD with the proposed method correctly recovered the original components because undershoot is avoided in the proposal.



Conclusion

- A tangentially constrained spline, which is a quartic C^2 -spline constrained by first derivatives at the tangential points is proposed for estimating envelopes without undershoot problem.
- A tangential points optimization method is also proposed so that an optimally smooth envelope among the proposed splines is obtained.
- Future works include considerations of appropriate boundary conditions.